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**Remote sensing for detection of cotton aphid (Homoptera: Aphididae) and spider mite-
(Acari: Tetranychidae) infested cotton in the San Joaquin Valley**

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1 **ABSTRACT.** We explored remote sensing methods for their potential to distinguish
2 aphid (*Aphis gossypii* Glover) and spider mite-infested (*Tetranychus* spp.) cotton (*Gossypium*
3 *hirsutum* L.) from uninfested cotton. Field plots were established using selective and disruptive
4 pesticides to establish a range of aphid and mite populations over 2 years. Aerial and satellite
5 remote sensing data in 2003 and 2004 were supplemented with ground-based remote sensing
6 data, in 2004, and by ground-truthing of arthropod populations in both years. Mite and aphid-
7 infested cotton was detected using aerial data in the green and near-infrared (NIR) wavelengths
8 in 2003, with sub-economic threshold aphid population levels. At the time aerial data were
9 collected, mite populations peaked at 95% leaves infested and exceeded treatment threshold
10 levels of 30-50% leaves infested. However, the number of mites per leaf in the treatments was
11 low to moderate (32, 9, 4, 6 and 2 average mites/leaf). Moreover, cotton infested with cotton
12 aphids above economic threshold levels was consistently detected using NIR wavelengths from
13 the satellite data in 2004. Similarly, aphid-infested cotton was detected at both sub and supra-
14 economic threshold aphid levels using NIR wavelengths from the ground-based remote sensing
15 data. Finally, accumulated mite-days were linearly correlated with a canopy, false color, and a
16 vegetation index using satellite data in 2004. Wavelengths in the NIR were fair to moderately
17 accurate predictors of aphid and mite infested cotton.

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19 **KEY WORDS.** Accumulated aphid-days, accumulated mite-days, Integrated Pest Management,
20 near-infrared, remote sensing.

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Introduction

Cotton aphids (*Aphis gossypii* Glover) and spider mites (*Tetranychus* spp.) are economically important pests of cotton (*Gossypium hirsutum* L.) in the San Joaquin Valley (SJV) of California. Cotton aphids feed by piercing the phloem, usually on the abaxial surface, and removing the assimilates. The carbohydrates from the phloem are metabolized by the aphids and are excreted, often deposited on the leaves of the cotton plant as a complex sticky substance referred to as honeydew. Honeydew on the leaves causes them to become glossy, as a result of the increase in light reflected off the honeydew, and may cause lint contamination, commonly called “sticky cotton”. Spider mites feed on the abaxial surface of cotton leaves by piercing plant cells with their stylets and assimilating the contents. The chloroplasts are removed, leaving the dead cell, which appears brown. Damage proceeds from small yellowish to reddish spots on the leaves, to leaf bronzing, and, in the case of a serious infestation, complete defoliation.

Cotton aphid management is most critical after the open boll stage of cotton in order to protect the exposed lint. The economic threshold for aphids before the open boll stage is 50-75 aphids per leaf for at least 7-10 days (5th mainstem node leaf from the top) (Godfrey et al. 2005a) and this threshold declines to 5 aphids per leaf after boll opening (Godfrey et al. 2004, 2005b). Although economic damage does not occur until 80% of the leaves are infested (6-9th mainstem node leaf from the top) (Wilson et al. 1991) with spider mites, the treatment threshold is set at 30-50% infested leaves (5th mainstem node leaf from the top) to allow time for the acaricide to manage the infestation before economic loss (Godfrey et al. 2005a).

Remote sensing is a precision technology, with roots in the aerospace and defense sectors, and acquires information about objects without being in contact with the objects. Airborne sensors can be used in agriculture, by capitalizing on their ability to detect light

47 reflected off of large areas of vegetation. Vegetation indices can be created from the data by
48 combining spectral measurements from different wavelengths; these are useful because they
49 convert multidimensional data into a single value.

50 Several studies in cotton have evaluated the potential of remote sensing methods to
51 facilitate arthropod pest management (Everitt et al. 1994, 1996, Summy et al. 1997, Brewster et
52 al. 1999, Fitzgerald et al. 1999, 2000, 2004, Willers et al. 1999, Riedel et al. 2001, Sudbrink et
53 al. 2003). Several studies found the wavelengths in the NIR to be the best indicators of
54 arthropod-damaged cotton (Everett et al. 1994, 1996, Summy et al. 1997, and Fitzgerald et al.
55 1999, 2000, 2004). Although the study done by Brewster et al. (1999) included a ground survey
56 and classification of crop types most likely to be attacked by whiteflies (*Bemisia argentifolii*
57 Bellows & Perring) and the study done by Sudbrink et al. (2003) included a visual rating of pest
58 damage, only the studies done by Summy et al. (1997) and Willers et al. (1999) incorporated
59 extensive ground-based insect monitoring (ground-truthing). For remote sensing to be a useful
60 management tool for Integrated Pest Management (IPM), consideration must be given to
61 threshold levels of arthropod infestations.

62 Many arthropod pests are distributed among patches within landscapes (Taylor 1984,
63 Hughes 1996). Both spider mites and aphid infestations begin in spatially heterogeneous areas
64 within fields (Hannah et al. 1996; L.D. Godfrey, unpublished data). In cotton, these infestations
65 begin as small “hot spots” and grow into larger infestations (Steinkraus et al. 2003; L.D.
66 Godfrey, unpublished data). Because of this spatial heterogeneity, the appropriate scale of
67 detection must be applied to reveal the distribution pattern of the aphid or mite infestation within
68 the field (Holland et al. 1999, Pearce and Zalucki 2006). Consequently, threshold levels of

69 arthropod infestations must be considered with regard to the spatial scale of the remote sensor
70 that is used.

71 The goal of this study was to incorporate ground-truthing with remote sensing methods
72 and to explore the application of this technique for cotton aphid and spider mite monitoring in
73 SJV cotton IPM. It was hypothesized that these arthropods, at sub-threshold levels, would
74 change leaf reflectance and that this could be detected using remote sensing. Both of these pests
75 occur in heterogeneous areas of the field, but current monitoring and treatment are done on an
76 intra-field basis. Using remote sensing in combination with ground-truthing, it may be possible
77 to shift these management efforts to an inter-field basis, which would potentially save resources
78 and protect the environment by limiting whole-field pesticide applications.

