

LIBRARY Michigan State University \mathbf{r}

This is to certify that the

dissertation entitled

The Simulation of Combine Harvester Performance as Affected by Bulk Crop Properties

presented by

Wilbur T. Mahoney III

has been accepted towards fulfillment of the requirements for

Ph.D. degree in Agric. Engr. Tech.

Augil K. Snieuwstanas/2/2/28

Date 2/2/88

MSU is an Affirmative Action/Equal Opportunity Institution

Ť

 $\bar{1}$

 $\overline{}$

 ~ 100 km s $^{-1}$

MART T RETURNING MATERIALS: Place in book drop to LIBRARIES | remove this checkout from **EXAMPLE 2001 YOUR RECORD.** FINES will MSU

MSU

Place in book drop to

LIBRARIES

your record. FINES when the charged if book is returned after the date MSU

Place in book drop to

LIBRARIES

LIBRARIES

Place in book drop to

your record. FINES will

be charged if book is

returned after the date

stamped below. stamped below.

THE SIMULATION OF COMBINE HARVESTER PERFORMANCE AS AFFECTED BY BULK CROP PROPERTIES

 \overline{a}

By

Wilbur 'Ihomas Mahoney, III

A DISSERTATION

Submitted to Michigan State university in partial fulfillment of the requirements for the degree of

DOCTOR OF PHILOSOPHY

Department of Agricultural Engineering

Copyright by WILBUR THOMAS MAHONEY, ...__
1988 III

 $\bar{\star}$

 $\frac{1}{2}$

 $\ddot{\bullet}$

ABSTRACT

THE SIMULATION OF COMBINE HARVESTER PERFORMANCE AS AFFECTED BY BULK CROP PROPERTIES

By

Wilbur Thomas Mahoney, III

Physical changes in crop properties have been reported to affect the operational characteristics of combine harvesters. Performance, measured as mass other than grain (M.O.G.) feed rates at fixed grain loss levels, varies as a result of crop property changes (Stephens and Babe, 1977). Combine performance may vary substantially during the course of a single day making comparisons of performance between combines in field tests difficult.

A research project was undertaken to determine the effect of bulk crop properties on combine performance. This thesis describes the collection and measurement of bulk crop properties, the correlation between property changes and combine performance, and the development of .a combine simulation model based on bulk crop properties.

Fourteen bulk crop properties were collected for wheat and barley from 1980 through 1984. In addition, the performance of the cleaning component and the straw walker was also measured on a conventional type John Deere 6620 combine. Crop properties were measured on grain, chaff, and straw components of wheat and barley. The performance criteria was chaff feed rate at 0.5 percent cleaner grain loss and total M.O.G. (mass other than grain) feed rate at 1.0 percent walker grain loss.

Grain density, grain angle of repose, chaff density, chaff coefficient of friction, chaff canpressibility nodulus, grain:M.O.G. ratio, straw density, straw canpressibility modulus, and straw coefficient of friction were shown to affect oanbine performance. In general, cleaner performance appeared to be three time more sensitive to crop changes than strawwwalker performance.

Cleaner performance and straw'walker performance prediction equations were developed which explained 92.0 percent of the variation in cleaning performance and 30.0 percent of the variation of the performance in straw'walker performance.

A.computer simulation model was developed using crop property and combine performance data. The model predicted cleaner, walker, and overall processing performance as functions of ground speed, width of cut, crop yield, and a set of crop properties which vary in a stochastic manner. Implemented as an interactive program, the user specifies initial crop properties and variability. Each property is then simulated over a range of selected moisture conditions. The model can be used by students and test engineers to study the effects of crop properties on combine performance. appeared to be three time more sensitive to
er performance.
Straw walker performance prediction
ch explained 92.0 percent of the variation
30.0 percent of the variation of the
performance.
codel was developed using crop p

1 Ajit R. Snivastava 2/2/88 \overrightarrow{A} jit K \overrightarrow{V} Srivastava ,_ . //f>/ (_

 $22/2/8$ Donald M. Edwards

ACKNOWLEDGMENTS

I wish to thank the following:

The Department of Agricultural Engineering and its head, Dr. Donald Edwards, for accepting me in the Agricultural Engineering Technology program.

Dr. Ajit K. Srivastava, my major professor, for his guidance, instruction, and assistance during my tenure as a graduate research assistante.

Drs. Thomas Burkhardt, Ivan Mao, and Alan Rotz for their guidance and assistance.

Deere and Company Harvester Works of Moline, Illinois for their financial support and assistance.

My wife for her enduring support.

My fellow graduate students, especially Steve Richey, my office partner and co-worker, for his help and advice.

iv

TABLE OF CONTENTS

 \mathcal{L}

LIST OF TABLES

LIST OF FIGURES

CHAPTER I

INTRODUCTION AND PROBLEM STATEMENT

1.1 <u>Background</u> 1.1 Background

combine harvester performance varies throughout the harvesting period. Performance shifts often occur within the same harvest day. Stephens and Babe (1977) reported performance decreased 30 percent in less than one week while conducting combine performance tests. During that week, grain moisture remained constant while straw moisture fell from 35 percent to 10 percent. Although the performance shift was not quantitatively explained much of the variation in performance was attributed to the change in straw moisture contrary to the popular belief that performance improves as straw moisture decreases. Research has been conducted to evaluate design changes or to compare machines in performance tests but little has been done to determine the effects of the crop parameters on combine performance.

Combine manufacturers conduct field and laboratory tests with prototype harvesters to determine if design changes significantly alter combine performance. Suspected crop property shifts often make comparisons between prototypes and production machines difficult. Presently, it is necessary to generate a performance curve for a production machine each time a prototype machine is operated to eliminate variability caused by the crop. Stephens and Rabe (1977) estimated 30—50 percent of testing time is spent generating the performance curve for the production machine.

Components are often tested with stored plant material in a laboratory setting. Performance of combines in field tests is not often

 $\mathbf{1}$

duplicated using stored plant material in a laboratory setting. The failure to duplicate field tests has been attributed to changes in properties of the crop as a result of storage.

1.2 Problem Statement

The goal of field and laboratory testing is to produce valid and repeatable results. Changes in crop properties are thought to cause combine performance shifts. A.method is needed to measure crop bulk properties and to explain the variation in combine performance as a function of crop properties. Ideally, a model of a combine harvester could be constructed such that production machine performance could be predicted thus minimizing the need for labor and cost intensive field performance curves.

1.3 Objectives

The purpose of this dissertation is to relate crop property changes to variations in combine harvester performance. The specific objectives are:

- 1) to identify and measure bulk crop property changes,
- 2) to correlate bulk crop property changes to combine harvester performance,
- 3) to predict losses as a function of crop properties, and
- 4) to develop a computer simulation of a combine harvester based on bulk crop property changes.

CHAPTER II

REVIEW OF LITERATURE

2.1 <u>Introduction</u> 2.1 Introduction

Almost all seed and grain crops in the united States are harvested by combine harvesters. Conventional combines, those with a threshing component, straw walker, and cleaning component are the focus of this review; An overview of combine harvest methods, terminology, operating principles, and the factors reported to affect combine performance are presented. CHA
REVIEW OF 1
2.1 <u>Introduction</u>
Almost all seed and grain cro
by combine harvesters. Convention
component, straw walker, and clean
review. An overview of combine ha
principles, and the factors report
presented.
2.2 <u>Har</u>

2.2 Harvest Methods and Principles

Two primary harvest methods are used for harvesting small grain crops. Direct cutting, the most common harvest method, involves cutting the standing crop and processing in a continuous operation. Processing in this case refers to threshing, separation, and cleaning. 'Windrowing, another harvest method, is also used in some parts of the united States. The practice is common in the Northern United States and Western Canada. Although windrowing requires an extra operation, the practice facilitates curing in areas where drying conditions are variable. Generally, conventional combines are of two types: self-propelled and pull-type. Self-propelled combines can be further categorized by three machine types: level land, hillside, and sidehill machines. Level-land combines are intended for use on level or nearly level land. Hillside machines, as the name implies, are designed to allow combining on hillsides. Hillside machines are equipped with automatic leveling devices that allow the machine to remain horizontal while the cutting mechanism follows the contour of the ground. Hillside combines are

 $\overline{\mathbf{3}}$

designed to operate on maximum slopes of 30-45 percent. Sidehill combines are essentially level land combines with slightly altered components which do not allow the material distribution to overload a component. For example the cleaning unit is equipped with baffles to assure even material distribution. Pull-type combines are usually PTO driven versions of level land machines. Their widest application is in those areas that windrow small grains. They are less expensive and less mneuverable than self-propelled machines. Also, pull-type combines do not provide the continuous operator adjustment available on self-propelled combines.

Combines perform four basic operations during the harvesting process:

- 1. Cutting standing plants or picking up the windrow,
- 2. Threshing,
- 3. Separation, and
- 4. Cleaning.

A conventional combine is equipped with several components to accomplish the basic operations of harvesting (Figure 1).

When direct cutting standing plants a feeding and cutting mechanism called a header is employed. The mechanism's principle components are a reel, a sickle bar cutter, and a conveying system. As the combine moves into a standing crop, the reel momentarily holds the plant in place for cutting and then directs the crop rearward for conveying into the feeder house.

The threshing component of the cylinder-concave is comprised of a drum or cylinder partially enclosed by a concave or grate mounted perpendicular to the crop flow. The concave is slotted such that grain

may fall through it at separation. As the cylinder spins, crop is fed between the cylinder and concave. Rasping, squeezing, and impact between the revolving cylinder and stationary concave detach the grain from the plant.

The straw walker or rack is typically an oscillating bed which separates grain from straw as it agitates the crop material rearward. Grain and chaff fall through openings in the rack onto grain return pans which lead to the cleaning component. The remaining material, predominantly straw, is carried out of the machine by the straw walkers.

The cleaning component, usually comprised of a chaffer sieve, a cleaning sieve, and a fan, is the final step in separation. Air directed on the mixture of grain, chaff, and broken straw carries the lighter debris out of the machine while the grain which is more dense falls through the sieves. The cleaned grain is then augured into the combine holding tank. Grain and chaff fall t
pans which lead to the
predominantly straw, is
The cleaning compo
cleaning sieve, and a f
directed on the mixture
lighter debris out of t
falls through the sieve
combine holding tank.
2.3 Combine Per separates grain from straw as
Grain and chaff fall through
pans which lead to the cleani
predominantly straw, is carri
The cleaning component,
cleaning sieve, and a fan, i
directed on the mixture of gr
lighter debris out o

2.3 Combine Performance

2.3.1 Performance Measurement

Combine performance of the cleaner and the straw walker is commonly evaluated by measured processing grain loss. Loss can be described as percentage of grain lost, rate of loss, or amount of loss per unit of land area (Hailander et al, 1983). The most common performance criterion found in the literature is loss expressed as a percentage of grain available on a component for a specified time interval and is the criterion adopted by the Society of Agricultural Engineers (Agricultural Engineers Yearbook, 1983).

Header loss (loss at the feeding and cutting stage) is considered more difficult to measure than walker or cleaner loss. When measured, header loss is usually expressed in one of the three forms previously

mentioned. Header loss is thought to account for most of the loss during combine harvesting and may vary from 0:5 to 2.0 percent (Ridenour et a1,l968).

The cylinder-concave component can be evaluated in a number of ways. Percentage of grain lost is sometimes used. Perhaps the most common standard of cylinder-concave performance is efficiency. Cylinder-concave efficiency is defined as the percentage of grain separated through the concave. Grain damage is also a measure of performance but varies in importance. Grain damage is very important in crops that are harvested for seed because germination is affected. In grain harvested for consumption, damage is not as important because there is no incentive for high quality grain beyond minimum requirements (Mailander et al, 1983).

2.3.2 Factors That Affect Combine Performance

combine performance is affected by many factors such as machine adjustments, crop conditions, and ground speed. ASAE standard 8396T assumes that increased feedrate causes increased total processing loss as well as increased walker loss and cleaning loss. Grain processing loss has been described as an exponential function of feedrate (Kirk et al, 1977; Kumar and Goss, 1978; Friesen, 1966) while other researchers have used a linear function of feedrate raised to a power (Wrubleski, 1977; Reed et al, 1968; Audsley, 1979). Although the relationship between loss and feedrate are documented, values for the coefficients describing the relationship between feedrate and grain loss vary substantially. The discrepancy between coefficients of various equations based on similar machines suggests that factors other than feedrate affect grain loss. Researchers have examined the effects of machine design, machine adjustments, and to a lesser extent crop

properties.

Cylinder speed and concave clearance have been shown to affect cylinder-concave performance when M.O.G. (mass other than grain) feedrate is constant (Vas and Harrison, 1964; Ridenour, 1968; Cooper, 1971; Rainer, Kepner, and Barger, 1980). Generally, faster cylinder speeds and more narrow concave settings are associated with higher threshing efficiency. Excessive cylinder speeds cause straw break up in some crops such that the cleaning component is loaded to a point that performance decreases. Goss et al (1958) and Vas and Harrison (1964) concluded that cylinder speed has a greater effect upon threshing efficiency and threshing loss than concave clearance. In addition to machine adjustments, orientation of the material entering the cylinder-concave has also been determined to affect separation efficiency (Arnold, 1964) .

The straw walkers are also similarly affected by factors other than feedrate. Machine parameters are known to significantly affect walker performance. Walker crank speed and crank throw were investigated by Reed, Zoerb, and Bigsby (1974) as was straw walker length. In general, walker performance is optimum when the material is aggressively tossed upward and moved rearward. Goss (1958) and Reed, Zoerb, and Bigsby (1970) found grain to straw ratio to have a negative affect upon walker performance (higher grain loss). Straw length was found to be of little importance by Reed, Zoerb, and Bigsby (1970) in contrast to Huisman (1977). Huisman also found relative humidity and bulk density of straw to correlate with walker performance. He also reported that straw coefficient of friction, straw moisture, grain moisture, and modulus of elasticity to be poorly correlated with walker performance. Conversely, Nath (1982) found grain moisture to be an important factor relative to

walker performance.

Nyborg, McColly, and Hinkle (1969) found reducing the grain to chaff ratio decreased cleaning shoe loss in some instances. Laboratory experiments have shown that shaker frequency, air flow characteristics and material entrance conditions affect cleaning performance (MacAulay and Lee, 1969; Rumble and Lee, 1970). Cylinder speed and concave clearance were reported to affect cleaning performance (Nath, Johnson, and Milliken, 1982). Other parameters such as chaffer opening, and cleaner slope also affect performance. In addition to feedrate and machine parameters, crop factors also affect cleaning performance. Nath, Johnson, and Milliken (1982) reported loss increased with increased grain moisture. Huynh (1982) reported increased moisture was responsible for increases in chaff coefficient of friction. Higher coefficients of friction resulted in faster conveying times over the component such that grain was not allowed to pass through the crop mat. experiments have shown that s
and material entrance conditi
and Lee, 1969; Rumble and Lee
clearance were reported to af
and Milliken, 1982). Other pa
cleaner slope also affect per
machine parameters, crop fact
Nath, Johnso

Nyborg, McColly, and Hinkle (1969) developed the following equation to describe cylinder loss in small grains:

n. - 4.76E—4 (m)1°5 c;/s"1'69 [11 where

 $TL = thresholding loss (percent)$

FR - M.O.G. feedrate (pounds per minute)

 G/S = ratio of grain to M.O.G. feedrate.

The correlation coefficient for the equation was reported as 0.50. Fairbanks, Johnson, Schrock, and Nath (1979) described threshing loss in grain sorghum by the following equations:

 $TL = 10.35 - 4.76(CS) + 0.27 1(CC) \dots \dots \dots \dots \dots \dots \dots \dots \dots \tag{2}$

 $TL = 3.46 + 0.217(M) - 0.261(CS) + 0.208(CC) \dots \dots \dots \dots \dots \dots \tag{3}$

 $\mathbf{9}$

where

 $TL =$ threshing loss (percent)

 $CS = cylinder speed (m/s)$

 $CC = concave clearance (mm)$

 $M =$ qrain moisture (percent)

Correlation coefficients were reported as 0.49 and 0.71, respectively.

Nath, Johnson, and Milliken (1982) developed the following equation to predict threshing loss in grain sorghum:

TL - 9.105 + 0.144(M) + 0.150(8) + (0.111)(C)(2613(F2) + 350.0 (Fs)(1o'4)) ⁺ (2573.0 (m2)(cs) ⁺ 16.0(MSZ)(C2)10'4 [41

where

```
M = grain moisture (percent)
      S = cylinder speed (m/s)C = concave clearance (mm)
      F = \text{feedback } (kq/s)The correlation coefficient for the model was 0.50.
     Huisman (1983) proposed what he termed a "simplified model":
     TL - TLF(FGT) ................................................. [5]
where
     TL = threshing loss (kq/s)
     TCF = threshing loss fraction
     FGT = grain feedbacke (kg/s)
```
Threshing loss fraction (TLF) is expressed as follows:

TLF - (l - TSE)(0.025) ... [6]

where

TSE - threshing separation efficiency

Huynh, Powell, and Siddall (1982) developed a stochastic model to describe the threshing and separation process in cereal grains. The

time required for a kernel of grain to be threshed after entering the cylinder concave, the time required for a kernel to pass through the straw'mat, and the time required for a kernel to pass through the concave grate were treated as random variables with characteristic distributions. They were able to determine the probability that a kernel would be threshed and separated before being carried out with the straw mat . time required for a ker
cylinder concave, the t
straw mat, and the time
concave grate were trea
distributions. They we
kernel would be threshe
straw mat.
2.5 <u>Straw Walker Models</u>

2.5 Straw Walker Models

Nyborg, McColly, and Hinkle (1969) proposed the following equation: m. - 0.102 ("0'82 (c/S)'1°73 [71

where

 $WL =$ rack or walker loss (percent)

 $FR =$ feedrate (pounds per minute)

 $G/S =$ grain to straw ratio

The correlation coefficient for the model was 0.74.

Reed, Zoerb, and Bigsby (1970) used a slightly different approach than Nyborg, McColly, and Hinkle (1969) to model walker performance. Reed, Zoerb, and Bigsby (1970) concluded that wheat separation can be described by a decaying exponential function and developed an equation to predict walker length for specified separation efficiency. They proposed evaluating walker performance by the walker length required for a given efficiency. The equation:

L - ln(100 — eff)/b ... [8]

eff - exp(-b * L) ... [9]

where

 $L = length(m)$ of the walker

 $b = an$ empirically derived value dependent upon the grain feedrate, the M.O.G. feedrate, grain to M.O.G. ratio, crop factors, and walker design

Huisman, Heining, van Loo, and Bergman (1974) determined that a model incorporating M.O.G. feedrate, grain feedrate, and relative humidity best described walker loss. Other factors such as grain moisture, straw moisture, and stubble length were considered to be less significant. The equations:

WL = $-11.96 + 1.40$ (FRS) + 1.64(ln(RH) + 0.017(MCS) + ln(MCG) +

3.2E—4(SL) + 0.021(FRG) [10]

where

WL = walker loss (kq/s) FRS = straw feedrate (kq/s) FRG = grain feedrate (kq/s) $RH =$ relative humidity (percent) $MCS =$ straw moisture (percent) $MCG =$ grain moisture (percent)

The model had a correlation coefficient of 0.91 and 0.77.

