

1 **Genetic Mapping and QTL Analysis for Disease Resistance Using F₂ and**
2 **F₅ Generation-based Genetic Maps Derived from Tifrunner × GT-C20 in**
3 **Peanut (*Arachis hypogaea* L.)**

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18
19 Abbreviations: QTL, quantitative trait loci; PVE, phenotypic variation explained; TSWV,
20 Tomato spotted wilt virus; LS, leaf spots.

21 Running Head: QTL analysis of disease resistance in peanut

22 Key words: Genetic map, Quantitative trait loci, thrips, leaf spots, Tomato spotted wilt virus

23

24 **Abstract**

25 One mapping population derived from Tifrunner \times GT-C20 has shown great potential in
26 developing high dense genetic map and identification of QTLs for important disease
27 resistance, Tomato spotted wilt virus (TSWV) and leaf spot (LS). Both F₂ and F₅ generation-
28 based genetic maps were constructed previously with 318 and 239 marker loci, respectively.
29 Higher map density could be achieved with the F₂ map (5.3 cM/locus) as compared to the F₅
30 (5.7 cM/locus). QTL analysis using multi-environment phenotyping data from F₈ and higher
31 generation for disease resistance identified 54 QTLs in the F₂ map including two QTLs for
32 thrips (12.14 – 19.43% PVE), 15 for TSWV (4.40–34.92% PVE) and 37 for LS (6.61–
33 27.35% PVE). Twenty-three QTLs could be identified in F₅ map including one QTL for
34 thrips (5.86% PVE), nine for TSWV (5.20 – 14.14% PVE) and 13 for LS (5.95–21.45%
35 PVE). Consistent QTLs identified in each map have shown higher phenotypic variance than
36 non-consistent QTLs. As expected, the number of QTLs and their estimates of phenotypic
37 variance were lower in the F₅ map. This is the first QTL study reporting novel QTLs for
38 thrips, TSWV and LS in peanut, and thus, future studies will be conducted to refine these
39 QTLs.

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44 **Introduction**

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46 Peanut has its global presence among growers and consumers with a total production of 37.7

47 Mton from 24.1 Mha in 2010 (FAO, 2012). The average yield was 1564 Kg/ha, and a wide

48 gap exists between the genetic potential of the modern cultivars and their actual yield in the

49 farmer's field. This gap has been heavily widened by several biotic and abiotic stress factors

50 in the past and it may be even worse at the current scenario due to the fluctuating climatic and

51 environmental conditions. Among the biotic stresses, early leaf spot (ELS) (caused by

52 *Cercospora arachidicola*), late leaf spot (LLS) (caused by *Cercosporidium personatum*) and

53 Tomato spotted wilt virus (TSWV) may cause significant yield loss (Nigam et al., 2012).

54 TSWV is generally spread by thrips and the farmers try to control TSWV indirectly with

55 insecticide spray. In 1997 and 1998, losses due to TSWV peanut was estimated ~ USD 40

56 million/year for Georgia alone in USA

57 (<http://www.caes.uga.edu/topics/diseases/tswv/peanut/intro.html>). Despite several chemical

58 treatments are available to control these diseases, host-plant resistance is the best control

59 mechanism which has the advantage of being cost-effective and eco-friendly.

60 Conventional breeding has been the major force in providing modern cultivars to the

61 farmers. Integration of genomics tools with conventional breeding has been successful in

62 some of the crops but peanut lagged behind in terms of genetic and genomic resources

63 required for such approach. However, the development in genetic and genomic resources in

64 peanut in recent years has provided the possibility for improving peanut through marker-

65 assisted selection to lead to the more rapid development of superior cultivars using

66 informative markers linked to desired traits. Although, marker-assisted breeding has been

67 applied on a limited scale (see Pandey et al., 2012), still peanut lacks availability of linked

68 markers for important traits. Already marker-assisted breeding in peanut has successfully

69 demonstrated its utility by using available limited resources in conversion of peanut cultivar
70 ‘Tifguard’ (Holbrook et al., 2008) into ‘high oleic Tifguard’ in 26 months (Chu et al., 2011).

