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Use of Spectral Radiance to Estimate In-Season Biomass and Grain Yield in Nitrogen- and Water-Stressed Corn

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Use of Spectral Radiance to Estimate In-Season Biomass and Grain Yield in Nitrogen- and Water-Stressed Corn

S. L. Osborne*, J. S. Schepers, D. D. Francis, and M. R. Schlemmer

while recently remote sensing techniques for estimating N status have **increased with limited research on the interaction between the two**
try, other leaf water content measurements, and canopy
temperature by infrared thermometry. Visual indica**stresses. Because plant water status methods are time-consuming and** temperature by infrared thermometry. Visual indicarequire numerous observations to characterize a field, managers could
benefit from remote sensing techniques to assist in irrigation and N
management decisions. A 2-yr experiment was initiated to determine
specific wavelen **The experiment was a split-plot design with three replications. The** cial to use remote sensing techniques to help managers **The experiment was a split-plot design with three replications. The** cial to use remote sensing **treatment structure had five N rates (0, 45, 90, 134, and 269 kg N** determine when and/or where a water stress exists. Early **ha¹) and three water treatments [dryland, 0.5 evapotranspiration** detection of a water stress could trigger irrigation before **(ET), and full ET]. Canopy spectral radiance measurements (350–2500** yield loss occurs.
 nm) were taken at various growth stages (V6–V7, V13–V16, and Recent researc nm) were taken at various growth stages (V6–V7, V13–V16, and

V14–R1). Specific wavelengths for estimating crop biomass, N concentration, grain yield, and chlorophyll meter readings changed with

growth stage and sampling **green and NIR regions were used to estimate total N and biomass** stressed canopies to have a lower spectral reflectance without water-
green and NIR regions were used to estimate total N and biomass stressed canopies to without water stress. Reflectance at 510, 705, and 1135 nm were in the NIR and red wavebands when compared with found for estimating chlorophyll meter readings regardless of year or unstressed canopies. A ratio of the two found for estimating chlorophyll meter readings regardless of year or **sampling date.** most successful in estimating the onset of stress. Moran

low available soil N can be an additional yield reduction partially-vegetated sites. The measurement can be calfactor. Denmead and Shaw (1960) found water stress culated from remotely sensed data (red and NIR) gathreduced corn grain yield by 25, 50, and 21% prior to, dur- ered with ground, aircraft, or satellite-based sensors.

brought about an awareness of efficiently utilizing our to investigate the relationship between canopy temperamental impacts of fertilizer leaching and runoff. Irriga-
tion management techniques involve several different
methods, including soil-, meteorological-, and crop-based
techniques. Soil-based techniques target irrigation e

ABSTRACT as crop coefficients, to estimate daily ET. Direct plant-**Current technologies for measuring plant water status are limited,** based measurements are limited to leaf water potential

et al. (1994) investigated the concept of a water deficit index, which is defined as the ratio of actual to potential AVAILABLE WATER is one of the most limiting factors ET. This index exhibits the ability to predict ET rate
and relative field water deficit for both full-cover and
law smillalaceil N son he and ditional violation and inter ing, and after silking, respectively, compared with the On-site measurements used in the calculation include nonstressed plots. Claasen and Shaw (1970) concluded net radiation, air vapor pressure deficit, air temperature, net radiation, air vapor pressure deficit, air temperature, that water stress at 75% silking resulted in greater reduc-
tion in corn grain yield (53% of the nonstressed), com-
(*Gossypium hirsutum* L.) showed reflectance in the in-(*Gossypium hirsutum L*.) showed reflectance in the inpared with stress during vegetative periods or 3 wk after frared spectra (810, 1665, and 2210 nm) increased as silking (15 and 30% of the nonstressed, respectively). relative water content decreased. Hope et al. (1986) lking (15 and 30% of the nonstressed, respectively). relative water content decreased. Hope et al. (1986)
Over the past decade, the increased use of irrigation integrated canopy reflectance, stomatal resistance, and Over the past decade, the increased use of irrigation integrated canopy reflectance, stomatal resistance, and and concern over environmental contamination has energy dynamics into a modeling system that was suited energy dynamics into a modeling system that was suited limited resources and decreasing the negative environ-
mental indices, that is, simple spectral ratio
mental impacts of fertilizer leaching and runoff. Irriga-
(NIR/red) and normalized difference vegetation index rological-based approaches depend on air temperature, and linear combinations of bands similar to those meanet radiation, vapor pressure, and wind speed, as well sured by sensors on the Landsat satellite. They determined water stress could not be detected until after

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tory, 2923 Medary Ave., Brookings, South Dakota 57006; J.S. Schep-
ers, D.D. Francis, and M.R. Schlemmer, Dep. of Agronomy and USDA- ability of all ratios to ARS, Univ. of Nebraska-Lincoln, Lincoln, NE 68583. Received 12 Dec. 2000. *Corresponding author (sosborne@ngirl.ars.usda.gov).