Nath, Johnson, and Milliken (1982) determined that walker loss in grain sorghum was a function of grain moisture, cylinder speed, cylinder-concave clearance, and feed rate. The correlation coefficient of the model was 0.51. The equation:

WL - 32.78 - 3.57(M) + 0.97(M2) - 0.091(C) - 7.8 (F)(0.00047(SC2)) + 0.87 - (227.04(m>2(r) + 0.4 (mzusm - 0.805(MSC)2)(10-4) [11]

where

 $M =$ grain moisture (percent) $S = cylinder speed (m/s)$

 $C = concave clearance (mm)$

 $F = M.O.G.$ feed rate (kq/s)

C = concave clearance (mm)
F = M.O.G. feed rate (kg/s
2.6 Cleaning Performance Models 2.6 Cleaning Performance Models

Nyborg, McColly, and Hinkle (1969) proposed a model to describe cleaning loss in Canadian wheat. The equation:

CL - 0.116(FR)°°37(c/S)'1'35 [12]

where

 $CL = cleaning loss (pounds)$ $FR = feed rate (pounds/minute)$ $G/S =$ qrain to straw ratio

Fairbanks, Johnson, Schrock, and Nath (1979) developed two models to predict cleaning losses in grain sorghum. The equations:

 $CL = 9.953 - 0.3382(MG) + 0.00069(MG^{2}) + 7.0(10-6)(CC^{3}) \ldots \ldots$ [13] $CL = 7.507 + 0.358(CS) + 0.00547(MGCS) \dots \dots \dots \dots \dots \dots \dots \dots \dots \quad [14]$

where

 $CL = cleaning loss (percent)$ $MG = grain moisture (percent)$ $CC = concave clearance (mm)$ $CS = cylinder speed (m/s)$

Huynh and Powell (1978) developed a probalistic model based upon two events: the migration of kernels through the chaffer openings, and crop dwell time on the cleaning component. The equation:

^R _ ^e -t/T........ .. [15]

where

- $R = the fraction of grain lost$
- $t =$ the reciprocal of the mean time required for the grain to pass through the material mat

 τ = the crop dwell time in the chaffer

2.7 Systems Research
2.7 Systems Research 2.7 Systems Research

2.7 Systems Rese
2.7.1 Definition 2.7.1 Definition

Systems research is an analytical study of a system and its sub-systems. The method is a means to rationally quantify the parameters of a system and the inter-relationships of the parameters. Systems research activities can be categorized as system analysis and system synthesis. System analysis involves the separation of a system into fundamental components, while system synthesis utilizes the information gained from the analysis to observe or modify the existing system (Manetsch and Park, 1982). parameters or a sys
Systems research ac
system synthesis.
into fundamental co
information gained
system (Manetsch an
2.7.2 System Models

2.7.2 System Models

Models are quantitative representations of a process (system). They are used to gain knowledge and convey information. Models are typically used for any or all of the following reasons:

- 1) economic considerations,
- 2) availability,
- 3) information.

There are two broad categories of models: deterministic and probabilistic. A deterministic model produces a repeatable set of outcomes while a probabilistic model introduces an element of uncertainty. The output from a probabilistic model varies if repeatedly provided with the same set of inputs while a deterministic model will yield the same values for a repeated set of inputs. Models are quantitative rep
They are used to gain knowledge
typically used for any or all of
1) economic considerations,
2) availability,
3) information.
There are two broad categories of
probabilistic. A deterministic
out

2.7.3 Testing and Implementation

After constructing a systems model, it is necessary to prove that the model is an adequate representation of the real process depicted. First the model must be verified. Verification is the process of

checking the mathematical correctness of the expressions in the model. Second, the model is validated to compare the output to reality. In some cases, it is not possible to validate the model because:

1) the real world process may not exist, or

2) there may be to little information about the

the model as he operated the combine through the simulated field of corn which was projected on the screen. The simulation allowed engineers to gather data in a laboratory setting where there was more control of the experiment and less cost. Systems research is a technique to examine a complete system. Agricultural engineers and other researchers have successfully employed this methodology to study existing or future systems. Second, the model is validat
some cases, it is not possib
1) the real world proc
2) there may be to lit
the model as he operated the
which was projected on the s
gather data in a laboratory
experiment and less cost. S
comp

2.8 Crop Properties Research

The bulk of crop properties research has been related to material handling of fruits and vegetables and for processing of commercial food products.

> Mohsenin (1965) stated," certain physical characteristics and engineering properties of material (food and agricultural products) should constitute important engineering data. Despite ever increasing applications of machinery, little is known about the physical properties of materials which influence the efficiency of the machine and the quality of the product.

Early research was conducted by Zink (1935) to determine the specific gravity of seeds and by Oxley (1944) who reported bulk densities for various grains. Most early research was conducted to aid the development of seed sorting and cleaning. Research has been

conducted to measure various aspects of many types of seed (Zoerb,l960; Harmond 1965; Kazarian and Hall, 1965; Garrett and Brooker, 1965; Brubaker and Pos, 1965; Chung and Converse, 1965; Zoerb, 1972;and Kusterman, 1984). Most testing of seed involved testing individual seeds and not bulk quantities.

Huisman (1977) investigated the effects of straw moisture content, bulk density of straw, straw modulus of elasticity, kinetic coefficient of straw on straw, and straw length distribution. In doing so, he developed several methods to measure the properties.

Straw bulk density was determined by placing a known volume of straw in a circular tub and loading the material to 120 Pa. The container was then shaken for one minute with a frequency of 33.1 cycles per second and an amplitude of 2.5 centimeters. The volume occupied by the straw was then used as the bulk density.

Modulus of elasticity was determined with a specially constructed instrumented test stand. Bundles of straw were subjected to a 3-point simple bending test from which the modulus of elasticity was derived. Coefficient of friction between straws was also determined by a specially constructed instrumented test stand. A straw stem was attached such that it was pulled across another similarly attached stem. The normal force was known and the frictional force was read directly from a force transducer.

CHAPTER III

COMBINE PERFORMANCE AND BULK CROP PROPERTY MEASUREMENT

COMBINE
3.1 <u>Introduction</u> 3.1 Introduction

Measured combine performance is known to change significantly during repeated performance tests. Much of the variation in combine performance can be attributed to changes in crop conditions such that slight changes in crop conditions can cause significant changes in combine performance. Stephens and Rabe (1977) reported that cloud cover or overnight frost caused changes in the crop which resulted in significant performance changes. Also, laboratory tests conducted with stored crops often are not duplicated in field tests and this led researchers to believe that the properties of the stored crop had changed.

A.research project was initiated to determine the effect of crop properties on combine performance. The results of the study have particular relevance for combine performance testing procedures. Prototype combine performance is evaluated by comparison to the performance of a production model combine. Typically, the performance of a production machine for a test day in the field is established first. Subsequent performance prototype tests are compared with this standard. The decision to establish a new standard of performance is not based on quantitative information but upon intuition and the time available. The information from this study will enable test engineers to determine when crop conditions have shifted such that performance is affected. Also, laboratory test results can be extrapolated for field conditions.

3.2 Combine Performance 3.2 Combine Performance

The performance of two combine components was measured during the study: the cleaner and the straw walker as defined by ASAE Standard 5343.1.

Performance measurement of each component was determined from loss curves which were generated with the bag catch method as described in .ASAE Standard 8396. In this method, the combine was operated at a predetermined ground speed and the material which exited the cleaner and the straw walker was caught in two separate bags and the time required to make the catch noted. The bags were then sieved to remove any grain. The amount of grain in each sample was recorded as percentage loss of the total grain processed during the the time the bags were open. Since the time required to collect the material was noted, the feedrate of the material on each component could be calculated. Material feedrates for the cleaner and the straw walker were recorded as the (MOS) feedrates expressed in metric tons per hour. The combine was operated at different ground speeds and the procedure repeated to include a range of feedrates. The performance for the walker or the cleaner was determined by plotting grain loss percentage versus the total MOG feedrate or chaff MOG feedrate (Figures 2 and 3). The relationship between grain loss and feedrate was found to best fit an exponential equation. Simple regressions of the natural logarithm of grain loss on feedrate were performed for each series of bag catches. One equation was calculated for the cleaner and one equation was calculated for the straw walker. The general form of the loss equation after the simple regression was:

1085' - a' + b * f ... [16] where

TYPICAL WHEAT CLEANER PERFORMANCE CURVE

TYPICAL WHEAT STRAW WALKER PERFORMANCE CURVE

loss'= natural log of grain loss (percent)

 a' = natural log of regression coefficient.

 $b =$ regression coefficient

 $f = \text{chaft or M.O.G. feedback}$ (tons/hour)

The equation was further manipulated by taking the exponential of each term in Equation 16 to yield the following general form.for cleaning loss or walker loss: 21

loss'= natural log of grain loss (percent)

a' = natural log of regression coefficient

b = regression coefficient

f = chaff or M.O.G. feedrate (tons/hour)

quation was further manipulated by taking the exponential of

where

 $loss = train loss (percent)$

 $a,b =$ regression coefficients

 $f = \text{chaft or M.O.G. feedback}$ (tons/hour)

Cleaner performance was expressed as the chaff M.O.G. feedrate at 0.5 percent grain loss and straw walker performance was expressed as the total MOG feedrate at 1.0 percent grain loss as calculated from each respective equation. This method was used to determine the performance of components both in the field and in the laboratory. During laboratory testing, the crop material was placed on a conveyer belt and the feedrate was varied by altering the conveyer speed. where

loss = grain loss (percent)

a,b = regression coefficients

f = chaff or M.O.G. feedrate (

Cleaner performance was expres

0.5 percent grain loss and straw wa

total MOG feedrate at 1.0 percent g

respective equati rcent grain loss and straw walker performance wa
MOG feedrate at 1.0 percent grain loss as calcul
tive equation. This method was used to determin
ponents both in the field and in the laboratory.
tory testing, the crop mate

Performance data was gathered on two types of John Deere conventional harvesters, a 6620 combine, and a 8820 combine. All performance information was expressed in terms of a 6620 combine which is know to have two—thirds the capacity of an 8820 machine. respective equation. Thi
of components both in the
laboratory testing, the c
the feedrate was varied b
Performance data was
conventional harvesters,
performance information w
is know to have two-third
3.3 <u>Measuring Bulk</u>

3.3 Measuring Bulk Crop Properties

3.3.1 Material Collection and Crop Properties Measured

Bulk samples of each crop component were collected from a production model combine harvester during generation of a loss curve. .A sample of chaff MOG was collected from the a bag catch collected during

the determination of each cleaner curve and a sample of walker 1006 was collected during the determination of each walker curve. Grain was collected directly from the storage bin of the combine. Approximately 20 kilograms of grain was collected from the grain auger outlet in the storage bin during the generation of a loss curve and used in the determination of grain properties.

Twelve property measurements were performed upon the collected crop material without sorting or grading. They were grain moisture, grain density, grain angle of repose, chaff moisture, chaff density, chaff coefficient of friction, chaff mean length, chaff compressibility modulus, straw moisture, straw density, straw coefficient of friction, and straw compressibility modulus. In addition to loss curves, grain to MOG ratios and chaff to MOG ratios were measured (Table 1). collected directly from the s
20 kilograms of grain was col
storage bin during the genera
determination of grain proper
Twelve property measurem
material without sorting or g
density, grain angle of repos
coefficient of fr

3.3.2 Crop Component Moisture

Grain moisture was determined with a John Deere portable moisture meter. The moisture cmtents of three to five grain sub-samples were determined and the mean moisture cmtent calculated. Chaff and straw moistures (dry basis) were determined by oven drying samples using guidelines established by ASAE Standard 8358.1. MOG ratios and chaff to
3.3.2 <u>Crop Component Mo</u>
Grain moisture was
meter. The moisture co
determined and the mean
moistures (dry basis) w
guidelines established
3.3.3 <u>Crop Bulk</u> Density

3.3.3 Crop Bulk Density

Grain bulk density (kq/m^3) was determined by weighing a 1-litre sub-sample of grain which was collected from the grain tank. Three to five sub-sample densities were measured and averaged to obtain the final value for entry into the data set.

Chaff bulk density (kg/m³) and straw bulk density (kg/m³) were both determined using an automated test stand (Figure 4). Initially, chaff density was determined using a pexiglass cylinder loaded with 400.0

TABLE 1

BULK CROP PROPERTY COLLECTION INFORMATION LISTED BY LOCATION, YEAR, CROP, TEST ENVIRONMENT AND MACHINE MEASUREMENTS. 23

TABLE 1

BULK CROP PROPERTY COLLECTION INFORMATION LISTED BY LOCATION,

YEAR, CROP, TEST ENVIRONMENT AND MACHINE MEASUREMENTS.

TEST STAND U SED TO MEASURE DENSITY AND COMPRESSIBILITY MODULUS OF CHAR? AND STRAW

FIGURE 4

grams of chaff. The cylinder was calibrated such that density was read directly from the side of the cylinder. A surface pressure of 280.0 (kPa) was used to empress the material for density determination because the pressure was thought to approximate the loading commonly found in a working combine. The stand was used to collect data during the 1980 and 1981 growing season. An automated test stand, constructed prior to the 1982 harvest season, consisted of a metal cylinder and a flat circular plate which was driven into the bore of the cylinder to compress either chaff or straw. The cylinder was mounted on three cantilevered strain gauged beams to sense loading and a potentiometer used to determine the distance of the plunger from the bottom of the cylinder. The crop was compressed as the circular plate was driven downward by a screw type drive attached to an electric motor. Figure 5 contains a typical graph of the output measurements produced by the test stand.

Chaff density and straw density were determined at the volume which the material was subjected to 280.0 (kPa). Three kilograms of chaff and one kilogram of straw were used for each respective test of crop material.

A simple regression of force over the range, 80.0 to 200.0 newtons and height (Figure 5) was performed for each sample. The range was chosen because 80 (N) corresponded to 280.0 (kPa) used in the previous stand. The equation:

F - a + b * h .. [18] where

 $F =$ force exerted by the plunger

 $a,b =$ estimated regression coefficients

 $h =$ height (m) of the plunger from the bottom of the tub was used to predict force for a given height and used in the calculation

TYPICAL GRAPH OF DATA FROM TEST STAND USED TO MEASURE
DENSITY AND COMPRESSIBILITY MODULUS OF CHAFF AND STRAW

 $\ddot{}$

of bulk density and compressibility modulus of chaff and straw. The calculation of density for chaff or straw'was performed using the known mass of material, the volume occupied at 280 (kPa), and the height at 280 (kPa) as calculated from.Equation 18.

3.3.4 compressibility Modulus

Compressibility modulus (kPa) is defined as:

a-mfig—r. [19] where

 Δp = the change in applied pressure to the bulk sample

 ΔV = the corresponding volumetric change and

 V_{α} = the initial sample volume

.As the circular plate descended, the volume of the crop material in the tub decreased while the area of the cylinder remained constant. The following calculating form.of compressibility modulus was used:

 $\beta = \frac{\Delta F}{A(\Delta L/L_o)} \dots \tag{20}$

where

 ΔF = the change in force(N) $A = the area(m) of the cylinder$ ΔL = the corresponding change in height(m)

 L_{Ω} = the initial height(m) of the sample

The change in force (ΔF) was predicted using Equation 18. As previously mentioned, the minimum value corresponds to the loading found in a combine. Two hundred newtons was chosen to standardize the fit and to provide the maximum number of linear points. ΔL = the corr
 L_0 = the init

The change in force

mentioned, the mini

combine. Two hundr

provide the maximum

3.3.5 Crop Friction

3.3.5 Crop Friction

Two types of friction measurements were conducted. The coefficient of friction between stainless steel and chaff or straw'was determined as was the internal friction of grain on grain as related by angle of repose.

Chaff and straw coefficients of friction were measured with an automated stand (Figure 6). A.sled was loaded with crop material and placed on a stainless steel surface which revolved when driven by an electric motor. The stand described by Hall and Husman (1981) was supplied by Deere and Company. An additional l-kilogram weight was added to the sled. The sled was adjusted such that only crop material was in contact with the stainless steel surface. The sled was attached by a length of wire to a strain-gauged cantilevered beam. As the steel surface revolved, the frictional force was sensed by the can cantilevered beam. Since the normal force was known, the coefficient of friction was easily determined. The sampling procedure was conducted such that one revolution of the table was the duration of the test. Thirty-three coefficients of friction were averaged for a single value. Grain angle of repose as defined by Hall and Huisman (1981) was measured by placing a l-litre sample of grain in a hollow cylinder (Figure 7). The cylinder was then slowly raised and the resulting cone height measured. The angle of repose was then determined knowing the volume and the height of the cone formed by the grain.

3.3.6 Particle Distribution

Chaff mean length was the only particle size measurement used. Measuring straw mean length was attempted by hand-counting a large straw sample. The measurement did not prove to be repeatable and required an excessive amount of time consequently, the determination of straw mean length was discontinued.

Chaff mean length is the mean particle size of a bulk chaff sample.

TEST STAND USED TO MEASURE CHAFF AND STRAW COEFFICIENTS OF FRICTION

FIGURE 5

TEST STAND USED TO MEASURE GRAIN ANGLE OF REPOSE

.A.test stand constructed by Deere and Company was used to sort chaff into various sizes (Figure 8). Four sieves (19.03, 12.70, 6.35, 3.175 mm) were cmtained in a metal shaker box. A l-kilogram sample was placed in the top of the box. The entire box was then driven by a crank mechanism for two minutes. The amount of material in each sieve was weighed as was the contents in the bottom pan and each weight expressed as a percentage of the total catch. A cumulative curve of percent catch versus sieve size (Figure 9) shows a typical sample distribution. Least squares linear regression was performed on the cumulative distribution data after taking the natural logarithm of each cumulative sieve contents. The resulting equation, where the independent variable was the sieve size of each tray and the dependent variable was the cumulative percentage of material in each tray, was used to calculate the particle size corresponding to 50.0 percent probability. squares linear regional
contents. The resu
the sieve size of e
cumulative percenta
the particle size c
3.3.7 <u>M.O.G.</u> Ratios

3.3.7 M.O.G. Ratios

Although not true crop properties, grain to M.O.G. ratio and chaff to M.O.G. ratio were measured by Deere and Company. Both ratios were calculated with the average feedrates of chaff, grain, and straw as determined by the bag catches for loss curve determination. Grain to M.O.G. ratio was the average grain feedrate (t/h) divided by the average total M.O.G. feedrate (t/n) while chaff to M.O.G. ratio was the average chaff feedrate (t/h) divided by the average total M.O.G. feedrate (t/h) . Although not tru
to M.O.G. ratio were
calculated with the a
determined by the bag
M.O.G. ratio was the
total M.O.G. feedrate
chaff feedrate (t/h)
3.3.8 Instrumentation

3.3.8 Instrumentation

.A Hewelett Packard 85 computer, Hewelett Packard 3497 data acquisition unit, and two specially constructed John Deere signal conditioners were used to collect data from instrumented test stands (Figure 10). A.K—tron electronic scale with digital read-out was used

FIGURE 8

TEST STAND USED TO SIEVE CHAFF INTO SIZE COMPONENTS

PIGURE 9

 $\hat{\boldsymbol{\beta}}$

33

ELECTRONIC INSTRUMENTATION USED FOR
CROP PROPERTY DATA COLLECTION

PIGURE 10

to weigh crop material. The accuracy of the instrument was rated at plus or minus two grams. \sim

 $\sim 10^{-10}$

 $\ddot{}$

CHAPTER IV

CORRELATION OF CROP PROPERTIES TO COMBINE PERFORMANCE AND THE EFFECT OF MOISTURE ON CROP PROPERTIES

CORRELA
AND
4.1 Introduction 4.1 Introduction

This chapter is presented in two major sections. The first portion of the chapter addresses the effects of crop properties on cleaner and straw walker performance. The second portion of chapter discusses the influence of moisture on each crop property.

Several methods were employed to determine the effect of a crop property on cleaner or straw walker performance. An important distinction must be made in this chapter, the effect of each property on performance was analyzed singly. No attempt was made to control the influence of other properties on performance during the analysis of a single property. The objective was to provide a field engineer with a set of information which will enable him to determine when the performance of a combine component has measurably changed by monitoring a single property. To accomplish the objective it was necessary to determine the changes in each property associated with a measurable shift in performance.

The complete data set is located in Appendix A. Means and standard deviations of all the crop property and machine performance measurements are recorded by test location in Table 2 while measurement errors for each property were estimated and recorded in Table 3. The measurement errors are an estimation of the ability to measure each property. For example, grain angle of repose can be measured with an accuracy of 0.40 degrees. A property measurement was the average of several sub-sample measurements. In the case of grain angle of repose, several

" Mean
"" () Standard Deviations " () Standard Deviations

TABLE 2

,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里,我们就不能在这里

MEASUREMENT ERRORS FOR EACH BULK CROP PROPERTY.