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Materials and Methods

81 **Field preparation.** Cotton was planted at ca.13 seeds/m on 5 May 2003 (Maxxa) and 26
82 April 2004 (NemX) at the University of California (UC) Shafter Research and Extension Center
83 near Shafter, CA. Twenty plots (~15.2 m long x ~8.1 m wide [8 rows]) were arranged in a
84 randomized complete block design (4 replications per treatment) with resident populations of
85 cotton aphids (*Aphis gossypii* Glover) and spider mites (*Tetranychus* spp.). Spider mites are a
86 perennial pest of cotton in the SJV and populations normally develop by July. In 2003, however,
87 spider mite populations did not develop until mid-August, while spider mite populations
88 developed in July 2004. Additionally, the spring in 2003 was unseasonably wet and the cotton
89 was planted later than usual. By mid-July 2003, there was only >1 aphid per leaf (5th mainstem
90 node leaf from the top of the plant). In contrast, aphid populations were ca.12 aphids per leaf by
91 mid-July 2004.

92 Cotton aphids have been a severe pest of the SJV during the last ten years. Because
93 aphid populations are favored by high nitrogen levels (Cisneros and Godfrey 2001), all the plots
94 were fertilized with 252 kg N/ha of ammonium sulfate (21-0-0) on 24 June and liquid urea
95 (Un32) on 9 July in 2003 to favor population development (this amount is on the upper end used
96 by growers). A more commonly used fertilizer regiment was employed in 2004, because aphids
97 were more abundant than in 2003. Thus, the fields were fertilized according to UC
98 recommendations and a nitrogen Un32 application was applied at 90 kg N/ha on 24 June 2004
99 (Basset et al. 1996). No potassium or phosphorus was applied to the field.

100 Weed control and furrow irrigation followed the recommendations of the UC (Bassett et
101 al. 1996, Bell et al. 1996, Vargas and Wright 2005). No plant growth regulators were used on
102 the tests in either year. Yields were assessed on 18 November 2003 and on 3 November 2004 by
103 harvesting two rows per plot using a John Deere 9910 cotton picker (Deere, Moline, IL), which
104 was adapted to pick into sacks. The sacks with cotton were weighed and seed cotton yields per
105 acre were calculated.

106 Arthropod populations were monitored in 2003 from 23 June to 18 September and in
107 2004 from 16 July to 1 September. Whiteflies populations were well below treatment levels in
108 2003, but in 2004, all the plots were treated with buprofezin (Courier at 298.1 g [AI]/ha, Nichino
109 America, Inc., Wilmington, DE) on 25 July as a prophylactic treatment; this eliminated any
110 interference that they would have caused though honeydew production.

111 **Differential population establishment.** To establish differential arthropod populations,
112 selective and disruptive pesticide treatments were applied at the onset of population build-up for
113 both cotton aphids and spider mites (Leigh 1980, Lasota and Dybas 1991, James 1997, Slosser et
114 al. 2001). Five treatments were established on 11 August 2003; Acetamiprid (Assail 70WP at

115 53.9 g [AI]/ha, Cerexagri, Inc., King of Prussia, PA) and abamectin (Zephyr 0.15EC at 0.32 g
116 [AI]/ha, Syngenta Crop Protection, Inc. Greensboro, NC) were applied to eliminate aphid and
117 spider mite populations in one treatment. Additionally, bifenthrin (Capture 2EC at 16.9 g
118 [AI]/ha, FMC Corp., Philadelphia, PA) was used to flare aphids in one treatment. Acephate
119 (Orthene 75S at 577.9 g [AI]/ha, Valent U.S.A. Corp. Walnut Creek, CA) and abamectin were
120 used to flare mite populations and eliminate aphids in another treatment. Finally, there was only
121 treatment where populations were untreated.

122 Differential populations were also established in plots on 25 July 2004. Acetamiprid
123 (Assail 70WP at 53.9 g [AI]/ha) and abamectin (Zephyr 0.15EC at 0.32 g [AI]/ha) were applied
124 to eliminate aphid and spider mite populations in one treatment. Lambda-cyhalothrin (Warrior at
125 3.8 g [AI]/ha, Syngenta Crop Protection, Inc. Greensboro, NC) was used to flare aphids in
126 another treatment. Acephate (Orthene 97 at 815.4 g [AI]/ha, Valent U.S.A. Corp. Walnut Creek,
127 CA) and abamectin were used to flare mite populations and eliminate aphids in another
128 treatment. In addition, acephate (Orthene 97 at 815.4 g [AI]/ha) was applied a second time on 5
129 August 2004 to the plots previously treated with Orthene 97, because spider mite populations
130 were slow to build. Acetamiprid and abamectin were used to eliminate both aphids and mites in
131 another treatment. Finally, there was only treatment where populations were untreated.
132 Lambda-cyhalothrin was substituted for bifenthrin in 2004 to increase aphid populations to a
133 greater degree than the previous year (Kerns and Stewart 2000, Slosser et al. 2001), although
134 aphid populations were much higher overall in 2004 than 2003.

135 **Sampling.** Ground-truthing data were collected by sampling for aphids and mites at
136 weekly intervals (approximate) within the plots. Both cotton aphids and spider mites were
137 sampled in 2003 by collecting 10-leaf samples (5th mainstem node leaf from the top of the plant)

138 per plot and counting the individuals in the laboratory under 50x magnification. Twenty-leaf
139 samples (5th mainstem node leaf from the top) were collected in 2004, making the analysis more
140 robust by increasing the power of obtaining a true measurement of population density. Aphids
141 were counted first and the leaves were soaked for 20 minutes in a ~0.5% bleach solution, with a
142 drop of liquid detergent. The leaves were then washed onto a 40 mesh, followed by a 100 mesh
143 screen, and the retained material was back-washed onto filter paper for storage and later
144 quantification of spider mite number. From the aphid and mite numbers, accumulated aphid-
145 days and accumulated mite-days were calculated for the period after treatment (Ruppel 1983).

146 **Remote sensing platforms.** 2003. Flight data, provided by Opto-Knowledge Systems,
147 Inc. (OKSI, Torrance, CA) through the USDA-ARS Western Integrated Cropping Systems
148 Research Unit, were collected on 20 August and 11 September 2003. The flight data from 20
149 August were used to correspond with a range of arthropod populations that included levels above
150 and below economic thresholds. The airplane was equipped with both a multispectral, Shafter
151 Airborne Multispectral Remote Sensing System (SAMRSS), and hyperspectral, Airborne Visible
152 Near Infrared (AVNIR), camera system. The SAMRSS system consisted of three 1024 x 1024
153 charged-coupled device cameras with filters that collected ~0.6 m resolution imagery, in four
154 bands: green (~550 nm, 10 nm bandwidth), red (~675 nm, 10 nm bandwidth), NIR (~850 nm, 40
155 nm bandwidth), and thermal (~8-12 μ m). Regions of interest (ROI) from these images were ~20
156 m² (55 pixels) for each plot. The AVNIR was a pushbroom hyperspectral system that collected
157 images in 60 bands ranging from ~430 to 1,012 nm in wavelength, at a 10 nm bandwidth, with a
158 1.6 m spatial resolution. ROI from these images were ~13 m² (40 pixels) for each plot. These
159 images were calibrated to ground-based reflectance and georectified by OKSI. The software
160 ENVI (Research Systems Inc., Boulder, CO) was used for visualization.