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Abbreviations: ET, evapotranspiration; MIDIR, midinfrared; NIR,

on the plant growth stage, soil background, and atmo- wavelengths and/or combinations of wavelengths indicspheric changes. The spheric changes ative of water stress and N deficiencies, and to evaluate

must be developed to assess the crop fertilizer N needs. corn grain yield. Methods used to detect N deficiencies during the cropping season include: chlorophyll meter readings, de-
structive plant sampling, and soil sampling. Blackmer structive plant sampling, and soil sampling. Blackmer
and Schepers (1994) found the chlorophyll meters were
useful for monitoring N status in irrigated corn. Nor-
malizing the readings to those from the highest N treat-
me of identifying excessive N availability during the later part of the growing season. The stalk nitrate test is plot with three replications. The treatment design was a three not practical during the growing season. Blackmer and by five factorial arrangement, with water treatment not practical during the growing season. Blackmer and by five factorial arrangement, with water treatment being the
Schepers (1996) found photographic data collected at whole-plot treatment and N rate as the split-plot tre Schepers (1996) found photographic data collected at θ R4 (dough) were more highly correlated with grain yield
than was chlorophyll meter data. They also found photo fertilizer (13 kg N ha⁻¹ and 19 kg P ha⁻¹ as ammonium polypthan was chlorophyll meter data. They also found photo-
graphic data could detect N deficiencies, which were
later verified using stalk nitrate-N sampling. Although
these methods have proven to be good indicators of N
str sample collection. The use of effective remote sensing techniques would eliminate the need for extensive field (1/2 sidedress at V1–V2 and 1/2 sidedress at V4–V5). Applica-
sampling, giving good deficiency detection capability, tions dates were 21 May and 10 June 1997, and 22

nm was best for separating N-deficient from non-N- late August. Chlorophyll meter readings (SPAD 502, Minolta deficient corn and sweet pepper (*Capsicum annuum* L.) Corp., Ramsey, NJ) were taken on the uppermost collared leaves, respectively. Bausch and Duke (1996) investigated until the ear leaf was present. Once the ear leaf was present of the ear leaf was present of the ear leaf was present of the ear leaf was used for the remainder of gated using a ratio of NIR/green reflectance as an N-
sufficiency index. Comparison of this method to the
SPAD chlorophyll meter measurements (Minolta Corp.,
Ramsey, NJ) demonstrated that the NIR/green ratio
and SPAD measu showed that N-deficient corn canopies had increased
red reflectance and decreased NIR reflectance when
compared with N-sufficient corn canopies. A ratio of
1997 and 1998. the average reflectance from 760 to 900 nm divided by reflectance from 630 to 690 nm exhibited a good separation between N treatments. Aase and Tanaka **mm** (1984) reported a relationship between green leaf dry matter and NIR/red ratios, and suggested that reflectance measurements could be used to estimate leaf dry matter or leaf area measurements in spring and winter wheat *(Triticum aestivum L.)*. Work by Stone et al. (1996) demonstrated that total plant N could be estimated using spectral radiance measurements at the red **mm** (671 nm) and NIR (780 nm) wavelengths. They calculated a plant-N-spectral-index for the amount of fertilizer N required to correct in-season N deficiency in winter wheat. **5** Sept. 1998 – 5 22

The objectives of this work were to determine specific $\ddot{\tau}$ **ET** = evapotranspiration.