 $-$.

sub-samples of grain were drawn and the grain angle of repose of each sub-sample determined. The mean and standard deviation of the sub—sample measurements was then calculated and recorded as the grain angle of repose of the sample. The error of measurement was estimated by calculating the mean of all the standard deviations for each property sample.

The correlation of crop properties to the performance of the cleaner and the walker was performed (Tables 4 and 5) but the correlations did not explain the property change required before a measurable performance shift occurred. Scatter plots of cleaner and walker performance versus each crop property are located in Appendix B. Single variable equations derived from simple regressions were generated to express performance as a function of properties for both the cleaner and the straw walker. The slope of each equation (Table 6) was evaluated to determine the property change associated with a 1—ton/hour feedrate performance increase. For example, the slope of the equation which described cleaning performance as a function of grain angle of repose was -0.34 (ton/hour)/(degree). The reciprocal of the slope was 2.95 (degree)/(ton/hour). In other words, based on the relationship between chaff feedrate at 0.5 percent grain loss and grain angle of repose, a 2.95 degree property change produced a 1—ton/hour change in chaff feedrate.

.A 10.0 percent performance shift was thought to be the minimum detectable difference in combine performance as opposed to a l-ton/hour change in feedrate which was equivalent to a 42.0 percent change in cleaner performance and a 13.5 percent change in walker performance. In fact, a 20.0 percent shift may be a more realistic figure.

It was desired to determine what property changes were required to

 ~ 1

" Observations

*** Significance Level

TABLE 4 continued TAIL! 4 continued

,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个人的人,一个

 $\frac{1}{2}$

 $\ddot{}$

* Correlation Coefficient
** Observations
*** Significance Level ' Correlation Coefficient

" Observations

*** Significance Level

TABLE 5 continued TAIL! 5 continued

TABLE 6

CLEANER AND STRAW WALKER PERFORMANCE AS FUNCTIONS OF CROP PROPERTIES. TABLE 6

SLOPES OF SINGLE VARIABLE REGRESSION EQUATIONS WHICH DESCRIBE

CLEANER AND STRAW WALKER PERFORMANCE AS FUNCTIONS OF CROP PROPERTIES. SLOPES OF SINGLE VARIABLE REGRESSION EQUATIONS WHICH DESCRIBE

observe a measurable shift in performance. Reference points were required to establish a basis for determining the chaff feedrate and total M.O.G. feedrate equivalent to a 20.0 percent shift in cleaner and walker performance. The overall performance means for cleaner and walker performance were used. The mean performance for the cleaner was 1.77 (tons/hour) chaff feedrate at 0.5 percent grain loss and 7.37 (tons/hour) total M.O.G. feedrate at 1.0 percent grain loss for the walker (Table 2). A 20.0 percent performance shift was equal to 0.35 (tons/hour) for the cleaner and l.47(tons/hour) for the walker. Property changes required to cause a 20.0 percent shift in performance from the overall mean were calculated by multiplying the 20.0 percent, feedrate by the inverse slope from each equation. For example, using data from Table 6, the property change for grain angle of repose was 1.04 degrees, or 2.95 (degrees)/(tons/hour) multiplied by 0.35 (tons/hour).

Probabilities were calculated to express the likelihood of observing a property change which would result in a measurable performance shift. The results of the these calculations for cleaner performance and walker performance are shown in Tables 7 and 8.

A.large sample of properties data was assumed such that the standard normal tables were used to determine the associated probabilities. The required change in each property which caused a measurable change in performance was normalized using each respective overall data set standard deviation (Table 2) and the probability determined from statistical tables. For example, "2" values for the standard normal curve were calculated for grain angle of repose by dividing 1.04 degrees (Table 6) times 2.0, the property change required for a measurable shift in cleaner performance, by 2.7 degrees (Table 2),

TABLE 7

PROPERTY CHANGES ASSOCIATED WITH A MEASURABLE CHANGE IN CLEANER TABLE 7
TABLE 7
PROPERTY CHANGES ASSOCIATED WITH A MEASURABLE CHANGE IN CLEANER
PERFORMANCE AND THE PROBABILITY OF SUCH A CHANGE. PERFORMANCE AND THE PROBABILITY OF SUCH A CHANGE.

TABLE 8

PROPERTY CHANGE ASSOCIATED WITH A MEASURABLE CHANGE IN STRAW WALKER 45

TABLE 8

PROPERTY CHANGE ASSOCIATED WITH A MEASURABLE CHANGE IN STRAW WALKER

PERFORMANCE AND THE PROBABILITY OF SUCH A CHANGE. PERFORMANCE AND THE PROBABILITY OF SUCH A CHANGE.

the standard deviation of grain angle of repose for the entire data set. The resulting "2" value, 0.77, describes a standardized distance from.the mean of grain angle of repose which lies at 0.0. The "2" value corresponds to an area under the curve for values of grain angle of repose which lie outside plus or minus 1.04 degrees from the mean of grain angle of repose. The area under the curve which is approximately 44.0 percent is the percentage of grain angle of repose values in the sample which were plus or minus 1.04 degrees from the mean value. 'This percentage is also the sample probability that a grain angle of repose is plus or minus 1.04 degrees from the mean.

Table 9 lists the performance to property ratios for the cleaner and the straw walker. The ratio of change in performance to change in each a property is an indicator of the relative importance of each crop property. The ratios were calculated by dividing 20.0 percent (measurable machine shift) by the change in property (Tables 6 and 7) expressed as percentage of the property mean. For example, the percentage change in grain angle of repose from the mean value of grain angle of repose is 1.04 (Table 2) divided by 22.4 (Table 2) multiplied by 100 percent which equals 4.6 percent. The ratio (change in performance/Change in property) is 4.3 (Table 9) or 20.0 percent divided by 4.6 percent. The ratio indicates how responsive performance was to corresponding property changes. expressed as percentage or the
percentage change in grain an
angle of repose is 1.04 (Tabl
by 100 percent which equals 4
performance/change in propert
by 4.6 percent. The ratio in
corresponding property change
4.2 Effect o ratios were calculated by dividing 20
chine shift) by the change in property
ercentage of the property mean. For e
nge in grain angle of repose from the
e is 1.04 (Table 2) divided by 22.4 (T
which equals 4.6 percent. The by 100 percent white
performance/chang
by 4.6 percent.
corresponding pro
4.2.1 <u>Grain Angle</u>

4.2 Effect of Crop Properties on Cleaner Performance

4.2.1 Grain Angle of Repose

Grain angle of repose was negatively correlated with cleaning performance. As a correlation of -0.75 (Table 4) indicates, performance tended to improve as grain angle of repose decreased. It appears that grain was less likely to pass through the chaff mat as the angle of

 $rac{1}{2}$ \mathfrak{a} \mathfrak{c}_3

-

 \mathbf{s} $\mathfrak{c}_{\mathsf{h}}$ s c_h

 $\mathbf{\hat{c}}$ \mathbb{C} s_t

 $\mathfrak{c}_{\mathfrak{r}}$

 $\frac{c_{h}}{h}$

TABLE 9

UNIT CHANGE IN CLEANER PERFORMANCE AND WALKER PERFORMANCE FOR A TABLE 9

UNIT CHANGE IN CLEANER PERFORMANCE AND WALKER PERFORMANCE FOR A

UNIT CHANGE IN A CROP PROPERTY. UNIT CHANGE IN A CROP PROPERTY.

repose increased.

Cleaner performance appeared to be most sensitive to changes in grain angle of repose based upon a calculated performance to property ratio of 4.3 (Table 9). A 1.04 degree change was required to observe a measurable change in cleaning performance (Table 7) while the likelihood of observing such a property change was 44.0 percent. The ability to measure a 1.04 degree change was well within the accuracy of measurement (Table 3) as the mean sample variation was 0.44 degrees. grain angle of repo
ratio of 4.3 (Table
measurable change i
of observing such a
measure a 1.04 degr
(Table 3) as the me

4.2.2 Grain Density

Grain density ($k\alpha/m^3$) was positively correlated with cleaning performance based on a correlation coefficient of 0.69 (Table 4). Cleaner performance increased 3.3 times for every corresponding increase in grain density (Table 9). A 43.0 (kq/m^3) change in grain density was required to detect a measurable shift in cleaner performance. The probability of such a property change was 21.0 percent. The required property change was well within the sampling variation of 9.5 (kq/m^3) (Table 3). Grain density (kg/
performance based on a
Cleaner performance inc
in grain density (Table
required to detect a me
probability of such a p
property change was wel
(Table 3).
4.2.3 Chaff Coefficient

4.2.3 Chaff Coefficient of Friction

Chaff friction was inversely related to cleaning performance. A correlation of -0.68 indicates a strong relationship to cleaning performance (Table 4). Higher levels of chaff friction tended to impede movement of the chaff mat across the cleaner and probably impeded the flow of grain through the chaff mat.

Cleaning performance had a performance to property ratio of 2.8 with chaff friction (Table 9) which indicates the cleaner is relatively sensitive to changes in chaff coefficient of friction. A friction change of 0.025 was required to observe a measurable performance shift

while the probability of a measurable property shift was 30.0 percent (Table 7). The property change was readily detected based upon a sampling variation of 0.015 (Table 3). while the probabili
(Table 7). The pro
sampling variation
4.2.4 <u>Chaff Density</u>

4.2.4 Chaff Density

Chaff density (kq/m^3) was positively related to cleaner performance based on a correlation coefficient of 0.36 (Table 4). This was not the expected result. It was believed that higher chaff densities, having less pore space, would be more difficult to clean. The decrease in voids was thought to inhibit grain movement through the chaff mat. The cleaner was sensitive to changes in chaff density as evidenced by a performance to property ratio of 1.40 (Table 9). A 5.62 (kq/m^3) change in chaff density was required to observe a measurable performance change (Table 7). The probability of observing a significant property shift was 11.0 percent (Table 7). Based on accuracy of measurement data, it was possible to detect shifts in chaff density. sampling variation of 0.015 (Table 4.2.4 Chaff <u>Density</u>

chaff density (kg/m³) was posibased on a correlation coefficient

expected result. It was believed t

less pore space, would be more diff

voids was thought to

4.2.5 Chaff Compressibility Modulus

Chaff compressibility modulus (kPa) was positively correlated with cleaner performance. The correlation coefficient was 0.85 (Table 4). The relationship seems intuitively correct because chaff with larger modulus values indicates resistance to volumetric change. Chaff which was resistive to volumetric change would likely have more pore space while being cleaned. A performance to property ratio of 1.0 indicates that a 20.0 percent performance shift required a 20.0 percent property shift (Table 9). An 0.90 (kPa) change in compressibility modulus was required to observe a measurable performance shift (Table 7) while the probability of observing a property shift was 37.0 percent. The accuracy of measurement for chaff compressibility modulus was 0.42 (kPa) which was well within the change required (Table 3) to detect a measurable property change. which was well within the chan
measurable property change.
4.2.6 <u>Grain and Chaff</u> Moisture

4.2.6 Grain and Chaff Moisture

Grain moisture appeared to have a negative effect upon cleaner performance based on a correlation coefficient of -0.31 while chaff moisture had little if any effect on cleaning performance based on a correlation of -0.03 (Table 4). The performance to property ratios for grain moisture and chaff moisture were 0.90 and 0.60, respectively (Table 9). A 3.2 percent shift in grain moisture and 4.0 percent shift in chaff moisture were required to observe a measurable performance shift (Table 7). The probability of observing a significant grain moisture shift was 18.0 percent and 5.0 percent for chaff moisture (Table 7). Both property changes are measurable based upon sampling accuracy. The mean variation for grain moisture was 0.24 percent and 0.40 percent for chaff moisture (Table 3). grain moisture and chaft
(Table 9). A 3.2 perce
in chaff moisture were
shift (Table 7). The p
moisture shift was 18.0
(Table 7). Both proper
accuracy. The mean var
0.40 percent for chaff
4.2.7 Chaff Mean Length

4.2.7 Chaff Mean Length

Chaff mean length (mm)'was inversely related with performance. The correlation between cleaner performance and chaff mean length is -0.37 (Table 4). A.higher mean length was a indication of more loading on the cleaning component and decreased performance. The performance to property ratio for chaff length was 0.80 (Table 9). A change of 2.33 (mm) was required to observe a measurable property shift (Table 7). The probability of observing a significant property shift was 5.0 percent (Table 7) while the mean sample variation for chaff mean length was 0.52 (mm) (Table 3).

4.2.8 <u>M.O.G.</u> Ratios 4.2.8 M.O.G. Ratios

Grain to M.O.G. ratio had a minimal effect upon cleaner performance as indicated by the low correlation of -0.19 (Table 4). Based on the poor correlation, the effect of grain to M.O.G. ratio on cleaner performance appears inconclusive.

Chaff to M.O.G. was more highly correlated to cleaner performance than was grain to M.O.G. ratio based on a correlation of 0.58. Cleaner performance tended to increase as chaff to M.O.G. ratio increased. A change of 0.05 in the chaff to M.O.G. ratio resulted in a 20.0 percent change in cleaner performance. The likelihood of observing a measurable change in cleaner performance as a result of a shift in chaff to M.O.G. ratio was 32.0 percent. Charl to M.O.G. was more
than was grain to M.O.G. rati
performance tended to increas
change of 0.05 in the chaff t
change in cleaner performance
change in cleaner performance
ratio was 32.0 percent.
4.3 Effect of Crop Prop M.O.G. ratio had a minimal effect upon clea
y the low correlation of -0.19 (Table 4).
on, the effect of grain to M.O.G. ratio on
pears inconclusive.
M.O.G. was more highly correlated to cleane
to M.O.G. ratio based on a co change in cleaner
change in cleaner
ratio was 32.0 pe
4.3 <u>Effect of Cro</u>
4.3.1 <u>Grain Angle</u>

4.3 Effect of Crop Properties on Straw Walker Performance

4.3.1 Grain Angle of Repose

Grain angle of repose (degrees) affected walker performance in much the same manner as cleaner performance. The correlation between grain angle of repose and walker performance was -0.44 (Table 5). As grain angle of repose increased, performance tended to decrease. Grain with larger values of angle of repose appeared to be less likely to pass through the straw mat.

The performance to property ratio for grain angle of repose and walker performance was 1.7 (Table 9). A 2.77 degree change in angle of repose was required to observe a measurable shift in walker performance. The probability of observing a property change was 5.0 percent (Table 8). The sample accuracy for angle of repose was 0.44 degrees (Table 3).
4.3.2 <u>Grain</u> Density 4.3.2 Grain Density

Grain density (kq/m^3) was positively correlated with walker performance as evidenced by a correlation coefficient of 0.38 (Table 5). Larger grain densities were associated with better grain movement through the straw mat.

A performance to property ratio of 1.4 indicated the influence of grain density on walker performance relative to the other properties (Table 9). A 113.0 (kq/m^3) change in grain density was required to observe a measurable performance shift (Table 8). The probability of observing such a property shift was less than 1.0 percent (Table 8). The sample accuracy was 9.5 (kq/m^3) (Table 3). Larger grain densities w
through the straw mat.
A performance to pr
grain density on walker
(Table 9). A 113.0 (kg/
observe a measurable per
observing such a propert
The sample accuracy was
4.3.3 Straw Coefficient

4.3.3 Straw Coefficient of Friction

Straw coefficient of friction was inversely related to walker performance. The correlation coefficient between straw friction and walker performance was -0.41 (Table 5). As the coefficient of friction increased, the movement of straw across the walker was reduced as was capacity.

The performance to property ratio for straw friction was 1.4 (Table 9). A.0.04 change in straw friction was required to observe a measurable change in walker performance (Table 7). The probability of observing a property change which corresponded to a measurable machine shift was 11.0 percent. Since the sample accuracy was 0.25, a property shift was measurable (Table 3).

4.3.4. Straw Compressibility Modulus

Straw compressibility modulus (kPa) was positively related to walker performance. The correlation coefficient was 0.43 (Table 5). Straw with higher modulus values tended to resist compaction and

maintained its porosity.

The performance to property ratio for compressibility modulus was 0.80 (Table 9). A 0.51 (kPa) change in compressibility was required to detect a measurable performance shift (Table 8). The probability of observing a property change which corresponded to a measurable performance shift was 2.0 percent. The sampling accuracy was 0.38 (kPa) (Table 3). maintained its porosity.
The performance to property
0.80 (Table 9). A 0.51 (kPa) cha
detect a measurable performance s
observing a property change which
performance shift was 2.0 percent
(Table 3).
4.3.5 Remaining Straw P

4.3.5 Remaining Straw Properties

The remainder of the walker performance related properties appeared to have little affect on performance. Grain moisture, straw moisture, straw density, and grain to M.O.G. all appeared to have a negligible affect on performance. All were poorly correlated and required extreme property changes to alter performance. Ine remain
to have little
straw density,
affect on perfo
property change
4.4 <u>Conclusions</u>

4.4 Conclusions

Cleaning performance was more sensitive to crop changes than walker performance and more properties were directly related to cleaning performance. In addition, property changes associated with measurable performance shifts were less for cleaning performance than for walker performance. For example, a 1.0 degree change (Table 7) in grain angle of repose was required for a measurable shift in cleaning performance while a 2.7 degree change (Table 8) in grain angle of repose was required for a measurable change in walker performance. Overall, the cleaner was approximately three times more sensitive to changes in crop properties than the straw walker based on the ratio of percentage performance change to percentage property change required for a measurable performance shift (Table 6). In general, the ratios were approximately three times greater for the cleaner than the straw walker.