161 2004. An airplane was also flown on 3 August 2004 before aphid and mite population
162 levels had built in the field. This imagery was provided by InTime, Inc. (Cleveland, MS), whose
163 airplane was equipped with a proprietary multispectral camera system that was similar to the
164 SAMRSS system used in 2003. Spatial resolution was 1 m and ROI from these images ranged
165 from ~78 to ~84 m² (78-84 pixels) for each plot. The fields were uniformly dry when all the
166 aerial imagery was collected.

167 QuickBird (DigitalGlobe, Inc., Longmont, CO) satellite imagery was also acquired in
168 2004. QuickBird was equipped with a multispectral system that detected 3 bands, at a spatial
169 resolution of 2.8 m, and a panchromatic sensor with a spatial resolution of 0.6 m. The images
170 provided by DigitalGlobe, through AgriDataSensing, Inc. (Fresno, CA), were calibrated to
171 reflectance factor at the earth's surface and were collected on 25 August 2004, 30 days after
172 treatment. The fields were uniformly wet from flood irrigation the morning the images were
173 taken. The images that were received were representative of information in the red, green, blue,
174 and infrared (IR) bands; a panchromatic image was also received. Data products were provided
175 from DigitalGlobe, which included the following images: canopy closure, green vegetation index
176 (GVI) (Kauth and Thomas 1976), adjusted vegetation index, and a vegetation adjusted soil index.
177 These data products were created by DigitalGlobe by processing the data through a proprietary
178 vegetation index-type algorithm, an advanced derivative of the TSAVI model (Baret et al. 1989,
179 Baret and Guyot 1991). All images were subjected to the same analysis in ENVI as the aerial
180 data. ROI for the panchromatic images ranged from ~54 to 67 m² (170-208 pixels) for each plot,
181 while they were ~50 m² (18 pixels) for each plot in the other images. More dates were not
182 available, because of the high cost of the data collection methods and the availability of

183 acceptable conditions for data collection (i.e. windless days for flight data, positioning of the
184 satellite, and cloudless days for both types of data).

185 The flight information was supplemented with measurements on 28 July, 4, 11, 18 and 26
186 August 2004 using a backpack spectrometer in combination with a hand-held contact probe
187 [FieldSpec® Pro FR (ASD), Analytical Spectral Devices, Inc., Boulder, CO]. Measurements of
188 reflectance were made, using the spectrometer, on the adaxial surface of five random leaves in
189 each plot. Several indices and narrow bands were analyzed (Table 1). In addition, the adaxial
190 surface of the same leaves was scanned in five equidistant locations around the leaf edges using a
191 Minolta Chlorophyll SPAD 502 meter (Konica Minolta, Osaka, Japan) (chlorophyll meter); this
192 measured the relative amount of chlorophyll in the leaf using the absorbance of a wavelength in
193 the red and NIR (exact wavelength not specified by manufacturer).

194 **Statistical analyses.** The calculated indices and representative reflectance values from
195 the airplane data, were each, individually, analyzed using two-way ANOVA ($P<0.05$).
196 Likewise, every band from each image in the QuickBird satellite data was subjected to a separate
197 two-way ANOVA ($P<0.05$). A two-way ANOVA ($P<0.05$) was also performed using the yield
198 data in 2003 and 2004. Finally, the hyperspectral ground data in 2004 were subjected to repeated
199 measures ANOVA ($P<0.05$), using each day that measurements were taken, for both field trials.
200 The split-plot conservative approach was taken, for the repeated measures ANOVA, by using the
201 interaction between the block and the treatment.

202 Tukey's HSD procedure was used for mean separation in most of the above analyses.
203 However, in the event that no factors were significantly different by Tukey's HSD procedure, the
204 P values for the differences among means were used for mean separation (Fisher's protected
205 LSD procedure). These results were considered marginally significant, since Fisher's protected

206 LSD is prone to Type I errors (Carmer and Swanson 1973). Finally, orthogonal contrasts were
207 used to test the null hypothesis that arthropod-infested plots were significantly different than
208 non-arthropod infested plots, when the interaction of day of measurement*treatment was
209 significant in repeated measures ANOVA.

210 Data that violated the assumptions of ANOVA were transformed or, in the case that this
211 failed for satisfying the assumption homoscedasticity, were subjected to a weighted ANOVA
212 ($P < 0.05$). Data that violated the assumptions of normality (Shapiro-Wilk value < 0.95) and could
213 not be successfully transformed using parametric transformations were subjected to non-
214 parametric ranked ANOVA. Data presented herein are untransformed arithmetic means and
215 standard errors.

216 For each treatment, the accumulated aphid-days, accumulated mite-days and yield were
217 correlated to the index and reflectance values from the spectrometer, airborne and satellite data
218 using Spearman rank-order correlation. Slopes were only presented if the correlations had P
219 values greater than 0.05.

220 Error, or confusion matrices, are often used to establish the accuracy (sensitivity) and
221 precision (specificity) of the classification model (Congalton 1991, Richards 1993, Brewster et
222 al. 1999). Cohen's kappa (κ) was developed to solve the problem of accounting for
223 measurements that may have been correct or incorrect due to chance (Cohen 1960, Hudson and
224 Ramm 1987, Congalton 1991, Richards 1993, Varbyla 1995, Brewster et al. 1999). The κ
225 statistic is expressed from -1 to 1 (if $\kappa < 0$, the classification is worse than random, if $\kappa = 0$ the
226 classification scheme is no better than a random classification, and if $\kappa > 0$, the classification is
227 better than random). Generally a Cohen's kappa value of < 0.2 represents a slightly accurate
228 assessment, 0.21-0.40 a fairly accurate assessment, 0.41-0.60 a moderately accurate assessment,

229 0.61-0.80 a substantially accurate assessment, and 0.81-1.00 an almost perfectly accurate
230 assessment (Groote et al. 2000, Kalra et al. 2003). Cohen's kappa does not assess accuracy
231 perfectly and is a conservative assessment. The kappa coefficient tends to overestimate the
232 agreement between classification sets that is due to chance, while underestimating the overall
233 classification accuracy (Foody 1992, Lo and Yeung 2002).