In order to efficiently apply N, in-season methods these wavelengths for estimating in-season biomass and

of 74 000 plants ha⁻¹. The experimental design was a split-Plots were 7.32 m \times 22.86 m with 0.91-m row spacing. Starter N ha^{-1}. Nitrogen treatments were applied as a split application sampling, giving good deficiency detection capability. tions dates were 21 May and 10 June 1997, and 22 May and
Increased scientific understanding of spectral re-
16 June 1998. Guardsman [Dimethenamid (2-chloro-N-[(1-Increased scientific understanding of spectral re-
sponses of crops is increasing the potential for using
remote sensing to detect nutrient stresses. Recent re-
search has focused on determining the appropriate
wavelength

	Water treatment						
Date	Dryland	$0.5 E T+$	1.0 ET				
		mm					
30 June 1997	22	22	22				
9 July 1997	14	14	14				
10 July 1997		14	29				
15 July 1997			28				
18 July 1997		27	33				
23 July 1997			36				
31 July 1997	37	36	38				
8 Aug. 1997		25	30				
		mm					
29 June 1998			16				
3 July 1998			14				
21 July 1998			30				
22 Aug. 1998			24				
5 Sept. 1998		5	22				

from the 350- to 2500-nm range, collected at 1-nm intervals gust in 1997, and 13 July and 16 September in 1998. Whole using a FieldSpec spectroradiometer (Analytical Spectral De- plants were weighed and then chopped with a chipper-shredvices, Boulder, CO). Measurements were taken at heights of der in the field to facilitate subsampling. Subsamples and ear 4.5 m and 10.6 m above the canopy, in 1997 and 1998, respec-
amples were weighed and then oven dried 4.5 m and 10.6 m above the canopy, in 1997 and 1998, respec-
tively, with an 18° field of view foreoptic. Measurement dates being reweighed to determine water content. All subsamples tively, with an 18° field of view foreoptic. Measurement dates being reweighed to determine water content. All subsamples were 17 July and 17 September in 1997 and 15 June and 17 were ground in a Wiley Mill to pass a 2-mm were 17 July and 17 September in 1997, and 15 June and 17 were ground in a Wiley Mill to pass a 2-mm sieve. Nitrogen
Iuly in 1998 Five canopy measurements were taken randomly concentration was determined on all subsamples July in 1998. Five canopy measurements were taken randomly
throughout each plot. Individual readings were averaged to
get a representative reading for the entire plot. Canopy mea-
surements were taken on cloud-free days a ing radiance. Reference panel measurements were taken before the initial canopy reading and repeated about every 15 min. **RESULTS AND DISCUSSION**

Spectral readings collected on the specified dates were aver-
aged across 5-nm intervals to decrease the amount of data for
analysis. Stepwise regression was performed, using the REG There was a significant linear

irrigation treatments (averaged across N treatments), the plots. Whole plants were weighed and ears were separated once distinguishable. Sampling dates were 15 July and 19 Au- respectively, compared with the full ET treatment in

analysis. Stepwise regression was performed, using the REG analysis. Stepwise regression was performed, using the REG reprocedure in SAS (SAS, 1988) on all data to develop multiple regression equations for predicting plant Biomass samplings were performed during the growing sea-
son by randomly collecting 12 plants from the east quarter of yields by 22.1 and 15.8% for the dryland and 0.5 ET 1997 (Table 2). These results are similar to previous $\frac{1}{1}$ Mention of trade name or proprietary products does not indicate findings from Denmead and Shaw (1960), who found a

*** Significant at the 0.05 probability level.**

**** Significant at the 0.01 probability level. *** Significant at the 0.001 probability level.**

† LOF, lack of fit.

endorsement of the USDA and does not imply its approval to the 25% reduction in yield with water stress prior to silking. exclusion of other products that may also be suitable. The dryland irrigation treatment resulted in a higher N

surements were correlated with changes in leaf N con-
centration. Problems associated with comparing read-
rates on chlorophyll content due to the extreme water ings from different growth stages, different hybrids, or stress present at that date. Immediately following chlodifferent locations make it difficult to make comparisons rophyll meter readings, all plots were irrigated to prethroughout the season (Schepers et al., 1992). Chloro- vent total crop loss for the dryland treatments (Table 1).

concentration relative to the 0.5 and full ET treatments.

There was no water stress present during the 1998 sea-

son, thus no effect of water on grain yield or total N.

Rainfall amount for the 1997 season was 247 mm, c N interaction for grain yield
and total N (Table 2). The 1999 season water treatments. Differences in chlorophyll content for
three water treatments were greatest between the 0 N
rate compared with the remaining N rates (

Chlorophyll Meter Readings Chlorophyll Meter Readings Chlorophied N were significant for all weekly Past research has shown that chlorophyll meter mea-
surements were correlated with changes in leaf N con-
from 31 July 1997 suggested a significant effect of N rates on chlorophyll content due to the extreme water