54
4.5 <u>Effect of Moisture</u> on <u>Crop Properties</u> 4.5 Effect of Moisture on Crop Properties 4.5 <u>Effect</u> of Mois
4.5.1 <u>Introduction</u>

4.5.1 Introduction

Moisture affects crop properties in a predictable fashion. Many combine operators base field adjustments on crop moisture. Grain moisture is the most common criterion due to ease of measurement and near instantaneous determination using an electronic moisture tester. Table 10 shows the correlation between each crop property and moisture. Specifically, grain properties are correlated to grain moisture, chaff properties to chaff moisture, and straw properties to straw moisture. Scatter plots of each crop plotted as a function of its component moisture are located in Appendix C. Table 10 shows the
Specifically, grain
properties to chaff
Scatter plots of ea
moisture are locate
4.5.2 <u>Grain Density</u>

4.5.2 Grain Density

Grain density tended to decrease as grain moisture increased. The overall data set correlation was -0.33. The correlation between grain density and grain moisture for wheat using data from the Coal valley, Illinois data set was -0.57 (Table 10). This data set was thought to be most representative of field conditions due to the number of observations, wide moisture range, and maturity level of the crop. density and grain
Illinois data set
most representati
observations, wid
4.5.3 <u>Grain Angle</u>

4.5.3 Grain Angle of Repose

Grain angle of repose decreased with associated increases in grain moisture as evidenced by a correlation 0.42 for the entire data set (Table 10). Once again, the Coal valley, Illinois data set illustrates the relationship for a single crop. The correlation for Coal valley was 0.80. 4.5.3 Grain Angle of
Grain angle of
moisture as evidenc
(Table 10). Once a
the relationship fo
was 0.80.
4.5.4 Chaff Density

4.5.4 Chaff Density

Chaff density did not correlate well with chaff moisture. The correlation coefficient was 0.0, however the relationship between chaff

 1982 1982 1983 1984 1984 1980 1980 1982 North Coal 1982 North Valley (Field) Valley
Wheat 0.001 0.001 1984 30 0.001 0.001 -0e57 -0.002
0.002
15.0 0.79 30 0.83 30 $\mathbf{I}=\mathbf{I}$ \blacksquare $\mathbf{I}=\mathbf{I}=\mathbf{I}$ $\mathbf{I}=\mathbf{I}$ \mathcal{A} 0.83 30 0.096 0.001 0.74 Wheat -0.21 38 0.101 $\begin{array}{c}\n\hline\n1984 \\
\hline\n\text{Coul} \\
\text{Valley}\n\end{array}$ 38 0.21 38 $\mathbf{i}=\mathbf{i}+\mathbf{i}$ $\mathbf{I}=\mathbf{I}-\mathbf{I}$ $\mathbf{1}=\mathbf{1}=\mathbf{1}$ $\mathbf{t}=\mathbf{t}-\mathbf{t}$ $\mathbf{t}=\mathbf{t}=\mathbf{t}$ 0.058 0.255 0.167 0.334 0.334 '0.53 \mathbf{a} 0.65 10 0.022 0.36
0.152
0.152 -0.17 -0.17 -3
- 9
- 9
- 9
- 9
- 9 Dakota Wheat 0.24 10 \bullet North 0.34 10 **L983** Valley $\begin{array}{cc}\n 10 \\
 10 \\
 0 \\
 0.077\n\end{array}$ 0.317 0.020 0.089 0.96 Coal 0.36 0.66 0.16 0.82 0.96 0.94 \bullet $\ddot{}$ $\ddot{}$ \bullet -10
 -10
 -376
 -376 -0.81
 10
 0.002 10 4 -
-
-
-
-0.376 - "
"0.81
" $-$ 0.001 0.001 -0e97 -0.11 Valley Wheat $\mathbf{I}=\mathbf{I}+\mathbf{I}$ $\mathbf{I}=\mathbf{I}=\mathbf{I}$ $\mathbf{1}=\mathbf{1}=\mathbf{1}$ $\mathbf{I}=\mathbf{I}=\mathbf{I}$ 1982
North
Dakota Wheat 0.092 0.117 0.016 0.037 0.036 0.065 0.041 0.474 0.80 0.03 -0.63 0.56 7 0.52 7 $\ddot{}$ 0.71 7 0.71 7 **7** $\ddot{}$ -0.70
7
7 California 0.347 0.005 0.469 0.106 0.130 0.387 0.411 0.33 -0.74 0.03 -0.41 0.19 11 0.286 '0.10 -0.09 Wheat 11 Ξ 11 11 11 11 1982 ה
ני
11 Barley 0.347 0.268 Idaho 0.64 13 0.009 0.68 13 0.005 -0.10 13 0.373 0.52 12 0.044 -0.12 13 ב
כ
-
ר $\mathbf{I}^{\top} \mathbf{I}^{\top} \mathbf{I}$ $\mathbf{I}=\mathbf{I}+\mathbf{I}$ 0.003 0.001 0.065 0.082 0.166 1980
Idaho Wheat -0.80 0.60 10 0.033 -0e3' 10 0.48 10 10 0.92 10 0.51 10 $\mathbf{t}=\mathbf{t}+\mathbf{t}$ $\mathbf{I}=\mathbf{I}+\mathbf{I}$ 0.055 0.001 0.001 0.003 0.001 0.42
0.001 -0.46 '0.33 70 0.48 93 0.001 0.31 55 0.011 0.44 95 88 0.21 59 ች
የ
" All Data \bullet 0.. I. Grain Density ' Grain Angle of Repose Chaff Compressibility
Modulus Straw Compressibility
Modulus Grain Angle of Repose Chaff Conpressibility Straw Compressibility Chaff Coefficient of
Priction Chaff Coefficient of Chaff Mean Length Chaff Mean Length Chaff Density Straw Density Grain Density Chaff Density

CORRELATION OF CROP PROPERTIES TO CROP MOISTURE LISTED BY LOCATION. CORRELATION OF CROP PROPERTIES TO CROP MOISTURE LISTED BY LOCATION.

* Correlation Coefficient ' Correlation Coefficient

tt Observations *' Observations

*** Significance Level *** Significance Level

and the construction of the construction o

Grain properties are correlated with grain moisture
Chaff properties are correlated with chaff moisture
Straw properties are correlated with straw moisture Straw properties are correlated with straw moisture Note: Grain properties are correlated with grain moisture Chaff properties are correlated with chaff moisture Note:

 \mathcal{S}

 \bar{z}

density and chaff moisture is generally positive for location sub—sets (Table 10). Chaff density appeared to increase with corresponding increases in chaff moisture. For example, increasing moisture causes crop material to loose its resiliency and compress more easily.

4.5.5 Chaff compressibility Modulus

Chaff compressibility modulus tended to decrease as moisture increased as indicated by a correlation of -0.46 for the entire data set (Table 10). As chaff became less moist, more pressure was required to change a volume of chaff. increases in chaff mois
crop material to loose
4.5.5 <u>Chaff Compressibi</u>
Chaff compressibil
increased as indicated
(Table 10). As chaff b
change a volume of chaf
4.5.6 <u>Chaff Coefficient</u>

4.5.6 Chaff Coefficient of Friction

Chaff coefficient of friction tended to increase as chaff moisture increased. This effect was evidenced by the overall data set correlation coefficient of 0.49 and the various location sub-set correlations (Table 10). (Table 10). As chaff b
change a volume of chaf
4.5.6 <u>Chaff Coefficient</u>
Chaff coefficient
increased. This effect
correlation coefficient
correlations (Table 10)
4.5.7 <u>Chaff Mean Length</u>

4.5.7 Chaff Mean Length

Chaff mean length tended to increase as chaff moisture increased based on the correlation of the overall data set (Table 10). The correlations of the location sub-sets do not support the theory that chaff length increases as moisture increases. It was theorized that wet crop material was less likely to break than dry material. 4.5.7 <u>Chaff</u> mean Lead Lead
Chaff mean len
based on the correl
correlations of the
chaff length increa
crop material was l
4.5.8 <u>Straw Density</u>

4.5.8 Straw Density

Straw density appeared to increase as straw moisture increased based on an overall data set correlation of 0.44 (Table 10). The correlation between straw density and straw moisture was strongest for the Coal valley, Illinois data set. The sub-set data was most indicative of the true relationship because Coal valley was the only

test location where the crop was harvested at less than optimal conditions. test location where the crop was ha
conditions.
4.5.9 <u>Straw Compressibility Modulus</u>

4.5.9 Straw Compressibility Modulus

Straw compressibility modulus tended to decrease as straw'moisture increased. An overall correlation coefficient of -0.34 supported the relationship. Examination of the relationships for the various sub—sets (Table 10) also tended to support an inverse relationship. Like chaff, as the material became drier, it became more resistive to changes in volume. 4.5.9 Straw Compressibili
Straw compressibili
increased. An overall c
relationship. Examinati
(Table 10) also tended t
as the material became d
volume.
4.5.10 <u>Straw Coefficient</u>

4.5.10 Straw Coefficient of Friction

Straw coefficient of friction tended to increase with increased moisture. The overall correlation coefficient of 0.56 indicated a positive relationship between friction and moisture (Table 10). 4.5.10 <u>Straw Coeff</u>
Straw coeffic
moisture. The ove
positive relations
4.5.11 <u>Conclusions</u>

4.5.11 Conclusions

.Although the relationships between properties were subject to considerable variability, there appeared to be discernible trends in most cases. Data from selected test sites appeared to be more representative of the true relationships than the relationships derived from the entire data set. The test program was conducted such that the machines were tested in a narrow range of conditions at a given location. 'While the moisture range may have been similar at different locations, the properties and the performance of the machines was vastly different. For example, the grain moisture at two sites is 12.0 percent but the crop properties are not the same nor is the performance of the machine. The overall data were useful but it should be noted that the variation in moisture was controlled by the nature of the testing program.

CHAPTER V

PREDICTION MODELS

5.0 <u>Introduction</u> 5.0 Introduction

Stepwise linear regression analysis was performed on the property data sets to develop predictive equations for the cleaner and the straw walker. The Statistical Package for the Social Sciences (SPSS) was used on the the Michigan State university Control Data Cyber 750 mainframe computer. Crop properties were chosen for the analysis such that cleaner performance was expressed as a function of grain and chaff properties. Only grain and straw properties were used to describe straw walker performance. It was assumed that the cleaner was affected by grain and chaff properties while the walker was affected by grain and straw properties and not chaff properties.

During the course of the experiment, several types of data transformations were performed on the data set before stepwise regression analysis was used. The transformations included logarithmic transformations properties, properties raised to powers, and properties expressed as mulitplicative combinations of one another. Models of the following general form explained the most variation in cleaning and walker performance: ' §

 $y_i = b_0 + b_1x_{1i} \dots b_kx_{ik} e_i \dots \tag{21}$ where

 y_i = M.O.G. feedrate (t/h) at a fixed grain loss b_1 = estimated regression coefficients x_i = crop properties e_i = error not accounted for by the model $i = 1, 2, ..., n$ observations

 $j = 1, 2, \ldots$, k independent variables

Equation 21 can be manipulated in to a power equation by expressing each term as an exponential:

 $y_i = b_0 * x_{1i}b_1 ... x_{ni}b_n$ [22] where

 y_i = M.O.G. feedrate (t/h) at a fixed grain loss b_i = estimated regression coefficients x_i = crop properties $i = 1, 2, ..., n$ observations $j = 1, 2, ..., k$ independent variables

Covariate models of cleaner and straw walker performance were analysed after stepwise regression was used to develop predictive equations for the cleaner and the straw walker. This was done to determine if location effects contributed to the explaination of combine performance after the effects due to crop properties was removed. Equation (21) was revised to add the classification variable, location. The general covariate model was:

xij - Li + blxij ... + on):i ^j [23] where

 $y_{i,j}$ = M.O.G. feedrate (t/h) at a fixed grain loss L_i = fixed effects (test sites) b_i = estimated regression coefficients $x_{i,j}$ = crop properties e_{ij} = error not accounted for by the model $i = 1, 2, ...$ n treatments (test sites) $j = 1, 2, ... k$ observations

5.1 Cleaner Prediction Equations

The entire set of chaff and grain properties were used for possible inclusion using the stepwise regression process. The most possible observations was insured by including all the cleaning data. The criteria for inclusion or exclusion from the resulting equation was chosen such that a 0.10 level of significance was maintained. Table 11 contains coefficients, constants, adjusted R square values, and number of observations used to develop the prediction equation. Figure 11, a scatter plot of predicted cleaner capacity versus observed cleaner capacity was constructed to graphically depict the overall correlation of the equation.

Grain angle of repose, chaff coefficient of friction, chaff density, and chaff mean length were selected in the that order to describe 72.0 percent of the variation in cleaning performance. Each variable entered the equation at the 10.0 percent level of significance.

In the previously described analysis, the maximum number of observations was made available for stepwise regression. Chaff compressibility modulus was not collected with the instrumented test stand until the 1982 growing season. Including chaff compressibility modulus in the analysis would not have allowed the maximum number of observations for the analysis because missing value option used by the statistical package would discard any data record with missing observations. The more recently collected properties were included in the analysis at the expense of twenty-two observations.

Table 12 lists the prediction equation coefficients, number of observations used in the stepwise regression, F ratios for each variable as they entered the model, and adjusted R squares as each variable entered the model. Figure 12 shows observed cleaner performance plotted

CLEANER PREDICTION EQUATION CDEFFICIEN'TS, ADJUSTED R-SQUARES, AND PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION. 61

TABLE 11

CLEANER PREDICTION EQUATION TO REFICIENTS, ADJUSTED R-SQUARES,

AND PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION.

PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL. PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL.

m: The model was developed using the entire data set consisting of 41observations.

CLEANER PREDICTION EQUATION CDEFFICIEN'TS, ADJUSTED R-SQUARES, AND PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION. 63

TABLE 12

CLEANER PREDICTION EQUATION TO DEFICIENTS, ADJUSTED R-SQUARES,

AND PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION.

PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL. PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL.

NOTE: The model was developed using data gathered after 1981 consisting of 23 observations.

against the predicted data using the equation of Table 12. Grain density, chaff compressibility modulus, chaff mean length, and chaff coefficient of friction were used to explain 92.0 percent of variation in cleaning performance. As before, the 10.0 percent level of significance was used to include or exclude a variable from the model. Grain properties were selected as the primary variable in both analyses. .Although 92.0 percent of the variation in the data set was explained when compressibility modulus was included, the set contained less than half the number of original observations. Also, the larger data set represented a wider range of crop conditions and combine performance. The additional variability explained by including location effects in a covariate model which included grain angle of repose, chaff coefficient of friction, chaff density, and chaff mean length was 2.0 percent (Table 13). Location effects explained an additional 3.0 percent of the variation in cleaning performance in the model which included grain density, chaff compressibility modulus, chaff mean length, and chaff coefficient of friction (Table 13). density, chaff compressibility modulu
coefficient of friction were used to
in cleaning performance. As before,
significance was used to include or e
Grain properties were selected as the
Although 92.0 percent of the variat

5.2 Straw Walker Prediction Equations

As with the cleaner analysis, stepwise regression was performed on various sets of straw and grain property data using transformed and untransformed data. When transformed, the variables were expressed logarithically, raised to powers, and combined multiplicatively with one another. Straw walker performance was best described by a power relationship as illustrated by Equation 22.

Results of stepwise regressions on straw and grain data are shown in Tables 14 and 15. A.set of properties data was selected such that the most possible observations were available for inclusion in the

EFFECT OF LOCATION ON CLEANER PERFORMANCE. THE LOCATION EFFECT WAS TESTED WITH COVARIATE MODELS USING 66

TABLE 13

EFFECT OF LOCATION ON CLEANER PERFORMANCE.

THE LOCATION EFFECT WAS TESTED WITH COVARIATE MODELS US

PROPERTIES PREVIOUSLY SELECTED BY STEPWISE REGRESSION PROPERTIES PREVIOUSLY SELECTED BY STEPWISE REGRESSION. 66
TABLE 13
OF LOCATION ON CLEANER PERFO
FECT WAS TESTED WITH COVARIA
EVIOUSLY SELECTED BY STEPWIS
Based on the Entire Data Set

STRAW WALKER PREDICTION EQUATION, ADJUSTED R-SQUARES AND PARTIAL F-RATIOS AS DERIVED BY STEPWISE REERESSION. 57

TABLE 14

STRAW WALKER PREDICTION EQUATION, ADJUSTED R-SQUARES

AND PARTIAL F-RATIOS AS DERIVED BY STEEWISE REGRESSION.

PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL. PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL.

n
NOTE: The model was developed using the entire data set consisting of 54 observations .

STRAW WALKER PREDICTION EQUATION, ADJUSTED R-SQUARES, PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION. 68

TARIE 15

STRAW WALKER PREDICTION ROUNTIED R-SQUARES,

PARTIAL F-RATIOS AS DERIVED BY STEPWISE REGRESSION,

PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL. PROPERTIES ARE LISTED AS THEY ENTERED THE MODEL.

NOTE: The model was based on data gathered after 1981 consisting of 33 observations.

equation. This resulted in an equation based on 54 observations. Straw coefficient of friction and grain angle of repose entered in the equation at the 10.0 percent level of significance. The equation explained 30.0 percent of the variation in straw walker performance.

Another regression analysis was performed on a properties data set which included straw compressibility modulus. The addition of straw compressibility modulus reduced the number of observations in data set because the property was not measured prior to 1982. The resulting equation based on 33 observations explained 30.0 percent of the variation in straw'walker performance. Grain angle of repose, straw density, and grain density were selected using the stepwise procedure.

Scatter plots of predicted straw walker performance versus observed straw walker performance are presented in Figures 13 and 14.

Location effects explained an additonal 27.0 percent in straw walker performance in a model which included straw friction and grain angle of repose (Table 16). Location effects were also significant in a model which contained straw compressibility modulus, straw density, and grain density. An additional 28.0 percent variation in straw walker was accounted for by the addition of location effects to the model (Table 16). angre of repose
model which con
grain density.
accounted for b
(Table 16).
5.3 <u>Conclusions</u>

5.3 Conclusions

Cleaner performance can be predicted by grain angle of repose, chaff friction, chaff compressibility modulus, chaff mean length, and grain density. Straw walker performance can best be predicted by straw coefficient of friction, straw compressibility modulus, straw density, and grain angle of repose. Ninety-two percent of variation in cleaning

EFFECT OF LOCATION ON STRAW WALKER PERFORMANCE. THE LOCATION WAS TESTED WITH A COVARIATE MODEL USING 72

TABLE 16

EFFECT OF LOCATION ON STRAW WALKER PERFORMANCE.

THE LOCATION WAS TESTED WITH A COVARIATE MODEL USING

PROPERTIES PREVIOUSLY SELECTED BY STEPWISE REGRESSION PROPERTIES PREVIOUSLY SELECTED BY STEPWISE REGRESSION. 72
TABLE 16
LOCATION ON STRAW WALKER PEI
WAS TESTED WITH A COVARIATE
EVIOUSLY SELECTED BY STEPWIS
Based on the Entire Data Set

performance was explained while only 30.0 percent of variation in walker performance was explained by the properties data. Location effects explained an additional 3.0 percent variation in cleaner performance and an additional 30.0 percent variation in walker performance.

CHAPTER'VI

CROP PROPERTY BASED COMBINE SIMULATION MODEL

CR
6.1 <u>Introduction</u> 6.1 Introduction

A.computer simulation model of a John Deere 6620 combine harvester is presented. The model, which was implemented in the Basic programming language on an IBM compatible micro computer will predict grain loss as a function of ground speed, yield, width of cut, and crop parameters. The program listing is found in Appendix D while examples of the program output and instructions for use are located in.Appendix E. language on an
a function of
The program li
output and ins
6.2 <u>Objectives</u>

6.2 Objectives

The objectives of the model were to:

- l. Predict grain loss on the major components of the combine as a function of ground speed, yield, width of cut, and crop bulk property The Objective
1. Predict
function
property
2. Graphica
combine
6.3 <u>Model</u> Concept
	- 2. Graphically show the relationship between crop changes and combine performance.

6.3 Model Concept

Figure 15 is a flow diagram of the combine simulation model. The flow of material can be traced from component to component. The user inputs to model are: grain moisture, chaff moisture, straw moisture, grain density, grain angle of repose, chaff mean length, chaff coefficient of friction, straw density, straw compressibility modulus, crop yield, grain to M.O.G. ratio, and chaff to M.O.G. ratio. Model outputs are: cleaning loss, walker loss, and total loss.

FIGURE 15

FLOW DIAGRAM OF COMBINE SIMULATION MODEL

COMBINE SIMULATION

 $\mathcal{L}_{\mathbf{a}}$

 \Box

 \vec{C}

6.4 Crop Propterty Simulation

The user can describe the properties used to model the crop or select a set of parameters for a give geographic location. The user may also specify various moisture levels for a give crop and/or select crop property data from actual test locations (California wheat, North Dakota wheat, and North Dakota barley) types for simulation.

Crop properties were assumed to be functions of crop type, and environment (weather and soil fertility). It was also assumed that the bulk properties for a given location were functions of moisture at crop maturity.

Equations were derived to express all crop properties in the model as functions of moisture. Specifically, chaff coefficient of friction was expressed as a function of chaff moisture while straw density was expressed as a function of straw moisture.

It was further assumed that crop component moistures can be expressed as functions of one another. Since grain moisture is the most common measurement performed by farmers and test personnel, it was decided to express chaff and straw moisture as functions of grain moisture.

Relationships were derived by regression to predict properties as functions of moisture. The relationships were assumed to be of forms raised to powers. Coefficients for each equation, R squares, F values, and the data used to develop each equation are listed in Table 17.

Crop property data was analyzed by location, by crop type and as a complete data set. numerous equations to predict a property as a function of moisture were generated by using subsets. The criteria used to select an equation was: best R square and'widest range of data. Simply using the entire data set to develop predictive equations was not

* Equation: Property = a*moistureb * Equation: Property = a*moistureb

 $\mathcal{O}(\mathcal{O}_\mathcal{O})$

used because the effects of moisture were controlled by the machine testing'method.

Single variable equations used to predict crop changes as functions of moisture were assumed to be representative for any small grain type. However, a typical grain density at 12.0 percent moisture in California wheat is 900 Kg/m^3 while a typical grain density for North Dakota barley is 700 kg/ m^3 at the same moisture content. The equation derived from Illinois wheat data is not an adequate predictor of California wheat or North Dakota barley unless the equation is adjusted.