71 Identification of linked markers is the base to improve peanut resistance for the important
72 diseases through marker-assisted breeding, and a mapping population derived from the cross
73 ‘Tifrunner’ × ‘GT-C20’ was developed for identification of linked markers. The parental
74 genotypes have several contrasting traits such as Tifrunner with high level of resistance to
75 TSWV, moderate resistance to early and late leaf spot (Holbrook and Culbreath, 2007) while
76 GT-C20 is susceptible to these diseases but has resistance to aflatoxin contamination (Liang
77 et al., 2005). Parental screening with ~5000 SSRs resulted in identification of 385
78 polymorphic loci which were genotyped on a set of 94 individuals of F₂ population. As a
79 result, a genetic linkage map was constructed with 318 mapped loci distributed on 21 linkage
80 groups with genome coverage of 1,674.4 cM and a marker density of 5.3 cM /locus (Wang et
81 al., 2012). Meanwhile, this population was advanced to F₅ generation and used for
82 development of another genetic map with 239 loci distributed on 26 linkage groups covering
83 a total genome distance of 1,213.4 cM and average map density of 5.7 cM/locus (Qin et al.,
84 2012). This population was then extensively phenotyped during the years for several
85 important traits including three important diseases.

86 Thus, this study reports the use of genotyping data generated at F₂ and F₅ generation and
87 phenotyping data generated at higher generations for identification of quantitative trait loci
88 (QTL) for thrips, TSWV and leaf spots including early and late leaf spots in this study. The
89 field phenotyping trials were conducted in multiple fields from 2010 to 2012. Late leaf spot
90 was predominate pathogen in all three years. Also, comparison was made for the effects of
91 identified QTLs and common genomic regions identified in the F₂ and F₅ maps.

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96 **MATERIALS AND METHODS**

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98 **Mapping Population**

99 A mapping population derived from the cross ‘Tifrunner’ × ‘GT-C20’ (referred to as T
100 population) was developed through single seed descent method (SSD) at Crop Protection and
101 Management Research Unit of USDA-ARS, Tifton, USA. The female parent, Tifrunner, is a
102 runner market-type cultivar with high level of resistance to TSWV, moderate resistance to
103 ELS and LLS, and has late maturity (Holbrook and Culbreath, 2007). The male parent, GT-
104 C20, is a Spanish-type breeding line with high susceptibility to TSWV and leaf spots but has
105 resistance to aflatoxin contamination (Liang et al., 2005). As of now, this mapping population
106 consists of 248 RILs (recombinant inbred lines) and has been phenotyped for several
107 agronomic traits including disease resistance.

108 **DNA Isolation, Polymorphism and Genotyping**

109 Initially the total genomic DNA was extracted from young leaflets of 94 F₂ plants along with
110 the parental genotypes (Tifrunner and GT-C20). Parental polymorphism screening and
111 population genotyping were conducted with SSRs markers available at UC-Davis and
112 Tuskegee University, USA. Simultaneously, the generation advancement was done from F₂
113 to F₅ generation and again the DNA was isolated from a subset of 158 F₅ individuals in order
114 to construct genetic map and use multiseason phenotyping data for QTL analysis. The details
115 of PCR reactions and complete genotyping and map construction were early published for F₂
116 map (Wang et al., 2012) and F₅ map (Qin et al., 2012).

117 **Phenotyping for Disease Resistance**

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119 The entire set of RILs with 248 individuals were phenotyped for several important traits
120 including resistance to thrips, Tomato spotted wilt virus (TSWV) and leaf spots (LS)
121 including both early and late leaf spots but late leaf spot was the predominate disease in all
122 three years. Therefore, the general term of leaf spots was used in this study. The field trials
123 were conducted using randomized complete block designs with at least 3 replications in 2010
124 at Dawson and Tifton, Georgia, and 2011 and 2012 at Tifton, Georgia. Late leaf spot was
125 predominate pathogen in all three years.