234 Utility of the green band (~550 nm), NIR band (~850 nm), narrow band wavelengths at
235 ~579 and ~880 nm, and the NDVI and GNDVI were assessed using Cohen's kappa coefficient
236 (κ) (Cohen 1960), using the data from 2003. Bands from the images that were significant by
237 ANOVA ($P < 0.05$) were also assessed for their utility using Cohen's kappa coefficient (κ) in
238 2004.

239 When using Cohen's kappa, the cutoff value for classification was the median of the
240 difference between values of the treatments. For example, in the 2004 canopy image, the
241 lambda-cyhalothrin treatment had an average value of 110.68, the acetamiprid treatment an
242 average value of 58.97, the untreated plots an average value 54.53, and the acetamiprid +
243 abamectin treatment an average value of 52.27. The values that were between 84.825 and 56.75
244 were predicted to be part of the acetamiprid treatment.

245

246 Results

247 **2003.** The pesticides helped manipulate mite levels among treatments and mite
248 populations in the treatments with the highest levels continued to build until the 4th week of the
249 study. Cotton aphid populations were also manipulated by the pesticides and they increased until
250 the 3rd week of the study, without much population increase after this point (Fig. 1).
251 Nonetheless, there were a range of aphid populations among the treatments when the remote

252 sensing data were collected (August 20, 2003) (Fig. 1). These numbers represent populations
253 that were below the economic threshold levels for aphids and above the treatment threshold level
254 for mites. Treatment thresholds for spider mites are set by the percentage of leaves infested.
255 Although there were a high percentage of leaves infested, actual mite numbers were low and the
256 infestation was still in an early stage when the remote sensing data were collected.

257 None of the yields among treatments were significantly different ($F = 0.82$; $df = 4, 12$; P
258 $= 0.5352$). There was an average of 15764, 15519, 15212, 14207 and 14079 kg/ha (seed cotton)
259 in the untreated, abamectin, acetamiprid + acephate, bifenthrin, and acetamiprid + abamectin-
260 treated plots, respectively.

261 *Multispectral data.* Treatments differed significantly in the green band (~ 550 nm) ($F =$
262 6.16 ; $df = 4, 12$; $P = 0.0062$) and in the NIR band (~ 850 nm) ($F = 9.27$; $df = 4, 12$; $P = 0.0012$)
263 (Fig. 2). In the green band, the average reflectance values for the acetamiprid + abamectin and
264 bifenthrin treatment were significantly higher than the abamectin treatment. Acetamiprid +
265 acephate and untreated plots had intermediate reflectance values and did not differ significantly
266 from the other values. In contrast, in the NIR band, the bifenthrin treatment had significantly
267 higher average reflectance values than the acetamiprid + abamectin and the abamectin
268 treatments. As in the green band, acetamiprid + acephate and untreated plots had moderate
269 average reflectance and did not differ significantly from the other values.

270 Although there were significant differences among the treatments in the green and NIR
271 bands, there were no correlations for the bands or vegetation indices between accumulated mite-
272 days or accumulated aphid-days. Some of the vegetation indices and bands were correlated with
273 yield (Table 2).

274 The Cohen's kappa value was 0 for the reflectance values in the green band (~550 nm).
275 Thus, the predictive quality of the reflectance values in the green band, in this case, was no better
276 than random. The Cohen's kappa value was the 0.313 for the NIR band (~850 nm).
277 Interestingly, although no significance was shown among the treatments for the NDVI ($F = 1.30$;
278 $df = 4, 12$; $P = 0.3230$) and GNDVI ($F = 1.18$; $df = 4, 12$; $P = 0.9355$), they both had a Cohen's
279 kappa value of 0.125.

280 *Hyperspectral data.* Treatments differed significantly in one of the wavelengths in the
281 green area tested (~579 nm) ($F = 3.69$; $df = 4, 12$; $P = 0.0351$) (Fig. 2). The average reflectance
282 values at ~579 nm for the acetamiprid + abamectin treatment were significantly higher than the
283 abamectin-treated plots; values for the other treatments were intermediate.

284 There were no correlations among any of the indices or narrow bands with accumulated
285 mite-days. There were weak correlations at ~569 nm and ~579 nm with accumulated aphid-days
286 (slope = 0.37; $P = 0.037$ and slope = -0.50; $P = 0.022$). There was weak correlation with yield
287 among some of the indices and narrow bands (Table 2).

288 The Cohen's kappa value was 0.25 for the reflectance values at ~579 nm. Analogous to
289 the multispectral data, although they did not significantly differentiate any of the treatments from
290 each other, the NDVI ($F = 1.65$; $df = 4, 12$; $P = 0.2265$), the GNDVI ($F = 1.57$; $df = 4, 12$; $P =$
291 0.2460), the green peak (~550 nm) ($F = 1.98$; $df = 4, 12$; $P = 0.1618$), and the NIR band (~880
292 nm) ($F = 0.27$; $df = 4, 12$; $P = 0.8942$) all had Cohen's kappa values of 0.188.

293 **2004.** When the QuickBird data was collected, there was an excellent array of aphid
294 populations among the treatments ranging from 39 to 1510 accumulated aphid-days (Fig. 3).
295 Spider mite populations increased more slowly than the aphid populations, but slightly exceeded
296 the treatment threshold during weeks 3 through 5 of the study (Fig. 3). As in 2003, although

297 there were a high percentage of leaves infested, actual mite numbers were low and the infestation
298 was still in an early stage when the remote sensing data were collected. All the treatments
299 exceeded mite treatment thresholds by the 6th week.

300 There were no significant differences in yield among the treatments ($F = 1.23$; $df = 4, 12$;
301 $P = 0.3508$). There was an average of 9905, 8951, 8286, 7939 and 7807 kg/ha (seed cotton) in
302 the plots treated with the acetamiprid + abamectin, untreated, acephate, abamectin, and lambda-
303 cyhalothrin-treated plots, respectively.

304 *Ground (hyperspectral spectrometer) data.* No significant differences were found among
305 treatments using the spectrometer on any given date. Additionally, the chlorophyll meter
306 measurements were the only dependent variable that was significant for time (day of
307 measurements) by treatment interaction (Table 3). Plots with natural infestations of aphids and
308 spider mites had chlorophyll meter values that were significantly lower over time than
309 acetamiprid + abamectin-treated plots. The response was quadratic, rather than linear ($F = 9.76$,
310 $P = 0.0028$). Moreover, all the narrow band and index values tested changed significantly over
311 time (day of measurements) (Table 3). There were no correlations among accumulated aphid-
312 days, mite-days calculated up to the 28 July, with yield, which was collected 3 November.