The following example illustrates the method used to adjust an equation. The equation which describes grain density as a function of grain moisture is based upon data gathered at the Coal valley, Illinois, test site during the summer of 1984. If it is desired to predict the change in grain density for a simulation of NOrth Dakota barley, the equation must be adjusted to predict the North Dakota condition. Grain density at 12 percent grain moisture as predicted by the equation is 868.0 (kg/m³). Based on prior personal experience, grain density for barley averaged 645.0 (kg/m³). These values are used to illustrate the adjustment.

The equation which describes grain density as a function of grain moisture, found in Table 17, must be altered as follows:

- 1. Linearize the equation,
- 2. Substitute the value of the property,
- 3. Substitute the value for moisture,
- 4. Solve for coefficient "a",
- 5. Convert equation to power form.
- The equation in its power form as shown in Table 17 is: Gden - 965.0 exp"°°°°8°'G"'°St" [24]

where

Gden = grain density (kg/m³)

 965.0 = estimated regression coefficient

 -0.0088 = estimated regression coefficient

 $Gmost = grain moisture (percent)$

The linearized equation, obtained by taking the natural logarithm of each side of the equation, is:

Ln(Gden) - Ln(965.0) + (-0.0088(Gmost)) [25] A.new coefficient can be solved for after substituting the new'values of grain angle of repose and grain moisture. The resulting equation is: The linearized equation
each side of the equati
Ln(Gden) = Ln(965
A new coefficient can b
grain angle of repose a
Gden = 717.0 exp⁽⁻
Typically, the values u
that correspond to the
particular crop.
6.5 <u>Feeding and Cuttin</u>

Gden - 717.0 exPI-0.0088(GIDSI;)) [25]

Typically, the values used to adjust the equation are selected as those that correspond to the optimum.performance of the machine for a particular crop.

6.5 Feeding and Cutting

The feeding and cutting component of the simulation model was developed using assumptions that feedrate is:

1. a function of width of cut,

2. a function of ground speed,

3. a function of yield.

The equation used to simulate feedrate is:

 $\mathbf{FR} = \mathbf{SP}(\mathbf{YD})\mathbf{WD}(\mathbf{1}/\mathbf{GMOG})(0.00329) \dots \tag{26}$

where

 $FR = M.O.G.$ feedrate (t/h) $SP = ground speed (miles/hour)$ Wd = cutting width (feet) $GMOG =$ grain to M.O.G. ratio 0.00329 = unit factor

The chaff feedrate was determined by the following equation: CF - CMDG(FR) [27] where The chaff feedrate was

CF = CMOG(FR)

where

CF = chaff feedrate (t/

CMOG = chaff to M.O.G r

FR = M.O.G feedrate (t

6.6 <u>Cleaning and Walker</u> Loss

 $CF = chaff feedback (t/h)$ $CMOG = chaff to M.O.G ratio$ $FR = M.0.G$ feedrate (t/h)

6.6 Cleaning and Walker Loss

Inputs to the cleaning component and straw'walker component in the model were crop properties, chaff feedrate, and total M.O.G. feedrate. In order to predict losses on each component, it was necessary to develop relationships which described cleaning and walker loss as a function of feedrates and crop properties.

The performance curves (Figures 2 and 3) which describe percentage of grain lost versus feedrate were developed using logarithmically transformed grain loss in a linear regression on feedrate. The form.of the relationship was:

Loss - a * exp((b)(FR)) [28] where

 $Loss = percentage of grain lost$ $a =$ estimated regression coefficient $b =$ estimated regression coefficient

 $FR = M.0.G$ feedrate (t/h)

Relationships were developed using multiple regression to describe cleaning and straw walker performance as functions of the performance curve regression coefficients. For example, a regression analysis using the "a" coefficients of all cleaning performance curves as the dependent variable and grain and chaff properties as the independent variables was performed. The resulting equation described the intercept as a function of properties. Likewise, a similar regression.was performed on the "b" coefficients of all cleaner curves. Both the intercept and the slope of a cleaner curve can be predicted as functions of crop properties. Loss for any feedrate can then be predicted. Equations of the following form were produced:

$$
a_{ij} = b_0 x_{i1}^{b_1 x_{i2}^{b_2} \dots x_{ik}^{b_k} \dots \quad [29]}
$$

$$
b_{ij} = b_0 x_{i1}^{b_1} x_{i2}^{b_2} \dots x_{ik}^{b_k} \dots \quad [30]
$$

where

 $b_{i,i}$ = estimated regression coefficients X_i = crop properties $i = 1, 2, \ldots, n$ observations

 $j = 1, 2, \ldots$, k independent variables

Straw'walker equations of the same form were also produced. The coefficients and statistics for the cleaner and walker loss equations used in the model are found in Tables 18 and 19.

The equations enabled the prediction of grain loss as a function of crop properties and feedrate. For a unique set of crop properties, a unique performance curve was described based on the predicted cofficients. Cleaner loss was calculated in the model using the chaff feedrate (FR) from.the feeding and cutting component (Eq. 27) the following form:

((b (CPI) ^c cleaning loss - ac exp [31] Straw walker loss was calculated using the total M.O.G. feedrate (FR) from the feeding and cutting component (Eq. 26) in an equation of the following form:

((b (FR)) walker loss - as exp ^s [32]

 \mathbf{I}

EQUATION COEFFICIENTS USED TO PREDICT CLEANER LOSS IN SIMULATION MODEL. PARTIAL F-RATIOS ARE LISTED AS 82

TABLE 18

EQUATION COEFFICIENTS USED TO PREDICT CLEANER LOSS

IN SIMULATION MODEL. PARTIAL F-RATIOS ARE LISTED AS

GENERATED BY STEPWISE REGRESSION. GENERATED BY STEPWISE REGRESSION. 82
TABLE 18
FFICIENTS USED TO PREDICT
MODEL. PARTIAL F-RATIOS
RATED BY STEPWISE REGRESSI
'a' Coefficient Statistics

NOTE: .Model based on 23 observations. Adjusted R—square is 0.64.

NOTE: Model based on 23 observations. Adjusted R-square is 0.32.

*Grain Loss = a*exp(b*chaff feedrate)

EQUATION COEFFICIENTS USED TO PREDICT WALKER LOSS IN SIMULATION MODEL. PARTIAL F-RATIOS ARE LISTED AS 83

TABLE 19

EQUATION COEFFICIENTS USED TO PREDICT WALKER LOSS

IN SIMULATION MODEL, PARTIAL F-RATIOS ARE LISTED AS

GENERATED BY STEPWISE REGRESSION. GENERATED BY STEPWISE REGRESSION.

NOTE: Model based on 34 observations. Adjusted R-square is 0.34.

NOTE: Model based on 34 observations. Adjusted R-square is 0.29.

** ess = a*exp(b*total M.O.G. feedrate)

.
6.7 <u>Total</u> Loss
... 6.7 Total Loss

Total losses can be calculated by adding the loss from the walker and the cleaner for each simulated groundspeed. 6.7 <u>Total Loss</u>
Total losses can be calcu
and the cleaner for each simul
6.8 <u>Random Variable</u> Generation

6.8 Random Variable Generation

A random variable generator was implemented to introduce variation into the model. Specifically; crop yield and the crop properties were treated as random variables in the simulation.

The inverse transformation method with piecewise piecewise approximation was used to code the Gaussian generator as described by (Manetsch and Park, 1985). Inputs to the generator to provide a normal distribution were the value of each property and variation expressed as a percent of the property. A random variable
into the model. Speci
treated as random vari
The inverse trans
approximation was used
(Manetsch and Park, 19
distribution were the
a percent of the prope
6.9 Simulation Results

6.9 Simulation Results

The relationships in the model have been checked for mathematical correctness. The model can be validated by plotting predicted feedrates versus actual feedrates used to develop the initial equations. Figure 16 shows the relationship of simulated chaff feedrates at 0.5 percent cleaner loss versus actual chaff feedrates at 0.5 percent cleaner loss. An R square of 0.62 was calculated. Figure 17 shows the relationship of simulated total M.O.G. feedrate at 1.0 percent walker loss to total M.O.G. feedrate at 1.0 percent walker loss. The simulated values explained 70.0 percent of the variation in walker performance. .A sensitivity analysis using the data in Table 20 was performed by holding all but one of the properties at its mean level while varying one property from a value equal to plus one standard deviation to minus one standard deviation from its mean value. The change in performance was expressed as the percentage change from the predicted performance

SIMULATED STRAW WALKER PERFORMANCE VERSUS ACTUAL STRAW WALKER PERFORMANCE
TABLE 20

87
TABLE 20
CROP PROPERTY MEANS AND STANDARD DEVIATIONS USED IN SENSITIVITY ANALYSIS. CROP PROPERTY MEANS AND STANDARD DEVIATIONS USED IN SENSITIVITY ANALYSIS.

 $\mathcal{L}(\mathcal{A})$ and

 \sim \sim

determined by the mean property values. The percentage change in performance was divided by the percentage change in the property to produce a dimensionless number to evaluate each.property in the loss equations.

Based on the sensitivity analysis using data from Table 21, the cleaner was most affected by changes in grain density and least affected by changes in grain angle of repose (Table 21). The straw'walker was most affected by straw density and least affected by grain density (Table 22). Cleaning performance and walker performance curves were generated by the model using actual data from test runs in North Dakota barley (Table 23). Figures 18 and 19 are scatter plots from an actual field test. The best fit line for the actual data is shown as is the simulated performance curve. The cleaning capacity at 0.5 percent grain was loss was 1.50 (t/h) while the predicted capacity was 1.47 (t/h). Actual walker capacity at 1.0 percent loss was 7.10 (t/h) compared to a predicted feedrate of 6.70 (t/h). Both simulated curves were generated with no crop variation. Chaff and grain property variation of 5.0 percent or less tendeds to produce the most realistic cleaner performance data while straw and grain property variation of 5.0 to 10.0 percent tended performance.

Figure 20 shows the relationship of cleaning and walker performance to grain moisture as predicted by the model. Inputs from Table 23 were used as initial property values to simulate performance for six different moisture contents. The initial or reference grain moisture selected was 12.0 percent. Crop variation for cleaner properties was 5.0 percent and crop variation for straw walker properties was 10.0 percent. As expected, the general trend in both cases was a decrease in performance as grain moisture increased.

SENSITIVITY ANALYSIS OF SIMJLATED CLEANER PERFORMANCE.

89

TABLE 21

SENSITIVITY ANALYSIS OF SIMULATED CLEANER PERFORMANCE.

*Properties varied plus or minus two standard deviations fran mean.

SENSITIVITY ANALYSIS OF SIMULATED STRAW WALKER PERFORMANCE.

*Properties varied plus or minus two standard deviations fran mean.

TABLE 23

CROP PROPERTIES, CROP YIELD, AND MACHINE PARAMETERS USED FOR SIWLATION OF CIEANER AND WAIKER PERFORMANCE. 91

TABLE 23

CROP PROPERTIES, CROP YIELD, AND MACHINE PARAMETERS

USED FOR SIMULATION OF CLEANER AND WALKER PERFORMANCE.

THE DATA IS AN ACTUAL BARLEY DATA SET. THE DATA IS AN ACTUAL BARLEY DATA SET.

PIGURE 18

.
6.9 <u>Conclusions</u> 6.9 Conclusions

The crop property based simulation model can predict trends in cleaning performance and in straw walker performance. Compared to actual data, the model explained 62.0 percent of the variation in cleaner performance and 70.0 percent of the variation in straw walker.

 \bullet

CHAPTER VII

SUMMARY AND CONCLUSIONS

7.0 Summary

Fourteen bulk crop properties of wheat and barley were measured and correlated to cleaner and straw'walker performance of a John Deere model 6620 combine harvester. Data was collected from 1980 through 1984 by Michigan State University and John Deere Harvester Works personnel. Crop properties are known to affect the performance of combines in field tests. Stephens and Rabe (1977) reported the affects of weather upon crop properties and the subsequent change in combine performance. Traditionally, crop moisture information has been used to make decisions such as when to harvest and when to make decisions concerning machine adjustments. Little was quantitatively known about the effect of properties on combine performance. Moisture content of grain, chaff, and straw have been shown to explain some of the variation in performance, but models dependent solely on moisture are site specific and fail to fully describe the physical complexities for a small grain crop.

The primary objective of the study was to relate changes in crop properties to changes in combine performance. This information was of particular interest to John Deere Harvester works engineers for use during field tests of prototype combines. The information also establishes the groundwork for a controller which senses changes in one or several crop properties and makes machine adjustments in an operating combine. Performance changes were influenced by changes in grain, chaff, and straw bulk properties.

The properties which displayed the most effect on performance were grain angle of repose, grain density, chaff coefficient of friction and straw coefficient of friction. In this study, moisture was found to have a minimal effect upon combine performance which was probably due to the narrow range of moisture examined. Most variation in crop properties was due to location differences and variety.

A combine simulation model was constructed based on the crop property data gathered during the study. A computer model was implemented on an IBM compatible micro—computer using the BASIC programming language. The model predicted grain loss on the cleaner and straw'walker as a function of ground speed, crop yield, width of cut, and a set of crop properties. The model provides an interactive means to change crop properties and view the result in a graphical fashion. Impremented on
programming lan
straw walker as
and a set of cr
to change crop
7.1 <u>Conclusions</u> property data gathered during t
implemented on an IBM compatibl
programming language. The mode
straw walker as a function of g
and a set of crop properties.
to change crop properties and v
7.1 <u>Conclusions</u>
7.1.1 <u>Effect o</u> ue to location differences and variety
imulation model was constructed based
thered during the study. A computer m
n IBM compatible micro-computer using
uage. The model predicted grain loss
a function of ground speed, crop

7.1 Conclusions

7.1.1 Effect of Crop Properties on Combine Performance

Cleaner performance appeared to be sensitive to changes in grain angle of repose, grain density, chaff coefficient of friction, chaff compressibility modulus, chaff density, grain:M.O.G. ratio, and chaff mean length. The straw walker appeared to be influenced by grain angle of repose, straw coefficient of friction, grain density, straw compressibility modulus, and straw density. Crop properties appeared to influence the performance of the cleaner more than the straw walker. In general, a given property change caused a greater change in cleaner performance than straw walker performance.

Cleaner performance can be predicted with a multi-variable equation consisting of the following properties: grain angle of repose, chaff coefficient of friction, and grain density. Straw walker performance can be predicted using the following variables in a multi-variable

equation: straw coefficient of friction, grain angle of repose, grain density, and straw density. equation: straw coeffici
density, and straw densi
7.1.2 <u>Combine Simulation</u>

7.1.2 Combine Simulation

Parameters which describe an exponential performance curve were predicted as a function of crop properties. A performance curve for a machine is characterized by estimated regression coefficients. Prediction equations were constructed to predict the intercept and the slope of a performance curve as a function of crop properties. A unique performance curve is predicted for a set of crop properties which characterize a crop condition. Inputs to the model were crop yield, width of cut, ground speed, grain density, chaff mean length, chaff coefficient of friction, grain angle of repose, straw density, straw compressibility modulus, and grain moisture. Prediction equations were constructed to
slope of a performance curve as a functi
performance curve is predicted for a set
characterize a crop condition. Inputs to
width of cut, ground speed, grain densit
coefficient of fr

Simulated cleaner data explained 62.0 percent of the variation in cleaner performance while simulated straw walker data explained 70.0 percent of the variation in straw walker performance. The simulated cleaner performance was most affected by grain density and least affected by grain density. Simulated straw walker performance was most affected by straw density and least affected by grain density.

Crop properties were varied as a percentage of a specified mean value. The most realistic simulated cleaner performance occurred when grain and chaff properties were varied approximately 5.0 percent. Straw walker simulation performance was most realistic when grain and straw properties were varied by 10.0 percent.

7.2 Recommendations For Further Research

This study establishes the groundwork for bulk measurements of crop properties and their effects upon the cleaner and the straw walker. A

similar approach should be taken to examine the effects of crop properties upon the threshing process. Also, the identification and measurement of aerodynamic properties should explain additional variation in cleaner and walker performance. Additional properties should be identified and measured to explain more of the variation in straw walker performance as only 30.0 percent of the variation of the in straw walker performance was attributed to crop properties.

Last, the range of moisture at a test site was limited because prototype testing was conducted when conditions were most favorable. A wider range in moisture conditions at a test site would result in more machine and property variation.

APPENDICES

 $\label{eq:2.1} \mathcal{L}(\mathcal{L}^{\text{max}}_{\mathcal{L}}(\mathcal{L}^{\text{max}}_{\mathcal{L}})) \leq \mathcal{L}(\mathcal{L}^{\text{max}}_{\mathcal{L}}(\mathcal{L}^{\text{max}}_{\mathcal{L}}))$

 $\mathcal{L}(\mathcal{L})$ and $\mathcal{L}(\mathcal{L})$.

 $\mathcal{L}(\mathcal{A})$ and $\mathcal{L}(\mathcal{A})$ and $\mathcal{L}(\mathcal{A})$

 $\mathcal{L}^{\text{max}}_{\text{max}}$ and $\mathcal{L}^{\text{max}}_{\text{max}}$

APPENDIX A

 \sim

 \sim \sim

 $\hat{\mathbf{r}}$

 \mathbb{R}^2

Crop Bulk Properties and Machine Performance Measurements Data Set

TABLE 21

List of variable names used to describe crop data and a TABLE 21
List of variable names used to describe crop data and a
brief description of each variable. brief description of each variable.

equation form:

Cl5mfr or Sllmfr = "a" * $exp^{("b"}$ * feedrate)

- Crop Location-Lichto 1980 Wheat -

 $\ddot{}$

 \bar{z}

- Crop Location-Loato 1980 Barley

- Crop Location-California 1982 Wheat

 $\ddot{}$

 $\overline{}$

 $\ddot{}$

 $\overline{}$

- Crop Location-North Dakota 1982 Wheat

 $\ddot{}$

 \bar{z}

Wheat - Crop Location-Coal Valley 1982

ł

106

 $\ddot{}$

 $\hat{\mathcal{L}}$

 $\overline{}$

7

ł

 $\overline{}$

 ~ 10

 $\ddot{}$

- Crop Location-Coal Valley 1984 Wheat

๛๏๏๏๏๏๏๏๏๏๏๏๏๏๏๏๏๏๘๚๚๚๚๚๚๚๚ \bullet \mathbf{a} **អន្តធ្ងន់មន្ត្រី ខេត្តក្នុងមន្ត្រី ដែល**
អន្តធ្ងន់មន្ត្រី មន្ត្រី ដែនមន្ត្រី ដែល ខេត្តក្នុង ខេត្តក្នុង ខេត្តក្នុង 2.4 S 888888888888888888887889.93
S 888888888888888888887889.939 និងទីប្តីមិនមិនមិនមិនមិនមិនមិនទីទីទីទីទីទីទីទីទីទីទីទីទីទី
វិនិងទីប្តីមិនមិនមិនមិនមិនមិនទីទីទីទីទីទីទីទីទីទីទីទីទីទី **ぬめぁヰなぉぉぉぉぉぉぉぉぉぉぉぉ。。。。。** こはははは 운용용왕악왕왕왕왕왕왕성 혁명급급급등급급급급급

- Crop Location-Coal Valley 1984 Wheat

888 8888
00000 8888 13520 2.788588 11185
11111 8888
0000 88888 88888
00000 00000 8888
0000 នុង
ការដូច
ស្រុក **95589**
9598 88888
88888 88888
00000 8888
 0000 8888
0000 **FBBB 3588**

 $\hat{\boldsymbol{\beta}}$

 $\ddot{}$

 $\ddot{}$

APPENDIX B

 $\sim 10^6$

 ~ 100 km $^{-1}$

 $\ddot{}$

 $\mathcal{L}^{\mathcal{A}}$

Scatter Plots of Combine Performance Versus Crop Properties

 $\mathbf{L2}$

SCATTER PLOT OF CHAFF FEEDRATE (t/h) AT 0.5 PERCENT
GRAIN LOSS VERSUS GRAIN ANGLE OF REPOSE (DEGREES)