Source	df	13 June	25 June	3 July	10 July	16 July	24 July	31 July	7 Aug.	14 Aug.
Replication		ns ₁	ns	ns	ns	ns	ns	ns	ns	ns
Water		ns	ns	ns	ns	ns	ns	*	ns	ns
Linear		ns	ns	ns	ns	ns	ns	*	ns	ns
LOF:		ns	ns	ns	ns	ns	ns	ns	ns	ns
Error (a)	4									
N rate	4	***	***	***	***	***	***	***	***	***
Linear		***	***	***	***	***	***	***	***	***
Quadratic		ns	***	***	***	***	***	***	***	***
Cubic		ns	ns	ns	ns	*	***	***	**	ns
LO		ns	ns	ns	ns	ns	ns	ns	ns	ns
N rate \times water	8	ns	ns	ns	ns	ns	ns	ns	*	ns

Table 3. Significance level from the analysis of variance, and single-degree-of-freedom contrasts for chlorophyll meter readings, Shelton, NE, 1997.

*** Significant at the 0.05 probability level.**

**** Significant at the 0.01 probability level.**

***** Significant at the 0.001 probability level.**

 \dagger ns $=$ not significant.

‡ LOF, lack of fit.

Table 4. Significance level from the analysis of variance, and single-degree-of-freedom contrasts for chlorophyll meter readings, Shelton, NE, 1998.

Source	df	17 June	24 June	30 June	8 July	15 July	27 July	4 Aug.	12 Aug.
Replication		nsi	ns	ns	ns	ns	ns	ns	ns
Water		ns	ns	ns	ns	ns	ns	ns	ns
Linear		ns	ns	ns	ns	ns	ns	ns	ns
LOF:		ns	ns	ns	ns	ns	ns	ns	ns
Error (a)									
N rate		***	***	***	***	***	***	***	***
Linear		***	***	***	***	***	***	***	***
Quadratic		***	***	***	***	***	***	***	***
Cubic		***	***	***	ns	**	**	ns	ns
LO		ns	**	ns	\mathbf{ns}	ns	ns	ns	ns
N rate \times water	\bullet	ns	\mathbf{ns}	ns	ns	ns	ns	ns	ns

**** Significant at the 0.01 probability level.**

***** Significant at the 0.001 probability level.**

† ns, not significant. ‡ LOF, lack of fit.

Two separate analyses were performed for the 17 July
1997 sampling due to differences between water stress
treatments. One analysis was performed for all treat-
treatments, while the other analysis was performed on only
th dicting total N and biomass with and without a water
stress. Results from water-stressed plants indicated re-
flectance in the red (600 and 700 nm) and a number (Table 5). flectance in the red (600 and 700 nm) and a number
of different wavelengths in the midinfrared (MIDIR),
between the water absorption bands (1400, 1900, and
2700 nm), were important for predicting biomass (R^2 = 0.87) (T 0.87) and total N ($R^2 = 0.87$) (Table 5). Reflectance in the MIDIR (1135, 1145, 2095, and 2110 nm) regions, with the MIDIR region is considered to be a function of MIDIR (1135, 1145, 2095, and 2110 nm) regions, with leaf thickness and water content (Lillesand and Kiefer, an $R^2 = 0.91$ (Table 5). Because plants had been under 1987). These wavelengths are similar to wavelengths a water stress for 2 wk during the 1997 growing season, used by past researchers to detect water stress (Moran the production of chlorophyll and plant growth were
et al. 1989: Carter 1991) Incontrast results from plants reduced. The decrease in growth, as well as differences et al., 1989; Carter, 1991). In contrast, results from plants adequately supplied with water indicated wavelengths in water content, could account for the contribution in the green region (550 nm) and the MIDIR (1425, from reflectance in the NIR and MIDIR regions in 1997. in the green region (550 nm) and the MIDIR (1425, 1465, 1490, and 2120 nm) were important in the predic-
tion of total N, having an $R^2 = 0.95$ (Table 6). Biomass data involved reflectance in the green (510 nm), NIR tion of total N, having an $R^2 = 0.95$ (Table 6). Biomass data involved reflectance in the green (510 nm), NIR was best predicted by wavelengths in similar narrow (705 nm), and MIDIR (1135 nm) with $R^2 > 0.82$ (Tawas best predicted by wavelengths in similar narrow waveband regions (505, 515, 545, 1455, 2045, 2180, and ble 5). Because these wavelengths were similar to those 2190 nm) with an $R² = 0.94$ (Table 6). This is similar used for the 1997 data, our results indicate the potential to work by Blackmer et al. (1994) who found reflectance of remote sensing for estimating chlorophyll meter read-