313 On 11 August, although there were no significant differences among treatments for any
314 of the variables tested by ANOVA, the DVI and NIR (~880 nm) were negatively linearly
315 correlated with accumulated aphid-days (Table 4). There were no correlations among any of the
316 indices or narrow bands with yield.

317 Like the previous week, on 18 August there were no correlations with accumulated mite-
318 days or yield. Furthermore, although there were no significant differences among treatments by
319 ANOVA, the narrow band at in the NIR wavelength (~880 nm) was negatively linearly

320 correlated with accumulated aphid-days (Table 4). There were no correlations among any of the
321 indices or narrow bands with yield.

322 On 26 August, there were correlations among various vegetation indices and narrow
323 bands vs. accumulated aphid-days (Table 4). Finally, there were no correlations among any of
324 the indices or narrow bands with yield.

325 *Airborne (multispectral) data.* Using the airborne data provided by InTime, none of
326 treatments were significantly different in any of the narrow bands or indices tested.

327 Additionally, there were no correlations among accumulated mite, accumulated aphid-days, or
328 yield with any of the bands or indices tested.

329 *Satellite (multispectral) data.* The lambda-cyhalothrin treatment was significantly
330 different from the other treatments in the canopy image, three of the four color infrared images,
331 the false color image, and one of the vegetation images (Table 5). None of the treatments were
332 significantly different using the panchromatic image, one of the color infrared images, the
333 soil/vegetation image, or the GVI image. Additionally, although treatments in the false color
334 image and one the vegetation images were significant by ANOVA, the Tukey groupings failed to
335 differentiate among the treatments (Fig. 4); accordingly, the letter groupings in the preceding
336 figure represents the *P* values for Fisher's LSD procedure.

337 Three images were significantly linearly correlated with accumulated mite-days, while
338 two images were significantly linearly correlated with accumulated aphid-days (Table 5). None
339 of the images were significantly correlated with yield. Furthermore, the Cohen's kappa values
340 for the significant image bands were 0.19 for the canopy image, 0.5 for the color infrared image
341 1, 0.44 for the color infrared image 2, 0.13 for the color infrared image 2, 0 for the false color
342 image and 0.06 for the vegetation index 1 image. The Cohen's kappa values for the treatments

343 that showed no significant differences were 0.25 for the panchromatic image and 0.46 for the
344 GVI image.

345

346

Discussion

347 In 2003, aphids were well below threshold levels throughout the study and, although
348 spider mite levels had exceeded threshold levels by August 20, the population levels were still
349 low. A similar phenomenon was observed in 2004, with spider mite levels above threshold
350 levels, but with low numbers. Albeit aphid levels were higher in 2004, the accumulated mite and
351 aphid-days were low enough that the arthropods had no significant on the yield in both years.

352 Although, the growth stage of the cotton in these experiments was consistent with the
353 normal occurrence of these pests in SJV cotton, these exploratory experiments were performed in
354 a manipulated controlled environment. Intrafield variation that is detected with remote sensing
355 methods alone is not adequate for prescribing a treatment regime against spider mite and aphid
356 pests. For example, an area of cotton that has few naturally occurring arthropod pests will have
357 variation that is attributable to factors such as plant growth stage, irrigation status, soil condition,
358 nutrient status and injury from other pest classes (e.g. weeds, nematodes, fungi). However,
359 arthropod-infested cotton was successfully detected over a two year period in this study. In
360 addition, aphids and spider mites were detected in fields that had dry soil in 2003 and in fields
361 that had wet soil in 2004. Finally, the field characteristics and the amount of nitrogen applied to
362 the plants were different both years, but both spider mites and aphids were still detected. Even
363 though arthropod-damaged cotton was detected over a range of conditions, each field must be
364 ground-truthed to demonstrate that an arthropod infestation is present and at what level it is
365 present.

366 In 2003, cotton with mite and aphid infestations in the early stages could be detected in
367 the green peak areas, using airborne imagery. Treatments with spider mites were detected using
368 bands in the green peak in both multispectral and hyperspectral data. However, spider mites
369 were only detected in NIR bands using multispectral (SAMRSS) data. Because treatment
370 decisions are based on percent infested leaves, mite numbers were above the treatment threshold
371 levels (30-50% infested mite leaves), but were only present in low numbers (2 to 31 average
372 mites/leaf). Thus, the infestations were detected at a stage that was early enough for a grower to
373 make a treatment decision. The presence of aphids lowered the reflectance values in the
374 abamectin-treated plots, while the presence of mites and aphids lowered the reflectance values in
375 the acetamiprid + abamectin treated plots.

376 Using the multispectral data from 2003, it is possible that plots with high numbers of
377 aphids, but low numbers of mites, and plots with low number of aphids, but low numbers of
378 mites (i.e. the untreated and acetamiprid + acephate-treated plots), were statistically similar
379 because the accumulated-mite and aphid-days offset one another. Moreover, aphid damage was
380 clearly detected in the green band (acetamiprid + abamectin plot values were significantly higher
381 than abamectin-treated plots), although untreated plots had accumulated mite and aphid-days that
382 were similar to abamectin-treated plots. Finally, the Cohen's kappa values calculated indicate
383 that the NDVI, GNDVI and the NIR band were better predictors of aphid and spider mite
384 infestations than the green band, with the NIR band being the best predictor of aphid and spider
385 mite infestations that was tested.

386 Using the hyperspectral data from 2003, it is conceivable that the cotton with few
387 accumulated aphid and mite-days (i.e. acetamiprid + abamectin-treated plots) had a higher
388 reflectance value at ~579 nm in the plots because it had less aphid and mite pressure. Also, plots

389 with high accumulated aphid-days, but few accumulated mite-days, had lower average
390 reflectance values at ~579 nm. This is similar to the findings in the green band using the
391 multispectral data.

392 The narrow band at ~579 nm had a Cohen's kappa value of 0.25. Thus, the predictive
393 quality of the reflectance values at ~579 nm, in this case, was better than random, but not
394 extremely accurate. Nonetheless, the Cohen's kappa values calculated indicate that the narrow
395 band ~579 nm (in the green wavelengths) was the best predictor of spider mites and aphid-
396 infested cotton, using the hyperspectral imagery. The NNIR, NR, OSAVI, MCARI, GDVI,
397 RVI₁, HM, YI and narrow bands at ~569 nm ~802 nm were not useful to detect aphid or mite-
398 infested cotton in both the 2003 multispectral and hyperspectral images.