FIGURE 22

FIGURE 23

SCATTER PLOT OF CHAFF FEEDRATE (t/h) AT 0.5 PERCENT GRAIN LOSSVERSUS CHAFF DENSITY (kg/m^3)

FIGURE 26

FIGURE 27

SCATTER PLOT OF CHAFF FEEDRATE (t/h) at 0.5 PERCENT GRAIN LOSS VERSUS CHAFF MEAN LENGTH (mn)

-4

APPENDIX C

 \mathcal{L}

 \bullet

 $\ddot{}$

Scatter Plots of Crop Properties Versus Crop Moisture

 \sim

 $\ddot{}$

STRAW DENSITY (kg/m³) VERSUS STRAW MOISTURE
(PERCENT DRY BASIS)

STRAW COMPRESSIBILITY MODULUS (KPa)
VERSUS STRAW MOISTURE (PERCENT DRY BASIS)

APPENDIX D

 \sim \sim

 Δ

 \mathcal{A}

Combine Simlation Model Source Code

 2 \rightarrow COMBINE SIMULATION 3 '* Wilbur Mahoney, Michigan State University Ag. Engineering 4 \rightarrow \bullet 6 $7¹$ **VARIABLES:** $8'$ Crop - array, contains crop property data $9₁$ Coeff - array, column one is " a " term of property versus moisture $10⁷$ equations, column two is "b" term $11'$ X, Y - arrays, interpolation tables for normal distribution $12⁷$ Cfeed, Mfeed - array, chaff feedrate and total M.O.G. feedrate $13'$ Closs, Wloss, Tloss - array, cleaner loss, walker loss, and total loss $14'$ Info – array, contains the property and loss data generated by each run $15[′]$ Variation\$ - character, turns random number generator on and off 16^r RS - character, determines if a new "a" calculated for property equations $17'$ 20 'crops 1=gmost 2=ganro 3 gdeno 4=cmost 5=cfrict 6=cmlntg 7=smost 8=smod 9=sdenwb 22 'coeff for props vs. moisture $column 1 = intercept$ $column 2 = slope$ 23 'cmost=f(qmost)3 smost=f(qmost)5 qanro=f(qmost)1 qdeno=f(qmost)2 24 'cfrict=f(cmost)4 smod=f(smost)7 sdenwb=f(smost)6 25 'coeff contains coeff for cla, clb, sla, slb from row 8-23 26 $'row$ 9 8 - 10 11 12 27 'cla-ganro cmlntg **cfrict** gdeno const 28 'row 13 14 - 15 **16** ganro 29 ' clb gdeno cfrict const 30 'row 19 18 - 17 31 'sla smost smod const 21 32 $'$ row $20₂$ 22 23 33 $'s1b$ qdeno sdenwb smost const 34 'pred contains rvs from original coefficients $35'$ 170 KEY OFF 175 RANDOMIZE 180 OPTION BASE 1 181 SCREEN 2 $185'$ 190 DIM X(100), Y(100), XX(100), YY(100), CROP(4, 15), COEFF(30,2), PRED(20), TLOSS(30,10) 200 DIM A(21), B(21), RATE(20), LOSS(20), CFEED(30, 10), MFEED(30, 10), $CLOSS(30, 10)$, WLOSS(30, 10), INFO(30, 25), MOIST(4, 30) 230 COUNT=1:0\$="OFF":TOG=1:R\$="OFF":GRAPH=1:LOOPS=0:Z\$="off" $231'$ 232' load crop data for California, and both N. Dakota sites $233'$ 240 DATA 9,20.7,885,9,.27,6.44,9,2.21,19,1.39,.14,.43,.04 250 DATA 12.06,21.16,800,9.65,.333,6.99,18.177,2.7,14.71,.836,.086,.379,.0379 260 DATA 11.9,24.2,681,9.9,.36,5.26,14.84,1.89,15.35,1.46,.015,.324,.032 $261'$ 262'load data for interpolation used by random number distribution generator