Hyperspectral Readings corn. Everitt et al. (1985) found plant leaf N content
to analyzes were negated for the 17 July was associated with reflectance at 500 nm.

near 550 nm was best for detecting N deficiencies in ings, which in the past has been used to estimate fertil-

Date	Growth stage	\mathbb{R}^2	Biomass
			- Mg ha ⁻¹ -
17 July 1997	V13-V16	0.87	$v = 10 - 122 \times R_{700} + 295 \times R_{1015} - 48 \times R_{1110} - 473 \times R_{1205} + 107 \times R_{1335}$ $+356 \times R_{1510} - 306 \times R_{1580} - 235 \times R_{1640} + 498 \times R_{1645} + 96 \times R_{1785}$ $-9 \times R_{1795} - 95 \times R_{1960} + 97 \times R_{1965} - 176 \times R_{2975} - 94 \times R_{2185}$ $+38\times R_{240}$
15 June 1998	$V6-V7$	0.59	$v = 4 - 1 \times R_{975} - 17 \times R_{2000} + 24 \times R_{2005} + 34 \times R_{2040} - 21 \times R_{2055}$ $-27 \times R_{\text{max}} + 16 \times R_{\text{max}} - 14 \times R_{\text{max}}$
17 July 1998	V14-R1	0.45	$v = 7 - 35 \times R_{2240}$
			Plant N
			- g kg ⁻¹ -
17 July 1997	V13-V16	0.87	$v = 1 + 158 \times R_{600} - 91 \times R_{705} - 5 \times R_{1130} + 12 \times R_{1155} - 11 \times R_{1450}$ $-26 \times R_{1515} - 18 \times R_{1700} + 22 \times R_{1725} + 12 \times R_{1775} + 11 \times R_{2045}$ $-10 \times R_{2000} - 4 \times R_{2155} + 6 \times R_{2100}$
15 June 1998	V6-V7	0.80	$y = 5 + 5 \times R_{1150} + 53 \times R_{2045} - 37 \times R_{2060} + 51 \times R_{2065} - 80 \times R_{2110}$
17 July 1998	V14-R1	0.95	$v = 1 - 15 \times R_{715} + 2 \times R_{935} + 18 \times R_{1435} + 7 \times R_{1955} - 15 \times R_{1990}$ $-8 \times R_{2010} - 45 \times R_{2020} + 32 \times R_{2110} + 9 \times R_{2200}$
			Chlorophyll meter reading
17 July 1997	V13-V16	0.91	$y = 48 - 7048 \times R_{440} + 8808 \times R_{510} - 1897 \times R_{705} + 3 \times R_{1135} + 51 \times R_{1145}$ $+285 \times R_{2095} - 298 \times R_{2110}$
15 June 1998	V6-V7	0.82	$v = 48 + 1840 \times R_{510} - 862 \times R_{705} + 28 \times R_{1135}$
17 July 1998	V14-R1	0.87	$y = 46 + 1006 \times R_{510} - 808 \times R_{705} + 87 \times R_{1135}$

Table 5. Regression equations for predicting biomass, plant N, and chlorophyll meter readings in corn grown under five N levels and three irrigation levels from canopy hyperspectral data by sampling date, Shelton, NE, 1997 and 1998 $(n = 45)$.

Table 6. Regression equations for predicting biomass, and plant N for corn grown at high water levels at the V13-V16 phenological growth stage, Shelton, NE, 17 July 1997 (*n* **15).**

izer N need for a growing crop (Blackmer and Schep-
ences in N content and also differences in the amount
of plant biomass present. The wavelengths used in the

phenological growth stages (Table 7). Prediction of grain yield was best for the 17 July 1998 sampling (V14–R1), with an $R^2 = 0.95$ using reflectance in the green (530) nm), red (675 nm), and MIDIR (1780, 1790, 2020, and 2035 nm). The equation developed for the 17 Sept. 1997 2035 nm). The equation developed for the 17 Sept. 1997 **CONCLUSIONS**
sampling date (R5–R6) used reflectance values in simi-
lar regions of the spectrum, with an $R^2 = 0.89$. Reflec-
The potential for hyperspectral re lar regions of the spectrum, with an $R^2 = 0.89$. Reflec-
tance in these spectral regions would account for differ-
detect water stress in irrigated corn was illustrated the

ers, 1995).
Stepwise regression was performed to estimate grain of plant biomass present. The wavelengths used in the Stepwise regression was performed to estimate grain regression equations changed throughout the growing Stepwise regression was performed to estimate grain regression equations changed throughout the growing yield and grain N using hyperspectral data for the four season and especially between years, suggesting seaseason and especially between years, suggesting seasonal differences in the amount of soil background present, difference in vegetation stages, and presumably a number of other agronomic factors.