399 Aphids were detected with the ground-based spectrometer measurements before they
400 reached economically damaging levels in the mite and aphid experiment in 2004. Again,
401 wavelengths in the NIR were correlated with aphid pressure and there was a relationship among
402 decreasing reflectance values and increasing accumulated aphid-days (Table 5). Based on these
403 correlations, the NIR wavelengths were the first to significantly distinguish aphid-infested plots
404 well below economic threshold levels. Treatments with higher aphid and mite pressure had
405 significantly lower chlorophyll levels over time, using the chlorophyll meter data. Finally,
406 aphid-infested cotton was detected using the satellite imagery from 2004, although the aphid
407 infested cotton had received a high amount of aphid pressure by the time the imagery was
408 collected. Because these images were correlated with accumulated aphid-days and not
409 accumulated mite-days, we used the images to differentiate the lambda-cyhalothrin treatment
410 based on aphid damage, rather than mite damage. Although the acetamiprid + acephate and the

411 untreated plots had more accumulated aphid and mite-days than the abamectin treatment (Fig. 3),
412 they were not significantly different from each other in any of the images received.

413 Spider mite damaged cotton was not detected with the ground-based spectrometer
414 measurements in 2004, although the chlorophyll meter detected lower chlorophyll levels in plots
415 with higher aphid and mite pressure. Spider mite infested cotton could be detected with airborne
416 imagery in 2003, and accumulated mite-days were linearly correlated with decreasing
417 reflectance, using airborne images in 2004. Mite damage above treatment threshold levels, but
418 with low mite numbers, could be detected in 2003 and 2004. Three images had values that were
419 linearly correlated with accumulated mite-days, using the QuickBird airborne data in 2004.

420 Airborne methods are different than ground based methods because they are subject to
421 sources of variation such as anisotropy, variable illumination, and water vapor, and aerosol
422 particles in the air. Many of the spectrometer measurements were made in conditions $>38^{\circ}\text{C}$,
423 although the ASD FieldSpec® Pro is designed to function in temperatures $<38^{\circ}\text{C}$. Airborne
424 methods way have been able to detect canopy differences in spider mite damaged cotton that the
425 spectrometer was not able to detect, because the contact probe used with the spectrometer
426 samples a small portion of the leaf and only 5 leaves per plot were sampled. Spider mite
427 damaged cotton may more seriously affect the canopy before affecting the reflectance of the
428 leaves, perhaps by eliciting a host response in the cotton, such as leaf curl or drooping. With
429 remote sensing technology, this could only be detected with canopy methods.

430 Unfortunately, in the mite and aphid test in 2004, there were no aerial or satellite data in
431 the time period between 8 days after treatment, when there were few aphids, and 30 days after
432 treatment, when the aphids had reached economic threshold levels. Aphid damaged cotton was
433 correlated with accumulated aphid-days using the spectrometer ground-based measurements in

434 the mite and aphid test. Nevertheless, it would be beneficial to have aerial or satellite data on an
435 intermediate number of aphids to determine if they could be detected with such data when the
436 plants were stressed, before yield loss occurred, and when the economic threshold was reached.
437 Aphid-infested cotton was detected using satellite data at levels well below the economic
438 thresholds in a separate test (unpublished data).

439 Some residue from the lambda-cyhalothrin treatment was visible on the leaves until
440 approximately 2 weeks after treatment in 2004. Also, some insecticides, such as acephate, can
441 change cotton plant physiology to indirectly facilitate mite infestations (Leigh 1980, Maggi and
442 Leigh, 1983, Beasley et al. 1996) and it is feasible that this physiological change could be
443 detected using remote sensing. However, the aerial remote sensing data taken 9 days after the
444 pesticide applications showed no significant differences among treatments. Additionally, the
445 spectrometer data were not significant or correlated with accumulated aphid-days until the aphid
446 pressure was high. Consequently, it is unlikely that the differences in reflectance were due to the
447 direct or indirect effects of the chemicals on the cotton. Rather, wavelengths in the NIR were the
448 most robust, in comparison with other indices and images tested in 2003 and 2004, when used to
449 distinguish aphid or mite damaged cotton from other cotton.

450 In 2003, the presence of low aphid numbers decreased reflectance values when compared
451 with uninfested cotton. Additionally, aphids with higher population levels also decreased
452 reflectance values in 2004. Both spider mite and aphid-infested cotton leaves decreased in the
453 NIR reflectance values over time. However, the NIR wavelengths alone cannot be used to
454 distinguish infested cotton. For example, the decrease in the values of the NIR wavelength is
455 consistent with a general decline in canopy cover (more bare soil exposed and more shadows) or
456 a senescing plant, because healthier plants reflect more light in the NIR (Richardson et al. 1975,

457 Summy et al. 1997, Brewster et al. 1999). Because both spider mite and aphids were detected by
458 a decrease in NIR wavelength values, monitoring the field populations is crucial to parse their
459 presence from other factors in the field that may cause similar damage. Once an infestation of
460 aphids or mites is discovered, the extent of the infestation can be monitored by comparison of
461 high aphid and/or spider mite populations in cotton to cotton with low aphid and/or spider mite
462 infestation levels; the highly-infested cotton should have a lower NIR wavelength value than the
463 cotton with low aphid populations. Cohen's kappa values from multispectral data in 2003 and
464 2004 indicate that fairly (0.21-0.40) to moderately accurate (0.41-0.60) predictions (Groote et al.
465 2000, Kalra et al. 2003) can be made using the NIR wavelengths for predicting areas of cotton
466 that are infested with a range of cotton aphid and spider mite populations.

467 Ground-truthing and remote sensing methods must be amalgamated for remote sensing to
468 be useful for the accurate identification of spider mite and cotton aphid infestations. The ground-
469 based information must confirm the dominant arthropod problem that is present. The remote
470 sensing imagery can assist, using supervised classification, to prescribe and appropriate
471 management strategy.

472 Because spider mites and aphids occur in heterogeneous areas of the fields, it is possible
473 that these "hot spots" can be differentiated from other sources of variation, using the
474 wavelengths in the NIR. The sensors that we tested had resolutions ranging from 0.6 - 2.8 m,
475 with ROI from 13-84 m². Hence, we have shown that remote sensing can detect spider mite and
476 aphid infestations in small areas. It must be demonstrated that this is possible outside of a
477 controlled experiment and that the costs associated with ground-truthing, processing of remote
478 sensing images, and variable rate pesticide applications are less than traditional sampling
479 methods and traditional pesticide applications. Thus, the results presented herein are exploratory

480 and represent one step in the process of integrating remote sensing techniques into the IPM
481 management practices of SJV cotton growers.