 $263'$ 270 DATA 0,0.05,0.1,0.15,0.2,0.25,0.3,0.35,0.4,0.45,0.5 280 DATA 0.55,0.6,0.65,0.7,0.75,0.8,0.85,0.9,0.95,1.00 290 DATA $-4.0, -1.645, -1.282, -1.036, -0.842, -0.674, -0.524$ 300 DATA $-0.385, -0.253, -0.126, 0.0, 0.126, 0.253, 0.385$ 310 DATA 0.524, 0.674, 0.842, 1.036, 1.282, 1.645, 4.0 $311'$ 312' load data for property prediction equations $313'$ **320 DATA** 5.784,.454,964.985,-.0883,.764,1.102,.122,.379,1.479,.896,8.129,.219 $6, 3.175, -.147, 7.639, .155$ 330 DATA $3.761, .852, 9.488, 1.837, 1.513, 1.1, .024, .026, -7.3, 1.55, -3.03, 1.26, -4.0$ 3,1.72,1.38E25,3.3E24,4.35,100,1.79e-15,100,-2.24,100,-.674,100,.614 ,100,460268,100 $331'$ 340 FOR I=1 TO 3:'read crop data 350 FOR J=1 TO 13 360 READ CROP $(1,J)$ 370 NEXT J 380 NEXT I $381'$ 390 FOR I=1 TO 21:READ XX(I):NEXT I:'read data for rand generator 400 FOR I=1 TO 21:READ YY(I):NEXT I:'read data for rand generator $403'$ 405 'read data for property prediction equations $406'$ 410 FOR I=1 TO 21: FOR J=1 TO 2: READ COEFF(I, J): NEXT J: NEXT I $411'$ 412' define function keys $413'$ 420 KEY 1, +CHR\$(13) 430 KEY 2, +CHR\$(13) 440 KEY $3, +CHRS(13)$ 450 KEY 5, +CHR\$(13) 460 KEY 4, +CHR\$(13) 470 KEY 6, +CHR\$(13) 480 KEY 7, +CHR\$(13) 490 KEY(1) $ON:KEY(2) ON:KEY(3) ON:KEY(4) ON:KEY(5) ON$ 500 KEY(6) ON: KEY(10) ON: KEY(7) ON: KEY(8) OFF: KEY(9) OFF: KEY(10) ON $502'$ 505 VARIATIONS="OFF" 'turn off random function generator 510 BEEP 518 CLS 520 G\$="off" 530 GOSUB 3840 'display main menu 540 ON KEY (1) GOSUB 5320 'select crop and moisture conditions 550 ON KEY(2) GOSUB 615 'run simulation 560 ON KEY(3) GOSUB 2150 'plot loss curves 570 ON KEY(5) GOSUB 1660 'stochastic parameters 580 ON KEY(4) GOSUB 3180 'scatter plots 590 ON KEY(6) GOSUB 4060 'print stats for each simulation 595 ON KEY (7) GOSUB 9000 'exit program 600 GOTO 520 610 END 615 FOR PASS = 1 TO CROPS 'start curve for selected crop

```
620 2$-"m"
630 GMOST=MOIST(PASS, MOISTURES+1)<br>631 GOSUB 4890 'adjust property equations for another curve
650 FOR MOISTPASS-1 TO MOISTURES 'start another curve at a new moist
    level
660 CROP(NUM(PASS), 1)=MOIST(PASS, MOISTPASS)
680 LIMIT-3.5
690 CLS: 700 m1:R$-"OFF"
720 LOOPS=LOOPS+1
730 CURVE(LOOPS)=LOOPS
740 LOCATE 20,20:PRINr"wait.....execution has begun......it may take a
    while"
741' initialize summation variables for property averages
743 GDSTAVG-O :CMOSTAVG-O : SMOSTAVC-O :GANRAVG-O :
    GDENAVG-O :CFRICTAVG-O :CMINIGAVG-O
744 SMODAVG-0: SDENAVG-0: MFEEDAVG-0:
    CFEEDAVG-0:CMOGAVG-0:GMOGAVG-0
745'
746' step thru each curve by 0.5 mph increments
750 FOR SPEED -.1 TO LIMIT STEP .5 ' limit is the number of points per
on a curve<br>760 SPEED(COUNT.LOOPS)=SPEED
765 60508 4700 ' get predicted property values
787 CMOG=CROP(NUM(PASS), 12)
789 GUS-CROP(MM(PASS) ,10)
794 Rs-"CN"
795 MFEED(COUNT, LOOPS)=SPEED*YIELD(PASS)*19*1/GMOG*.00329 'calc mog
    feedrate
830 CFEED(COUNT, LOOPS)=CMOG*MFEED(COUNT, LOOPS) 'calc chaff feedrate
860 '870' NCOUNT-9
880' FOR N=1 TO 15<br>890' Y5=COEFF(NCOUNT, 1): S5=COEFF(NCOUNT, 2)
890' Y5-COEFF(NCOUNT)<br>890' Y5-COEFF(NCOUNT)<br>900' IF N<9 THEN S5-<br>930' PRED(N)-Y5<br>940' NCOUNT-NCOUNT+1
900' IF N<9 THEN S5=COEFF(NCOUNT.2)
910' IF N>=9 THEN S5=COEFF(NCOUNT, 2)
930' PRED(N)-Y5
940' NCOUNT=NCOUNT+1<br>950' NEXT N
957'958' calculate "a" and "b" terms for cleaner loss and walker loss
959'
960 CLA=CROP(NUM(PASS), 2) ^3.76*CROP(NUM(PASS), 6) ^3.31*
         CROP(NUM(RASS),5)'8.03*3.34E-06
970 CLB-CROP(NUM(RASS),3)'-5.03*CROP(NUM(RASS),5)'—3.41*9.18E+12
980 SLA=CROP(NUM(PASS), 3) ^4.438*1.796E-15
990 SLB-CROP(NUM(RASS),3)'—2.24*CROP(NUM(RASS),8)"-.674*
         9 CROP(NUM(RASS),9)'.614*4602681
91'
995' calculate losses
996'
1000 CLOSS(COUNT, LOOPS)=CLA*EXP(CFEED(COUNT, LOOPS)*CLB)
1001 IF CLOSS(COUNT, LOOPS)>=100 THEN CLOSS(COUNT, LOOPS)=100
1005 WIOSS(COUNT, LOOPS)=SLA*EXP(MFEED(COUNT, LOOPS)*SLB)
1006 IF WIDSS(CGJNT,LOOPS)>-lOO THEN WLOSS(COUNT,LOOPS)-100
1020 TLOSS(COUNT, LOOPS)=WLOSS(COUNT, LOOPS)+CLOSS(COUNT, LOOPS)
1021 IF TLOSS(COUNT, LOOPS)>=100 THEN TLOSS(COUNT, LOOPS)=100
```
1330 INFO(DOOPS,24)-NUM(EASS) 1340 NEXT MOISTPASS,PASS 1350 CLS 1360 RETURN 1361' 1370 'subroutine normal distribution 1371' 1380 FOR $J = 1$ TO 100 1390 Kl-RND(1) 1400 FOR I = 1 TO 21 1410 IF x1 < XX(I) THEN 1430 1420 NEXT I 1430 Y2=(X1-XX(I-1))*(YY(I)-YY(I-1))/(XX(I)-XX(I-1))+YY(I-1) 1440 YB-YS + SS*Y2 IP CIDSS(CQJNI',LOOPS)<01 'I'HEN CIDSS(COINI',IOOPS)-1E—33 1027 IF WLOSS(COUNT, LOOPS)<0! THEN WLOSS(COUNT, LOOPS)=1E-33 1028 IF TLOSS(COUNT,LOOPS)<0! THEN TLOSS(COUNT,LOOPS)=1E-33 1150 GCOUNT=COUNT 1160 IF COUNT >1 THEN GOSUB 3100 GMOSTAvG-CROP(NUM(PASS),1)+GMOSTAVG: .MFEEDAVG-MFEED(COUNT,LOOPS)+MFEEDAVG GANRAVG—CROP(NUM(PASS),2)+GANRAVG: CFEEDAVG-CFEED(COUNT,LOOPS)+CFEEDAVG GDENAVG-CROP(NUM(RASS),3)+GDENAVG CMOSTAVG-CROP(NUM(EASS),4)+CMOSTAVG CFRICTAVG-CROP(NUM(PASS),5)+CFRICTAVG CMLNTGAVG-CROP(NUM(PASS),6)+CMLNTGAVG SMOSTAVG-CROP(NUM(PASS),7)+SMOSTAVG SMODAVG-CROP(NUM(PASS),8)+SMODAVG SDENAVG-CROP(NUM(RASS),9)+SDENAVG 1199 GMOGAVG-GMOG+GMOGAVG:CMOGAVG-CMOG+CMOGAVG:COUNT-COUNT+1 1200 NEXT SPEED 1202 COUNT=COUNT-1 1210 GCOUNT-COUNT 'pass count to plot INFO(LOOPS,10)-GMOGAVG/OOUNT:INFO(LOOPS,11)-CMOGAVG/OOUNT 1221 INFO(LOOPS, 1)=GMOSTAVG/COUNT: INFO(LOOPS, 2)=GANRAVG/COUNT INPO(LOOPS,3)-GDENAVG/COUNT 1223 INFO(LOOPS, 4)=CMOSTAVG/COUNT 1224 INFO(LOOPS, 5)=CFRICTAVG/COUNT INFO(LO0PS,6)-CMLNTGAVG/OOUNT 1226 INFO(LOOPS, 7)=SMOSTAVG/COUNT 1227 INFO(LOOPS, 8)=SMODAVG/COUNT INFO(LOOPS,9)-SDENAVG/COUNT 1229 MFEEDAVG-MFEEDAVG/COUNT CFEEDAVG-CFEEDAVG/COUNT 1270 INFO(LOOPS, 12) = (-.69-BOC)/B1C 1280 INFO(LOOPS, 13) =- BOW/B1W 1290 INFO(LOOPS, 14) = $(.69 - B0T)/B1T$ 1300 INFO(LOOPS, 15)=R2C: INFO(LOOPS, 16)=R2W: INFO(LOOPS, 17)=R2T 1310 INFO(LOOPS, 18)=BOC: INFO(LOOPS, 19)=BOW: INFO(LOOPS, 20)=BOT INPO(LOOPS, 21)-BlC: INPO(LOOPS, 22)-BlW: INFO(LOOPS, 23)-BlT 1450 NEXT J 1460 RETURN 1465' 1470 'subroutine least squares 1475' 1480 B2=0:B3=0:B4=0:B6=0:B7=0:R3=0:B0=0:B1=0

```
1490 FOR I=1 TO COUNT
1500 83-83+RATE( I ) :B4-B4+LOSS( I )
1510 B6-BG+RATE( I ) '2 :BZ-BZ+RATE( I )*LOSS( I)
1520 R3=R3+LOSS(I)<sup>-2</sup>
1530 NEXT I
1540 Sl-B6-OOIM*(B3/OaNr)'2
1550 S2=R3-COUNT* (B4/COUNT) ^2
1560 B7=B3^2
1570 B8=B2-B3*B4/COUNT
1580 W9-86-B7/CCKN1'
1590 Bl-B8/W9
1600 BO=B4/COUNT-B1*(B3/COUNT)<br>1610 R4=B2-B3*B4/COUNT
1620 R5=(B6-B7/COUNT)*(R3-B4^2/COUNT):R5NEW=(.0001/R4)^2:IF R5<=R5NEW
      THEN R2=.000001:GOTO 1640
1625 R2-R4/R5'.5
1630 S3-SZ-Bl'2*Sl
1640 'S4-SQR( (S3/(CGJNT-2) ))
1650 RETURN
1660 'parameter change subroutine
1661 \overline{\text{C}} \overline{\text{C}}1662 IF Z$<>"ON" THEN LOCATE 10,20:PRINT"You have not selected any crop
      information": I.20: FOR I=1 TO 2000: NEXT I:CLS: RETURN
1690 R$="OFF":VARIATION$="ON" 'turn on random generator
1700 GOSUB 7500'go to get crop variation
1740 IF TEMP-13 THEN EXAMPLE 11, 20: FOR I=1 TO<br>1690 R$="OFF": VARIATION$="ON" 'turn on r<br>1700 GOSUB 7500'go to get crop variation<br>1740 IF TEMP-13 THEN C$="1 % WALKER LOSS"
1740 IF TEMP-13 THEN C$-"1 % WALKER LOSS"<br>1830 CLS
1840 RETURN
1845'
1850 'plotting routine
1860 '
1870 XRANGE-300/(XMAX—XMIN)
1880 YRANGE-lZO/(YMAX-YMIN)
1890 PSET (330,130)
1900 DRAW "u120 r300 d120 1300"
1910 FOR I=1 TO GCOUNT
1920 X(1) = 330 + ABS( (XRANGE * (XMIN-RATE(1)) ) )1930 Y(I)-130-ABS((YRAmE*(YMIN-LOSS(I))))
1940 NEXT I
1950 FOR I=GRAPH TO GCOUNT
1960 IF X(I)>638 0010 2030
1970 IF Y(I)<20 0010 2030
1980 IF Y(I)<0 GOTO 2030
1990 IF X(I)<0 0010 2030
2000 PSET (X(I), Y(I))2010 IF G$-"off" THEN 00808 3930 'get symbol to plot
2020 IF G\text{S}="on" THEN GOSUB 4000
2030 NEXT I
2040 Y=10:X=330
2041 XLABEL-40
2045 FOR LABEL-XMIN 'IO XMAX STEP (XMAX-XMIN)/5
2046 LOCATE 18, XLABEL: PRINT USING"###.##";LABEL
2047 XLABEL-XLABEL-fl
2048 NEXT LABEL
2049 TLABEL-Z
2050 FOR LABEL = YMAX TO YMIN STEP - \frac{1}{2050} FOR LABEL = YMAX TO YMIN STEP - \frac{1}{200}2060 LOCATE YLABEL, 34: PRINT USING"###.##"; LABEL
```
2070 YLABEL-YLABEL+3 2080 NEXT LABEL 2090 PSET(330,130) 2100 FOR I=1 TO 5:X=(300/5)+X:PSET(X,130):DRAW"U8":NEXT I 2110 PSET(330,10) 2120 FOR I=1 TO 5 2130 Y=(120/5)+Y:PSET(330, Y):DRAW"R15":NEXT I 2140 RETURN 2150 'subroutine to toggle between curves 2160 GCOUNT=0:GRAPH=1:NEWLOOPS=LOOPS:G\$="off" 2170 CLS 2180 LOCATE 10,20: PRINT"The first five or less curves are plotted 2190 LOCATE 11,20: PRINT"by default. To select specific curves 2200 LOCATE 12,20: PRINT" strike the space bar. To select the 2210 LOCATE 13,20: PRINT"default condition strike any key 2220 RS=INREYS 2230 IF LEN(K\$)=0 GOTO 2220 2240 IF K\$=" " THEN GOSUB 5030 2250 CLS 2260 IF TOG>3 THEN TOG-1 2270 IF TOG-1 THEN L\$="CLEANING LOSS CURVE" 2280 IF TOG-2 THEN LS="WALKER LOSS CURVE" 2290 IF TOG=3 THEN L\$="TOTAL LOSS CURVE" 2300 GCOUNT=COUNT 2310 IF NEWLOOPS>5 THEN NEWLOOPS=5 2320 IF TOG><1 THEN GOTO 2440 2330 FOR OO-1 TO NEWLOOPS 2340 NN=CURVE(OO) 2350 FOR J=1 TO COUNT: RATE(J)=CFEED(J, NN): LOSS(J)=(CLOSS(J, NN)): NEXT J 2360 XMIN=0:XMAX=5:YMIN=0:YMAX=5 2370 GOSUB 1850 2380 NEXT OO 2390 GOSUB 2930 2400 C\$="CHAFF FEEDRATE (T/H)" 2410 GOSUB 5200 'label x-axis 2420 C\$="CLEANER LOSS %" 2430 GOSUB 5230 'label y-axis 2440 IF TOG><2 THEN GOTO 2560 2450 FOR OO-1 TO NEWLOOPS 2460 NN=CURVE(OO) 2470 FOR J=1 TO COUNT:RATE(J)=MFEED(J,NN):LOSS(J)=(WLOSS(J,NN)):NEXT J 2480 XMAX=15: YMAX=5: XMIN=0: YMIN=0 2490 GOSUB 1850 2500 NEXT OO 2510 GOSUB 2930 2520 C\$="MOG FEEDRATE (T/H)" 2530 GOSUB 5200 'label x-axis 2540 C\$="WALKER LOSS %" 2550 GOSUB 5230 'label y-axis 2560 IF TOG <3 THEN GOTO 2680 2570 FOR OO-1 TO NEWLOOPS 2580 NN=CURVE(OO) 2590 FOR J=1 TO COUNT:LOSS(J)=CLOSS(J, NN)+WLOSS(J, NN):NEXT J 2595 FOR J=1 TO COUNT:RATE(J)=SPEED(J,NN):NEXT J 2600 XMAX=5: YMAX=5: XMIN=0: YMIN=0 2610 GOSUB 1850 2620 NEXT OO

```
2630 GOSUB 2930
2640 C$="ground speed (m/h)"<br>2650 GOSUB 5200 'label x-axis<br>2660 C$="TOTAL LOSS %" 'label y-axis<br>2670 GOSUB 5230 'label y-axis
2680 LOCATE 1,50:PRINT L$
2690 TOG-10044
2700 LOCATE 22,1:PRINT"strike the space bar to exit plot"
2710 LOCATE 23,1:PRINT"strike any other key to continue"
2720 K$=INKEY$<br>2730 IF LEN(K$)=0 GOTO 2720
2740 IF K$<>" " 0010 2250
2750 CLS
2760 RETURN
2765 '
2770 'graph all curves on same screen
2771'
2780 IF 0S-"CN" THEN 0010 2850
2790 PSET(20,10):GOSUB 2890
2800 PSET(230,10):GOSUB 2890
2810 PSET(440,10):GOSUB 2890
2820 LOCATE 1,10:PRINT"CLEANER LOSS CURVE"
2830 LOCATE 1,35:PRINT"WALKER LOSS CURVE"
2840 LOCATE 1,60: PRINT" TOTAL LOSS CURVE"
2850 PSET(X, Y)
2860 DRAW"EZ 04 E2 P2 H4"
2870 GS-"ON"
2880 RETURN
2890 DRAW"DlOO R190 0100 L190"
2900 DRAWMDZO R190 020 L190 020 R190 020 L190 080"
2910 DRAW"R38 0100 R38 0100 R38 0100 R38 0100"
2920 RETURN
2930 ' print info about each curve at plot edge
2940 PSET(180,3)
2950 YPOS-O
2960 LOCATE 1,1
2970 FOR INN-1 '10 LOOPS
2980 00508 3930 'get symbol and plot
2990 00508 4980 'GET CROP LABEL
3000 PRINT HS
3010 PRINT "GRAIN MOISTURE = "; INFO(NN, 1)
3020 IF TOG-1 THEN PRINT"CAPACITY = ";INFO(NN,12);"(T/H)"<br>3030 IF TOG-2 THEN PRINT "CAPACITY = ";INFO(NN,13);"(T/H)"
3040 IF TOG-3 THEN PRINT"SPEED = ";INFO(LOOPS, 14); "(M/H)"
3050 YPOS-YPOS+32
3060 PSET(180,YPOS+3)
3070 PRINT
3080 NEXT NN
3090 RETURN
3095'
3100 ' subroutine calculates statistics for each loss curve<br>3101 ' "a", "b" and r-square
3102'
3110 FOR J=1 TO COUNT:
     RATE(J)-CFEED(J,LOOPS):LOSS(J)-LOG(CLOSS(J,LOOPS)):
     NEXT J
3120 00808 1470 :BOC-BO :BlC-Bl :R2C-R2
```

```
3130 FOR J=1 TO COUNT:
     RATE(J)-MFEED(J,LOOPS):LOSS(J)-LOG(WIOSS(J,LOOPS)):
     NEXT J
3140 00808 1470:80w-BO:BlW-81:RZW-R2
3150 FOR J=1 TO COUNT:
     LOS(J)=LOG(WLOS(J,LOOPS)+CLOS(J,LOOPS)):
     NEXT J
3155 FOR J=1 TO COUNT:
     RATE(J)=SPEED(J,LOOPS):NEXT J
3160 GOSUB 1470:B0T=B0:B1T=B1:R2T=R2
3170 RETURN<br>3175'<br>3180 'subroutine to plot properties and loss information
3185 '
3190 CLS:G$-"on"
3200 LOCATE 1,1: PRINT"1 = grain moisture"
3210 PRINT"2 = grain angle of repose"
3220 PRINT"3 = grain density"
3230 PRINT"4 = chaff moisture"
3240 PRINT"5 = chaff friction"
3250 PRINT"6 = chaff mean length"
3260 PRINT"7 = straw moisture
3270 PRINT"8 = straw modulus"
3280 PRINT"9 = straw density"
3290 PRINT"10 = qrain to mog ratio"
3300 PRINT"11 = chaff to mog ratio"
3310 PRINT"12 = chaff feedrate at 1/2 % loss"
3320 PRINT"13 = mog feedrate at 1 % loss"
3330 PRINT"14 = mog feedrate at 2 \text{\$} total loss"
3340 LOCATE 1,40: PRINT" select the x-axis variable"
3350 LOCATE 2,40:1NPUT XVAR
3360 LOCATE 4,40:PRINT"select the y-axis variable"
3370 LOCATE 6,40:INPUT YVAR
3380 XMAx-9999I:YMAx-—9999I
3390 MN-99999!:YMIN-99999I
3400 FOR I-l TO LOOPS
3410 IF INFO(I, XVAR) > XMAX THEN XMAX=INFO(I, XVAR)3420 IF INFO(I,XVAR)<XMIN THEN XMIN-INFO(I,XVAR)
3430 IF INFO(I,YVAR)>YMAX THEN YMAX-INFO(I,YVAR)
3440 IF INEO(I,YVAR)<YMIN THEN YMIN-INFO(I,YVAR)
3450 NEXT I
3460 XMIN-XMIN*.9:XMAX-XMAX*1.1
3470 YMIN-YMIN*.9:YMAX-YMAX*1.1
3480 GCOUNT-0
3490 CHECK(l)-INFO(1,24)
3500 N=13510 00808 4980 ' go get crop label
3520 YPOS-O
3530 LOCATE 1,1
3223 FOR 0-1 TO LOOPS:RATE(O)-INEO(O,XVAR):LOSS(O)-INFO(O,YVAR)
3560 GCOUNT-GCOUNT+1
3570 GRAPH=GCOUNT 'setup to plot one point and return
3580 IF 0-1 THEN CLS
3590 GOSUB 1850
3610 FLAG-0
3620 NN-O:M-O
```

```
3630 CHECK(O)-INFO(O,24)
3631 IF O-1 THEN LOCATE 1,1: PRINT H$:PSET(180,3):GOSUB 4000:GOTO 3710
3640 FOR JJ-l '10 M—l
3650 IF CHECK(JJ)-INFO(O,24) THEN FLAG-1
3660 NEXT JJ
3670 IF FLAG- 1 6010 3710
3671 YPOS-YPOS+25
3672 PSET(180,YPOS+3)
3673 LOCATE O+1,1
3681 PRINT
3690 GOSUB 4980 'get crop label because it is not a duplicate
3700 PRINT H$ 'print label
3703 03808 4000 'pick up symbol
3710 NEXT 0
3720 FOR 1-1 TO 2
3730 IF I=1 THEN TEMP=XVAR
3740 IF I=2 THEN TEMP=YVAR
3750 GOSUB S620 'pick up property labels
3760 IF I-l THEN 00808 5200 'label x-axis
3770 IF 1-2 THEN GOSUB 5230 'label y—axis
3780 NEXT I
3790 LOCATE 23,1:PRINT "strike any key to continue"
3800 Ks-INKEYS
3810 IF LEN(K$)-0 0010 3800
3820 CLS
3830 RETURN
3835'
3840 'subroutine to display main menu
3845'
3850 LOCATE 10,20:PRINT"F1. select crop location"
3860 LOCATE 11,20:PRINT"F2. run sinulation"
3870 LOCATE 12,20:PRINT"F3. plot loss curves"
3880 LOCATE 13,20:PRINT"F4. plot scatter plots of properties"
3890 LOCATE 14,20:PRINT"F5. set parameters for stochastic process"
3900 LOCATE 15,20:PRINT"F6. display historical data from simulation run" 3901 LOCATE 16,20:PRINT"F7. exit program
3920 RETURN
3925'3930 'subroutine to determine which symbol to plot
3935'
3940 IF NN—l THEN DRAW"U3 d3 h3 f3 L3 R3 93 e3 d3 u3 f3 h3 r3 13 e3"
3950 IF INN-2 THEN DRAW"u3 13 d6 r6 u6 13"
3960 IF NN=3 THEN DRAW"e3 q6 e3 h3 f6"
3970 IF INN-4 THEN DRAW"e3 16 f6 16 e3"
3980 IF INN-5 TEEN DRAW"U3 D6 L3 U6 R6 06 L3 R3 U3 L6"
3990 RETURN
3995'
4000 IF INFO(0,24)=1 THEN DRAW"U3 d3 h3 f3 L3 R3 q3 e3 d3 u3 f3 h3 r3 13
     e3"
4010 IF INFO(0,24)=2 THEN DRAW"u3 13 d6 r6 u6 13"
4020 IF INFO(0,24)=3 THEN DRAW"e3 q6 e3 h3 f6"
4030 IF INFO(0,24)=4 THEN DRAW"e3 16 f6 16 e3"
4040 IF INFO(O,24)-5 THEN DRAW"03 D6 L3 06 R6 06 L3 R3 U3 L6"
4050 RETURN
4055'
2326'subroutine to display historical information for each curve
4057'
```

```
4060 CYCLE-1:RAT-1:CLS:SCREEN 0
4070 WHILE RAT
4080 NN=CYCLE: GOSUB 4980
4090 LOCATE 1,20: PRINT H$; " AT"; INFO(CYCLE, 1); "PERCENT GRAIN MOISTURE"
4100 LOCATE 4, 1: PRINT"
                                      CLEANING LOSS
                                                               WALKER LOSS
     TOTAL LOSS"
4110 LOCATE 7, 1: PRINT"R SQUARE
4120 LOCATE 7,21: PRINT USING "#.##"; INFO(CYCLE, 15)
4130 LOCATE 7,41: PRINT USING "#.##"; INFO(CYCLE, 16)
4140 LOCATE 7,66: PRINT USING"#.##"; INFO(CYCLE, 17)
4150 LOCATE 9,1: PRINT "INTERCEPT 'a'"
4160 LOCATE 9,20: PRINT USING "##. ##^^^^"; INFO(CYCLE, 18)
4170 LOCATE 9,40: PRINT USING"##.##^^^^"; INFO(CYCLE, 19)
4180 LOCATE 9,65: PRINT USING"##.##^^^^"; INFO(CYCLE, 20)
4190 LOCATE 11, 1: PRINT"SLOPE 'b'"
4200 LOCATE 11, 20: PRINT USING"##.##^^^^"; INFO(CYCLE, 21)
4210 LOCATE 11,40: PRINT USING"##.##^^^^";INFO(CYCLE,22)<br>4220 LOCATE 11,65: PRINT USING"##.##^^^^";INFO(CYCLE,23)
4230 LOCATE 13, 1: PRINT "FEEDRATE (t/h)"
4240 LOCATE 14,1: PRINT" (see below)"
4250 LOCATE 13, 20: PRINT USING"##.##"; INFO(CYCLE, 12)
4260 LOCATE 13, 41: PRINT USING "##.##"; INFO(CYCLE, 13)
4270 LOCATE 13,65: PRINT USING"##.##"; INFO(CYCLE, 14)
4280 LOCATE 16, 1: PRINT"GRAIN ANGLE OF REPOSE ="
4290 PRINT"GRAIN DENSITY ="
4300 PRINT "CHAFF MOISTURE ="
4310 PRINT"CHAFF FRICTION ="
4320 PRINT "CHAFF LENGTH ="
4330 LOCATE 16,40: PRINT "STRAW MOISTURE ="
4340 LOCATE 17,40: PRINT"STRAW MODULUS ="
4350 LOCATE 18, 40: PRINT "STRAW DENSITY ="
4360 LOCATE 19,40: PRINT "AVG. GRAIN: MOG ="
4370 LOCATE 20,40: PRINT "AVG. CHAFF: MOG ="
4380 LOCATE 16,28: PRINT USING "###.###"; INFO(CYCLE, 2)
4390 LOCATE 17,28: PRINT USING "###.###"; INFO(CYCLE.3)
4400 LOCATE 18,28: PRINT USING "###.###"; INFO(CYCLE, 4)
4410 LOCATE 19,28: PRINT USING "###.###"; INFO(CYCLE, 5)
4420 LOCATE 20,28: PRINT USING "###.###"; INFO(CYCLE, 6)
4430 LOCATE 16,65: PRINT USING "###.###"; INFO(CYCLE, 7)
4440 LOCATE 17,65: PRINT USING "###.###"; INFO(CYCLE, 8)
4450 LOCATE 18,65: PRINT USING "###.###"; INFO(CYCLE, 9)
4460 LOCATE 19,65: PRINT USING "###.###"; INFO(CYCLE, 10)
4470 LOCATE 20,65: PRINT USING "###.###"; INFO(CYCLE, 11)
4480 LOCATE 22,1:PRINT"cleaner - chaff feedrate at 1/2 % cleaner loss
     walker - mog feedrate"
4490 LOCATE 23,1:PRINT"at 1 % walker loss total - mog feedrate at 2 %
     total loss"
4500 CYCLE=CYCLE+1
4510 IF CYCLE>LOOPS THEN RAT=0
4520 IF CYCLE>LOOPS THEN CYCLE=1
4530 K$=INKEY$
4540 IF LEN(K$)=0 THEN 4530
4550 IF K$="P" THEN GOSUB 4600
4560 '
4570 WEND
4580 SCREEN 2
4590 RETURN
```

```
4591'
4600
'subroutine to dump screen contents to printer
4610'
4610<br>4620 WIDTH "LPT1:",80<br>4630 FOR ROW = 1 TO 24
4640 FOR COL = 1 TO 80
4650
CHAR-SCREEN(RGV,COL)
4660 IF CHAR=0 THEN CHAR=32<br>4670 LPRINT CHR$(CHAR);
4670 LPRINT CHRS(1)<br>4680 NEXT COL, ROW
4690 RETURN
4700'
4701 ' subroutine to calculate properties as functions of moisture<br>4702 '
4701
IF RS-"OFF" THEN YS-CROP(NUM(PASS),1):SS-CROP(NUM(RASS),1)*GNOSTVAR
4702
IF RS-"OFF' THEN'MDISTEMP-YS:DEVTEMP-SS
4703 4705
4710 CROP(NUM(PASS).4)=CROP(NUM(PASS).1)^COEFF(3.2)*COEFF(3.1)
4712 IF VARIATIONS="ON" THEN GOSUB 1370: CROP(NUM(PASS). 4)=Y3
4711 15-CROP(NOR)(PASS), 4): 55-CROP(NOR(PASS), 4)-CROSIVAR<br>4712 IF VARIATION$="ON" THEN GOSUB 1370: CROP(NUM(PASS), 4)=Y3<br>4720 CROP(NUM(PASS), 7)=CROP(NUM(PASS), 1)^COEFF(5, 2)*COEFF(5, 1)
4721
Y5-CROP(NUM(RASS),7):SB—CROP(NUM(RASS),7)*SMOSTVAR
4722 IF VARIATIONS="ON" THEN GOSUB 1370: CROP(NUM(PASS), 7)=Y3
4730
IFVARIATIONS-"ONM THEN GOSUB 1370:CROP(NUM(RASS),7)-Y3 CROP(NUM(PASS),2)—CROP(NUM(PASS),1)'00EFF(1,2)*OOEFF(1,1)
4731
Y5-CROP(MM(PASS),2):SS-CROP(NUM(PASS).2)*GANGVAR
4732
IF VARIATICN$-"CN" THEN GOSUB 1370:CROP(NUM(PASS),2)-Y3 CROP(NUM(PASS),3)-CROP(NUM(PASS),1)'OOEFF(2,2)*COEFF(2,1)
4740 CROP(NUM(PASS), 3)=CROP(NUM(PASS), 1) \text{CDEF}(2,2) * COEFF(2, 1)
4742 IF VARIATIONS="ON" THEN GOSUB 1370:CROP(NUM(PASS), 3)=Y3<br>4750 CROP(NUM(PASS), 5)=CROP(NUM(PASS), 4)^COEFF(4,2)*COEFF(4,1)
4750 CROP(NUM(PASS),5)=CROP(NUM(PASS),4)^COEFF(4,2)*COEF<br>4751 Y5=CROP(NUM(PASS),5):S5=CROP(NUM(PASS),5)*CFRICTVAR
4752 IF VARIATIONS="ON" THEN GOSUB 1370: CROP(NUM(PASS). 5)=Y3
4760
IF VARIATIONS-"ON" THEN GOSUB 1370:CROP(NUM(RASS),5)-Y3 CROP(NUM(PASS).8)-CROP(NUM(RASS),7)"COEFF(7,2)*COEFF(7,1)
4761
Y5-CROP(NUM(RASS),8):SS-CROP(NUM(PASS),8)*SMODVAR
4762 IF VARIATIONS="ON" THEN GOSUB 1370: CROP(NUM(PASS), 8)=Y3
4770
IF VARIATICN$-"(N" THEN GOSUB 1370:CROP(NUM(PASS),8)-Y3 CROP(NUM(PASS),9)-CROP(NUM(RASS),7)'COEFF(6,2)*OOEFF(6,1)
4771
Y5-CROP(NUM(PASS),9):SS-CROP(NUM(PASS),9)*SDENVAR
4772
IF'VARIATIONS-"ON" THEN GOSUB 1370:CROP(NUM(EASS),9)-Y3 CROP(NUM(PASS),6)-CROP(NUM(PASS),4)'00EFF(8,2)*COEFF(8,1)
4776 CROP(NUM(PASS), 6)=CROP(NUM(PASS), 4) \text{CDEF}(8,2) \times \text{COFF}(8,1)4777
Y5-CROP(NUM(RASS),6):SS-CROP(NUM(RASS),6)*CMLNTGMAR
4778
IF VARIATIONS-"0N" THEN GOSUB 1370:CROP(NUM(RASS),6)-Y3
4780 RETURN<br>4800'<br>4801' adjus<br>4802'
4890
COEFF(1,1)-CROP(NUM(RASS),2)/GMOST"COEFF(1,2)
4900
COEFF(2,1)-CROP(NUM(RASS),3)/GNOST'COEFF(2,2)
4910
OOEFF(3,1)-CROP(NUM(RASS),4)/GMOST"COEFF(3,2)
4920
COEFF(4,1)-CROP(NUM(RASS),5)/CROP(NUM(PASS),4)"OOEFF(4,2)
4930
COEFF(S,1)-CROP(NUM(RASS),7)/GMOST"OOEFF(5,2)
4940
COEFF(6,1)-CROP(NUM(RASS),9)/CROP(NUM(PASS),7)'OOEFF(6,2)
4950 COEFF(7,1)=CROP(NUM(PASS),8)/CROP(NUM(PASS),7)^COEFF(7,2)<br>4951 COEFF(8,1)=CROP(NUM(PASS),6)/CROP(NUM(PASS),4)^COEFF(8,2)
4970 4975'
4976'
subroutine to assign labels used to plot loss curvesLPRINT CHR$(CHAR);
       FOR CO<br>CHAR=S<br>IF CHA<br>LPRINT<br>NEXT C<br>RETURN
4702 '
IF RS-"ON" THEN Y5-MOISTEMP:SS-DEVTEMP
       IF VARIATION$="ON" THEN GOSUB 1370:CROP(NUM(PASS),1)=Y3<br>CROP(NUM(PASS),4)=CROP(NUM(PASS),1)^COEFF(3,2)*COEFF(3,1)
Y5-CROP(NUM(RASS),4):SS-CROP(NUM(RASS),4)*CMOSTVAR
Y5-CROP(NUM(PASS),3):SS-CROP(NUM(RASS),3)*GDENVAR
4801' adjust each property curve for moisture reference
4951 COEFF(8,1)=CROP(NUM(PASS), 6)/CROP(NUM(PASS), 4)<sup>2</sup>COEFF(8,2)
4970 RETURN
```
4977' 4980 IF NUM(NN)=1 THEN HS="CALIFORNIA WHEAT" 4990 IF NUM(NN)=2 THEN H\$="NORTH DAKOTA WHEAT" 5000 IF NUM(NN)=3 THEN HS="NORTH DAKOTA BARLEY" 5010 IF NUM(NN)=4 THEN H\$="Special Blend" 5020 RETURN $5021'$ 5022' subroutine to select loss curves to plot $5023'$ 5030 CLS: LOCATE 5.20: PRINT LOOPS:" are available to choose from" 5040 LOCATE 10.20: PRINT"You may plot (1-5) curves on the same plot" 5050 LOCATE 11.21: INPUT "Enter the number of curves you wish to plot";NEWLOOPS 5060 IF NEWLOOPS \leq 5 GOTO 5110 5070 REEP: CLS 5080 LOCATE 10,20: PRINT" you choose more than 5 curves 5090 LOCATE 11,20: PRINT"try again (1-5) 5100 GOTO 5030 5110 CLS 5120 FOR NN-1 TO LOOPS 5130 GOSUB 4980 ' go get crop identification label 5140 PRINT "Curve $(\overline{N};N_N;")$ " $\overline{N};H\overline{S},$ " at"; INFO(NN, 1) 5150 NEXT NN 5160 FOR I=1 TO NEWLOOPS 5170 LOCATE 10.40: INPUT"Enter curve number". CURVE(I) **5180 NEXT I** 5190 RETURN $5191'$ 5200 'label x-axis subroutine 5205 ' 5210 LOCATE 20,42+(37-LEN(C\$))/2: PRINT C\$ 5220 RETURN $5225'$ 5230 'label y-axis subroutine 5235 5240 A\$="":L=LEN(C\$):L1=(18-L)/2 5250 FOR X=1 TO L1:A\$=A\$+"":NEXT X 5260 AS=AS+CS 5270 FOR X=LEN(A\$) TO 18:A\$=A\$+"":NEXT X 5280 FOR X=3 TO 20 5290 LOCATE X, 33: PRINT MID\$ (A\$, X-2, 1) 5300 NEXT X 5310 RETURN 5315 ' 5320 'crop and grain moisture selection routine 5325 5321 z\$="ON": CLS 5340 LOCATE 10,15: PRINT"You may select a maximum of 4 crops per simulation" 5350 LOCATE 11,15: PRINT" and a maximum of 5 moisture levels per crop." 5360 LOCATE 18,25: PRINT "strike any key to continue" 5370 KS=INKEYS 5380 IF LEN(K\$)=0 GOTO 5370 5381 cls 5400 LOCATE 7,20: INPUT"HOW MANY CROPS DO YOU WANT TO USE"; CROPS 5410 FOR I=1 TO CROPS 5420 GOSUB 5560 'display crop codes for the user

```
5430 LOCATE 9,20: INPUT"ENTER THE CROP"; NUM(I)
5440 IF NUM(I)>4 THEN BEEP: CLS: LOCATE 9, 20: PRINT" there are 4 crops and
     no more":LOCATE 10,20:INPUT"ENTER THE CROP AGAIN";NUM(I)
5450 LOCATE 11, 20: INPUT"ENTER THE CROP YIELD"; YIELD(I)
5460 IF NUM(I)=4 THEN GOSUB 5780
                                          'enter crop properties
5470 CLS:LOCATE 7,20:INPUT"HOW MANY MOISTURE LEVELS ": MOISTURES
5480 FOR J=1 TO MOISTURES
5490 LOCATE 9+J, 20: INPUT"ENTER THE GRAIN MOISTURE LEVEL
     DESIRED": MOIST(I.J)
5500 NEXT J
5510 LOCATE 11+J, 20: INPUT" ENTER THE REFERENCE MOISTURE FOR CURVE
     ADJUSTMENT"; MOIST(I, J)
5520 NEXT I
5530 BEEP
5540 CLS
5550 RETURN
5551'5552' subroutine to display available crop data
5553'5560 CLS
5570 LOCATE 2,30: PRINT "(1)=california wheat"
5580 LOCATE 3,30: PRINT "(2)=north dakota barley"
5590 LOCATE 4,30: PRINT "(3)=north dakota wheat
5600 LOCATE 5,30: PRINT "(4)=you describe crop"
5610 RETURN
5611'5612' property labels
5613'5620 IF TEMP-1 THEN C$-"GRAIN MOISTURE"
5630 IF TEMP=2 THEN C$="ANGLE OF REPOSE"
5640 IF TEMP=3 THEN C$="GRAIN DENSITY"
5650 IF TEMP=4 THEN CS="CHAFF MOISTURE"
5660 IF TEMP=5 THEN CS="CHAFF FRICTION"
5670 IF TEMP=6 THEN CS="CHAFF LENGTH"
5680 IF TEMP=7 THEN C$="STRAW MOISTURE"
5690 IF TEMP=8 THEN CS="STRAW MODULUS"
5700 IF TEMP=9 THEN C$="STRAW DENSITY"
5710 IF TEMP=10 THEN C$="GRAIN: MOG RATIO"
5720 IF TEMP=11 THEN C$="CHAFF:MOG RATIO"
5730 IF TEMP=12 THEN CS="1/2 % CLEANER LOSS"
5740 IF TEMP=13 THEN C$="1 % WALKER LOSS"
5750 IF TEMP-14 THEN CS-"2 % TOTAL LOSS"
5770 RETURN
5771'5772' subroutine to enter crop properties
5773'5780 CLS: LOCATE 2, 20: PRINT" CROP PROPERTIES ROUTINE"
5790 FOR TEMP=1 TO 11
5800 GOSUB 5620
5810 LOCATE TEMP+5,15: PRINT CS
5820 NEXT TEMP
5830 LOCATE 6,40: INPUT "* ", CROP(NUM(I), 1)
5840 LOCATE 7,40: INPUT "* ",CROP(NUM(I),2)<br>5850 LOCATE 8,40: INPUT "* ",CROP(NUM(I),3)<br>5860 LOCATE 9,40: INPUT "* ",CROP(NUM(I),4)
5870 LOCATE 10,40:INPUT "* ",CROP(NUM(I),5)
5880 LOCATE 11,40:INPUT "* ",CROP(NUM(I),6)
```
5890 LOCATE 12,40:INPUT "* ",CROP(NUM(I),7) 5900 LOCATE 13,40:INPUT"* ",CROP(NUM(I),8) 2320 LOCATE 14,40:INPUT"* ",CROP(NUM(I),9)
5920 LOCATE 15,40:INPUT"* ",CROP(NUM(I),10)
5930 LOCATE 16,40:INPUT"* ",CROP(NUM(I),12) 5910 LOCATE 14,40:INPUT"* ",CROP(NUM(I),9) 5920 LOCATE 15,40:INPUT"* ",CROP(NUM(I),10) 5940 RETURN 5941' 3322' subroutine to input variation for each crop property 5943' 7500 CLS: LOCATE 2, 20: PRINT" CROP VARIATION ROUTINE" 7510 FOR TEMP-1 'IO 9 7520 GOSUB 5620 8000 LOCATE TEMP+5,15: PRINT C\$ 8010 NEXT TEMP 8020 LOCATE 6,40:INPUT "* ",GMOSTVAR 8030 LOCATE 7,40:INPUT "* ",GANGVAR 8040 LOCATE 8.40:1NPUT "* ",GDENVAR 8050 LOCATE 9,40:INPUT "* ",CMOSTVAR 8060 LOCATE 10,40:INPUT "* ",CFRICTVAR 8070 LOCATE 11,40: INPUT "* ", CMLNTGVAR 8080 LOCATE 12,40:INPUT "* ",SMOSTVAR 8090 LOCATE 13,40:INPUT"* ",SMODVAR 8095 LOCATE 14,40: INPUT"* ", SDENVAR 8400 RETURN 9000 CLOSE:END 9001 RETURN
APPENDIX E