detect water stress in irrigated corn was illustrated the

First season (1997) of this study, but these results could
not be repeated due to the lack of water stress in the
1998 growing season. The presence of water stress influ-
1998 growing season. The presence of water stress i enced the wavelengths used for estimating plant N con-
tent compared with wavelengths used without the pres-
Everitt, J.H., A.J. Richardson, and H.W. Gausman. 1985. Leaf reflecence of a water stress. Predicting plant N and grain yield
resulted in an R^2 of 0.80 or better for each sampling date,
but the wavelengths were not always consistent across and relationships between spectral reflectanc but the wavelengths were not always consistent across lated relationships between spectral reflectance, thermal emissions,

phenological stages Reflectance in the green (510 nm) and evapotranspiration of a soybean canopy. phenological stages. Reflectance in the green (510 nm) , and evapotranspiration of a soub-22:1011–1019. NIR (705 nm), and MIDIR (1135 nm) regions predicted

chlorophyll meter readings with an *R²* > 0.82 for all

Sampling dates, indicating the potential of using hypers-

Jackson, R.D., P.N. Slater, and P.J. P sampling dates, indicating the potential of using hypers- Jackson, R.D., P.N. Slater, and P.J. Pinter, Jr. 1983. Discrimination pectral imagery as a tool to detect and map variations in

plant chlorophyll. While there are a number of different

specific wavelengths that were used to estimate N con-

Lillesand, T.M., and R.W. Kiefer, 1987. Remote se tent, biomass, and grain yield, and the specific equations interpretation. 2nd ed. John Wiley & Sons, New York.

developed may not be transferable to other areas the Moran, M.S., T.R. Clarke, Y. Inoue, and A. Vidal. 1994. developed may not be transferable to other areas, the
range of wavelengths could be useful in determining
which reflectance bands are important for detecting N
246–263.
The and spectral vegetation index. Remote Sens. Envir content, biomass, or grain yield with and without the Moran, M.S, P.J. Pinter, Jr., B.E. Clothier, and S.G. Allen. 1989. Effect
of water stress on the canopy architecture and spectral indices of
of water stress on the cano presence of a water stress at different times in the grow-
ing season.
Peterson, T.A., T.M. Blackmer, D.D. Francis, and J.S. Schepers. 1993.

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-
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-
-
- tance compared with other nitrogen stress measurement in corn
leaves. Agron. J. 86:934–938. Thomas, J.R., and G.F. Oerther. 1972. Estimating nitrogen content
Bowman, W.D. 1989. The relationship between leaf water status, g
- exchange, and spectral reflectance in cotton leaves. Remote Sens. Environ. 30:249–255.
- Carter, G.A. 1991. Primary and secondary effects of water content Effects of nitrogen nutrition on the growth, yield, and on the spectral reflectance of leaves. Am. J. Bot. 78:916–924. characteristics of corn canopies. Agr on the spectral reflectance of leaves. Am. J. Bot. 78:916–924.
-
-
-
-
-
-
- Lillesand, T.M., and R.W. Kiefer. 1987. Remote sensing and image interpretation. 2nd ed. John Wiley & Sons, New York.
-
-
- Using a chlorophyll meter to improve N management. NebGuide G93-1171-A Univ. of Nebraska Cooperative Extension, Univ. of **REFERENCES** Nebraska, Lincoln, NE.
- Aase, J.K., and D.L. Tanaka. 1984. Effects of tillage practices on soil Ritchie, S.W., J.J. Hanway, and G.O. Benson. 1997. How a corn plants develops. Spec. Pub. 48. Iowa State Univ. of Sci. and Tech. Coop.
	-
	-
	-
- and wheat spectral reflectance. Agron. J. 76:814-818.

and wheat spectral reflectance. Agron. J. 76:814-818.

Sax Institute. 1988. SAS/STAT procedures. Release 6.03 ed. SAS

Institute. 1988. SAS/STAT procedures. Release 6.
	- of sweet pepper leaves by reflectance measurements. Agron. J. 64:11-13.
	- Walburg, G., M.E. Bauer, C.S.T. Daughtry, and T.L. Housley. 1982.
Effects of nitrogen nutrition on the growth, yield, and reflectance