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Table 1.

Vegetation Index		Equation	Reference
Normalized Difference Vegetation Index (NDVI)	NDVI =	$\frac{(R_{NIR} - R_{RED})}{(R_{NIR} + R_{RED})}$	(Rouse et al. 1974)
Normalized Near Infrared (NNIR)	NNIR =	$\frac{(R_{NIR})}{(R_{NIR} + R_{RED} + R_{GREEN})}$	(Ojala unpublished)
Normalized Red (NR)	NR =	$\frac{(R_{RED})}{(R_{NIR} + R_{RED} + R_{GREEN})}$	(Ojala unpublished)
Optimized Soil Adjusted Index (modified) (OSAVI)	OSAVI =	$\frac{(R_{NIR} - R_{RED})}{(R_{NIR} + R_{RED} + 0.16)}$	(Rondeaux et al. 1998, Ojala unpublished)
Modified Chlorophyll Absorption in Reflectance Index (MCARI)	MCARI =	$[(R_{770} - R_{670}) - 0.2 * (R_{770} - R_{550})] * (R_{770} - R_{670})$	(Daughtrey et al. 2000)
Green Differential Vegetation Index (GDVI)	GDVI =	$R_{GREEN} - R_{RED}$	(Ojala unpublished)
Ratio Vegetation Index (RVI ₁) [or Simple Ratio (SR ₁)]	RVI or SR =	$\frac{(R_{NIR})}{(R_{RED})}$	(Jordan 1969)
Ratio Vegetation Index (RVI ₂) [or Simple Ratio (SR ₂)] ^x	RVI or SR =	$\frac{(R_{RED})}{(R_{GREEN})}$	(Jordan 1969)
Green Vegetation Index (GNDVI) ^y	GNDVI =	$\frac{(R_{NIR} - R_{GREEN})}{(R_{NIR} + R_{GREEN})}$	(Gitelson and Merzlyak 1998)
Difference Vegetation Index (DVI) ^x	DVI =	$(R_{NIR} - R_{RED})$	(Tucker 1979)
Visible Atmospherically Resistant Index (VARI _{GREEN}) ^x	VARI _{GREEN} =	$\frac{(R_{GREEN} - R_{RED})}{(R_{GREEN} + R_{RED} - R_{450})}$	(Gitelson et al. 2002)
Half Max (HM) ^z	HM =	$R_{675} + \frac{(R_{760} - R_{675})}{2}$	(Ojala unpublished)
Photochemical Reflectance Index (PRI) ^z	PRI =	$\frac{(R_{520} - R_{580})}{(R_{520} + R_{580})}$	(Gamon et al. 1992)
Yellowness Index (YI) ^z	YI ≈	$\frac{(R(\lambda_i) - 2R(\lambda_0) + R(\lambda_{+1}))}{\Delta\lambda^2}$	(Adams et al. 1999)
Narrow Bands	~550 nm ^y =	~R ₅₅₀	
	~553 nm =	~R ₅₅₃	
	~569 nm ^z =	R ₅₆₉	
	~579 nm ^y =	R ₅₇₉	
	~802 nm ^z =	R ₈₀₀	
	~850 nm ^x =	R ₈₅₀	
	~880 nm =	~R ₈₈₀	

x: Only tested these in 2004, hyperspectral spectrometer data.

y: Only tested these in 2003, hyperspectral data and 2004 hyperspectral spectrometer data.

z: Only tested these in 2003, hyperspectral data.

Table 2. Spearman correlations for the indices and narrow bands versus yield, 2003.

	Index or Band	Slope	<i>P</i>
Multispectral	NDVI	0.73	0.0002***
	NNIR	0.72	0.0004***
	RVI ₁	0.49	0.0287*
	GNDVI	0.82	<0.0001***
	RVI ₁	0.49	0.0287*
Hyperspectral	NNIR	0.52	0.0197*
	NR	-0.49	0.0265*
	OSAVI	0.49	0.0298
	RVI ₁	0.54	0.0141*
	GNDVI	0.52	0.0188*
	~550 nm	-0.75	0.0001***
	~569 nm	-0.78	<0.0001***
~579 nm	-0.76	0.0001***	

* $P < 0.05$.

** $P < 0.01$.

*** $P < 0.001$.

Table 3. Results from repeated measures ANOVA. Day of measurement represents the weekly hyperspectral spectrometer measurements on 28 July, 4, 11, 18, and 26 August 2004. chloro.

meter = chlorophyll meter.

Index or Band	Treatment ^x		Day measured ^y		Day measured*Treatment ^z	
	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>	<i>F</i>	<i>P</i>
NDVI	2.23	0.1273	101.73	<0.0001***	1.14	0.3406
NNIR	1.26	0.3368	66.86	<0.0001***	1.07	0.4010
NR	2.32	0.1158	124.05	<0.0001***	1.00	0.4706
OSAVI	0.86	0.5147	81.31	<0.0001***	1.40	0.1731
MCARI	0.45	0.9623	25.00	<0.0001***	0.77	0.6823
GDVI	0.63	0.6490	59.75	<0.0001***	1.59	0.0989
RVI ₁	1.81	0.1926	109.25	<0.0001***	0.91	0.5616
RVI ₂	0.49	0.7437	286.05	<0.0001***	1.06	0.4146
GNDVI	0.47	0.7601	115.64	<0.0001***	1.16	0.3239
DVI	0.30	0.8725	58.53	<0.0001***	0.86	0.6148
VARI _{GREEN}	2.39	0.1092	10129.2	<0.0001***	1.27	0.2472
~553 nm	0.27	0.8924	10.07	<0.0001***	1.30	0.2284
~579 nm	0.41	0.7894	31.44	<0.0001***	0.70	0.7860
~880 nm	1.00	0.4466	4.03	0.0059**	1.40	0.1731
chloro. meter	1.33	0.3159	7.57	<0.0001***	1.99	0.0310*

x: Corrected df = 4,12.

y: Corrected df = 4,59.

z: Corrected df = 16,59.

* $P < 0.05$.

** $P < 0.01$.

*** $P < 0.001$.

Table 4. Spearman correlations for hyperspectral spectrometer indices and narrow bands vs. accumulated aphid-days.