 ~ 10

 $\ddot{}$

 \sim

 \sim \sim

Combine Simlation Interactive Session

SIMULATION INSTRUCTIONS AND SAMPLE PROGRAM OUTPUTS

The simulation program was written in Microsoft Basic and implemented on an Ian-compatible micro—computer which utilized an Intel 8086 central processing unit. The simulation of a single loss curve required approximately five minutes to complete when the program was executed as interpreted code. The execution time was reduced to approximately one minute per curve simulation by compiling and linking the source code into a single executable module.

The documentation of the program is contained within the source code. The major variables are explained in a block of comment lines at the beginning of the program.

Function keys (F1 - F7) are used to select program options from a menu display. The inputs to the program are a series of crop properties and the random variation of each property expressed as a percentage, grain to M.O.G. and chaff to 11.0.6. ratios, ground speed (mph), header width (feet), and crop yield (bushels). The simulation provides options to display cleaner, straw walker, and total loss curves and a means to construct scatter plots of each property or machine parameter expressed as a function of another property or machine parameter.

The following text and figures describe the execution of the combine simulation program, Throughout the instruction, input from the user will be highlighted. Some inputs must be terminated by pressing the RETURN key which is denoted as (RET). The instructions assume that you are already familiar with the MS-DOS operating system and are able to boot the computer and begin the execution of a program. A typical interactive session begins as follows:

154

2. Wait for the following screen display:

F1. select crop location F2. run sinulation F3. plot loss curves F4. plot scatter plots of properties F5. set parameters for stochastic process F6. display historical data from simulation run F7. exit program

3. Press ftmction key Fl. At this point it is necessary to select the number of machine performance curves to simulate and the crop properties values which describe a crop. This example will select one set of crop properties and three moisture conditions to sinulate. The following message is displayed:

> You may select a maximum of 4 crops per simulation and a maximum of 5 moisture levels per crop. Strike any key to continue. Press any key.

How many crops do you want to use? Enter 1 (REF)

Select the crops you wish to use. Enter 4 <RET> Note: You may choose crop properties which are representative of a California wheat crop, a North Dakota wheat crop, or a North Dakota barley crop. You may also elect to describe the crop by its properties. If you are unfamiliar with the range of crop property values reference Table 2 on page of Chapter IV.

Crop Properties Routine .

How many moisture levels? Enter 3 <RET>

Enter the grain moisture level desired? 12 (BET) Enter the grain moisture level desired? 13 <RET> Enter the grain moisture level desired? 14 (BET) Enter the reference moisture level? 12 (BET)

4. It is now necessary to select the amount of variation for each crop as a percentage. Press F5.

Crop variation Routine

Grain Moisture 1.0 Angle of Repose 1.0 Grain Density 1.0 Chaff Moisture 1.0 Chaff Friction 1.0 Chaff Length 1.0 Straw Moisture 1.0 Straw Modulus 1.0 Straw Density 1.0

Note: The simulation of cleaner performance appears most realistic when chaff and grain properties vary by approximately 5.0 percent. Likewise, walker performance appears most realistic when the properties in the walker loss equation vary by 10.0 percent.

5. Press F2 to simulate the performance of a combine harvester. 6. Press F3 to produce loss curves for the cleaner, the walker, or the total loss curve. You will be prompted for the curves to display from a menu or select the first five performance curves. For this example press (BET), default to display the performance curves for one crop at three moisture levels. You may toggle between each type of curve by pressing the space bar. The display begins with cleaner curves. Press the space bar to display the walker curves. Each time you press the spacebar another set of curves is displayed.

7. Press F4 to create a scatter plot of one property versus another property or to plot the machine performance versus a property. For example, to plot grain density versus grain moisture, select the appropriate property from the display by entering the number of the property at the prompt(Figure 21).

Enter a property. 1 <RET>

Enter a property. 3 (RET)

8. Press F6 to display a summary of the crop property mean values, the coefficients of the predicted loss curves, and the predicted feed rates. Figure 22 is an example of the statistics provided for each simulation run. For example, there are three such summaries created by this example set of program inputs. Press the spacebar to display the next summary.

9. Press F8 to exit the program and return to DOS.

BY PRESSING THE F3 FUNCTION KEY

158

BY PRESSING THE F3 FUNCTION KEY

159

strike any key to continue

 $\frac{1}{2}$

CALIFORNIA WHEAT

l,

 $\ddot{}$

FIGURE 52

EXAMPLE OF SIMULATION STATISTICS DISPLAYED
BY PRESSING THE F6 FUNCTION KEY EXAMPLE OF SIMULATION STATISTICS DISPLAYED BY PRESSING THE F6 FUNCTION KEY

 $\frac{1}{2}$

BIBLIOGRAPHY

 $\mathcal{L}^{\text{max}}_{\text{max}}$ and $\mathcal{L}^{\text{max}}_{\text{max}}$

 \sim α

 $\hat{\mathcal{L}}$

 $\sim 10^7$

 $\label{eq:2.1} \frac{1}{\sqrt{2}}\int_{\mathbb{R}^3}\frac{1}{\sqrt{2}}\left(\frac{1}{\sqrt{2}}\right)^2\frac{1}{\sqrt{2}}\left(\frac{1}{\sqrt{2}}\right)^2\frac{1}{\sqrt{2}}\left(\frac{1}{\sqrt{2}}\right)^2\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac{1}{\sqrt{2}}\frac$

BIBLIOGRAPHY

- Agricultural Engineers Yearbook. 1983. American Society of Agricultural Engineers. St. Joseph, Michigan.
- .Allen, R. R., and L. D. Hollingsworth. 1981. Combine header performance in lodged grain sorghum. Transactions of the ASAE 24(6):1426—1428, 1431.
- .Arnold, R. E. 1964. EXperiments with rasp bar thresh-ing drums: some factors affecting performance. Journal of Agricultural Engineering Research 9(2):99—l34.

المستحقة

- Boyce, B. R., T. R. Pringle, and B. M. D. Wills. 1974. The separation characteristics of a combine harvester and a comparison of straw walker performance. Journal of Agricultural Engineering Research 9:77-84.
- Cervinka, V. 1974. Multiple regression analysis of combine harvester design and operational parameters. Transactions of the ASAE 17:221-224.
- Chung, D. s. and H. H. Converse. 1965. Effect of moisture content on some physical properties of grain. Transactions of the ASAE 14(4):612-614,620.
- Claar, P. W. et al. 1982. Simulated tractor chassis suspension system. Transactions of the ASAE 25(3):590-594.
- Cooper, G. F., J. H..A. Lee, and'w. H. Knapp. 1969. Standard Terminology for combines and grain harvesting. ASAE Paper No. 69-650. Chicago, Illinois.
- Cooper, 6. F. 1977. Cylinder/Concave performance from laboratory tests-II. Proceedings, First International Grain and Forage Harvesting Conference. Ames, Iowa, pp. 101-103.Cooper, G. F. 1981.
How crops behave in a laboratory environment. ASAE Paper No. 81-1565.
- Chicago, Illinois.
- DeKoning, X. 1973. Measurement of some parameters of different spring wheat varieties affecting combine harvesting losses. Journal of .Agriculutural Engineering Research 18:107-115.
- Dodds, M. E., and F. W. Bigsby. 1968. The breakup of wheat straw by combine cylinders. Canadian Agricultural Engineering 10(1):43,44.
- Griffin, G. A. 1973. Fundamentals of Machine Operation combine Harvesting. John Deere Service Publications. Moline, Illinois.
- German, R. F. and J. H. Lee. 1969. Grain separation on an oscillating sieve as affected by air volume and frequency. Transactions of the ASAE 12(6):883-885. '
- Hall, C. W. and G. G. Zoerb. 1960. Some mechanical and rhelogical properites of grains. Journal of Agricultural Engineering Research $5(1)$:83–93.
- Hall, J. W. and J. F. Huisman. 1981. Correlating physical properties with combine performance. ASAE Paper No. 81-3538.
- Haman, J. S. 1977. Influence of some physical properites of grain on harvesting conditions. Proceedings, First International Grain and Forage Harvesting Conference. Ames, Iowa. pp. 69-73.
- Huisman, W. 1977. Moisture content, coefficient of friction and modulus of elasticity of straw in relation to walker lossees in a combine harvester. Proceedings, First International Grain and Forage Conference. Ames, Iowa. pp. 49-54.
- Huisman, W., J. J. Heining, J. van Loo and O. C. Bergman. 1974.
Automatic feed-rate control on a combine harvester. Paper 74-111-106, GIGR Conference. Holland.
- Huisman, W. 1983. Optimum cereal combine harvester operation by means of automatic machine and threshing speed control. Doctoral Thesis. .Agricultural University, wageningen, Netherlands.
- Huynh, v; M. and T. E. Powell. 1978. Cleaning shoe performance prediction. ASAE Paper No. 78-1565.
- Kirk, T. G. et a1. 1977. Evaluation of ^a simulation model of the combine harvester. Proceedings, First International Grain Forage Harvesting Conference. Ames, Iowa. pp. 23-27.
- Kumar, R. and J. R. Goss. 1981. Computer simulation of harvesting alfalfa seed. Transactions of the ASAE 24(5):1135—1140.
- Manetsch, T. J. and G. L. Park. 1982. System Analysis and Simulation with Applications to Economic and Social Systems. Michigan State university, Department of Electrical Engineering.
- Mark, A. H., J. M. Godlewski and J. L. Coleman. 1963. Evaluating combine performance: a global approach. Agricultural Engineering 44(3):136-137.
- Moshenin, N. N. 1980. Physical properties of plant and animal materials. Gordon and Breach Science Publishers. pp.583-584.
- Murray, D. A. et a1. 1977. Recent development in grain threshing and separating mechanisms. Proceedings, First International Grain and Forage Harvesting Conference. Ames, Iowa. pp. 178-185.
- Neal, A. E. and G. F. Cooper. 1968. Performance testing of combines in the lab. Agricultural Engineering **:397-399.
- Nil. N. H. et al. 1975. SPSS: Statistical Package for the Social Sciences. Second Edition. McGraw—Hill, New York.
- Nyborg, E. O. 1964. A test procedure for determining combine capacity. Canadian Agricultural Engineering 6(1):8-10.
- Nyborg, E. O. 1969. Grain combine loss characteristics. Transactions of the ASAE 12(6):727-732.
- Oxley, T. A. 1944. The properties of grain in bulk. Transactions of the Society of Chemical Industry 63:53-57.
- Ridenour, H. E. 1968. Combines and Combining. The Ohio Agricultural Education curriculum Materials Service.
- Reed, W. B., G. C. Zoerb and F. W. Bigsby. 1974. A laboratory study of grain-straw separation. Transactions of the ASAE 17:452-460.
- Scheuller, J. K. 1985. The current status of automation on self-propelled grain combines manufactured in North America. Proceedings of the Agri-mation. Chicago, Illinois. pp. 307-311.
- Turner, R. J. 1985. The development of ^a corn harvesting combine simulator. ASAE Paper No. 85-1578.
- van Loo, I. J. 1977. An automatic feedrate contol system for combine harvester. Wageningen University Department of Agricultural Engineering, Netherlands.
- Vas, f. M. and H. P. Harrison. 1969. The effect of selected mechanical threshing parameters on kernel damage and threshability of wheat. Canadian Agricultural Engineering 11(2):83—87.
- Wrubleski, P. D. and E. O. Nyborg. 1977. Prairie Agricultural Institute fiekl evaluation of grain combines. Proceedings, First
Internation Grain and Forage Conference. Ames, Iowa. pp. 118-123.
- Wrubleski, P. D. and L. G. Smith. 1980. Separation characteristics of conventional and non-conventional grain combines. Transactions of the ASAE 23(3):530-534.

 \bullet