Date	Index or Band	Slope	<i>P</i>
11 August 2004	~880 nm	-0.57	0.0082**
18 August 2004	GDVI	-0.48	0.0307*
	RV _{I2}	0.49	0.0276*
	VARI _{GREEN}	-0.52	0.0180*
	~880 nm	-0.46	0.0422*
26 August 2004	NR	0.64	0.0025**
	OSAVI	-0.63	0.0026**
	GDVI	-0.69	0.0008***
	RV _{I1}	-0.61	0.0041**
	RV _{I2}	0.68	0.0010**
	DVI	-0.74	0.0002***
	VARI _{GREEN}	-0.58	0.0072**
	~553 nm	-0.49	0.0279*
~880 nm	-0.62	0.0035**	

* $P < 0.05$.

** $P < 0.01$.

*** $P < 0.001$.

Table 5. Images provided by DigitalGlobe, Inc. that were able to significantly distinguish aphid infested cotton (lambda-cyhalothrin treatment) from other cotton. 25 August 2004- 31 days after treatment.

Index or Band	Lambda-cyhalothrin treated plots vs. other plots ^z		Images correlated with accumulated mite-days		Images correlated with accumulated aphid-days	
	<i>F</i>	<i>P</i>	Slope	<i>P</i>	Slope	<i>P</i>
Canopy	7.4	0.0030**	0.59	0.0059**	0.46	0.0396*
Color Infrared 1	6.72	0.0045**	-0.40	0.0797	-0.75	0.0001***
Color Infrared 2	6.61	0.0048**	-0.38	0.0949	-0.77	<0.0001***
Color Infrared 3	6.71	0.0045**	-0.39	0.0870	-0.79	<0.0001***
False Color	3.41	0.0439*	0.67	0.0011**	0.28	0.2399
Veg. Index 1	3.38	0.0452*	0.68	0.0010**	0.27	0.2521

z: Corrected df = 4,12.

* $P < 0.05$.

** $P < 0.01$.

*** $P < 0.001$.

Figure Legends

Fig. 1. Average accumulated aphid and mite-days calculated after treatment, over time, 2003.

Arrow denotes the date that airborne data were collected. ad: acetamiprid; an: abamectin; u: untreated; ae: acephate; bn: bifenthrin

Fig. 2. Average reflectance values of multispectral data (~550 and ~850 nm) and hyperspectral

data (~579 nm), 2003. Letters represent mean separation by Tukey's HSD procedure. ad: acetamiprid; an: abamectin; bn: bifenthrin; ae: acephate; u:untreated

Fig. 3. Average accumulated aphid and mite-days calculated after treatment, over time, 2004.

Arrow denotes the date that airborne data were collected. ad: acetamiprid; an: abamectin; u: untreated; ae: acephate; ln: lambda-cyhalothrin

Fig. 4. Average QuickBird (satellite) false color image and vegetation image values (+SE)

among treatments. Letter groupings represent *P* values obtained from Fisher's LSD procedure.

Groupings represent marginal significance. ad: acetamiprid; an: abamectin; ln: lambda-cyhalothrin; ae: acephate; u: untreated

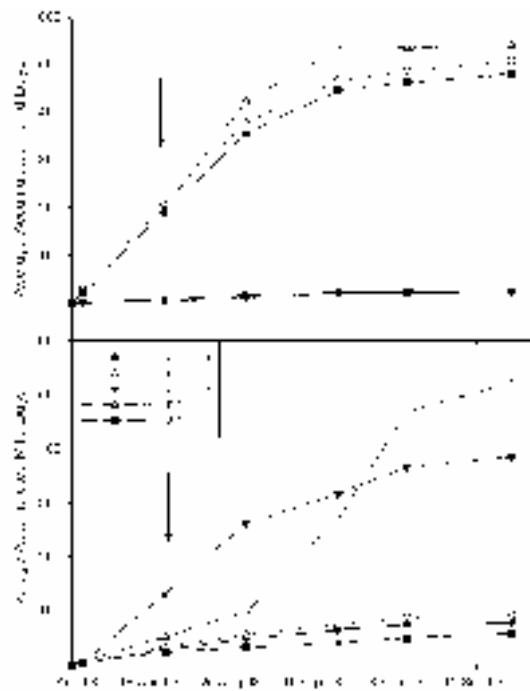


Figure 1; Reisig and Godfrey.

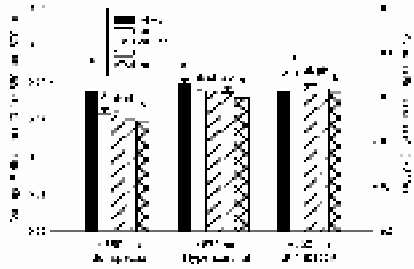


Figure 2; Reisig and Godfrey.

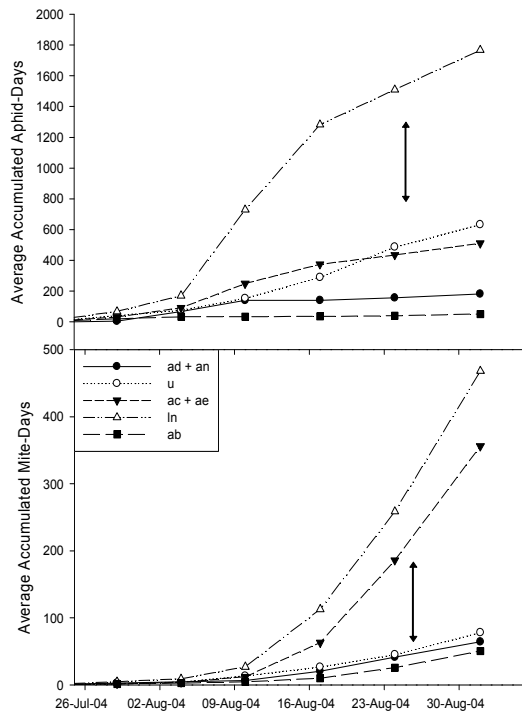


Figure 3; Reising and Godfrey.

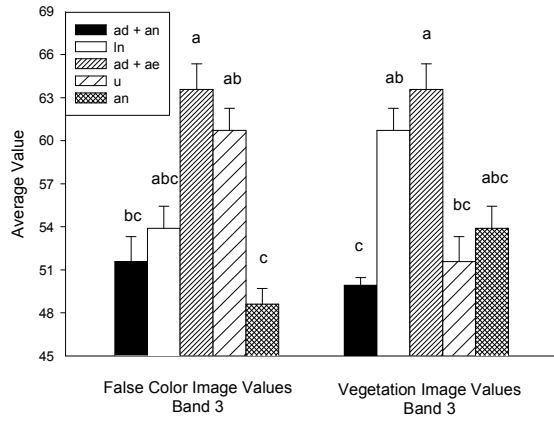


Figure 4; Reisig and Godfrey.