126 In Tifton, Georgia, two separate field trials were conducted at the Belflower Farm in
127 all three years. Soil type is Tifton loamy sand (Fine-loamy, siliceous, thermic Plinthic
128 Kandiudult). In each year, one experiment was planted in April to maximize potential for
129 development of spotted wilt epidemics (Li et al., 2012) and one was planted in May to reduce
130 potential for spotted wilt epidemics and increase the likelihood of heavy leaf spot epidemics.
131 Experiment plots were 6.0 m long, separated by 2.4 m alleys. Peanut seeds were planted in
132 91-cm-spaced twin-row plots.

133 Severity of TSWV was assessed using a 0 to 5 severity scale adapted from Baldessari
134 (2008) based on visual determination of presence of symptoms and estimation of the degree
135 of stunting (reduction in plant height, width, or both) for symptomatic plants. Leaf spot
136 severity was evaluated using the Florida 1 to 10 scale (Chiteka et al., 1988) where 1 = no leaf
137 spot; 2 = very few lesions on the leaves and none on upper canopy; 3 = very few lesions on
138 upper canopy; 4 = some lesions with more on upper canopy with 5% defoliation; 5 =
139 noticeable lesions on upper canopy with 20% defoliation; 6 = numerous lesions on upper
140 canopy with 50% significant defoliation; 7 = numerous lesions on upper canopy with 75%
141 defoliation; 8 = Upper canopy covered with lesions with 90% defoliation; 9 = very few leaves
142 covered with lesions remain and some plants completely defoliated; 10 = plants dead.

143 This population was phenotyped for thrips for one season (TPS_DW10) at Dawson in

144 2010 while TSWV for four seasons i.e., at Dawson in 2010 (TSWV_DW10E1), at two
145 locations of Tifton in 2010 (TSWV_TF10E2 and TSWV_TF10E3) and at Tifton in 2011
146 (TSWV_TF11E4). This population was screened for leaf spot for a total of 10 seasons which
147 include screening at Dawson in 2010 (LS_DW10E1 and LS_DW10E2), at Tifton in 2010
148 (LS_TF10E3) and in 2011 (LS_TF11E4, LS_TF11E5, LS_TF11E6, and LS_TF11E7) while
149 three in 2012 (LS_TF12E8, LS_TF12E9 and LS_TF12E10).

150 **Nomenclature Uniformity between Genetic Maps**

151 The genetic maps were constructed at two different institutions using the two different
152 generations of the same cross i.e., using 94 F₂ individuals at Tuskegee University and 158 F₅
153 individuals at USDA-ARS (Tifton), and published in the same year i.e., late 2012 (Wang et
154 al., 2012) and early 2012 (Qin et al., 2012), respectively. The panel of markers screened on
155 parental genotypes was different, hence, some differences in nomenclature used for names of
156 markers were found. This was more frequent with the naming of unpublished markers having
157 long IDs e.g., the markers developed through BAC-end sequencing were named with prefix
158 “ARS” in the F₅ map while with prefix “GNB” in the F₂ map. Here we retained the names as
159 such for all the published markers used in these two maps while few changes were made to
160 keep size of names manageable and better viewing such as ‘pPGP....’ and ‘sPGP....’ were
161 abbreviated to ‘seq.....’ in order to bring uniformity with recently published high dense
162 consensus genetic maps (Gautami et al., 2012; Shirasawa et al., 2013). The purpose of all the
163 above exercise was to bring the genetic information in uniformity which has helped in
164 comparison of genetic maps between each other and also with published consensus genetic
165 map. It is important to mention that the genetic map information generated using F₅
166 population (Qin et al., 2012) was used for construction of both the consensus genetic maps
167 (Gautami et al., 2012; Shirasawa et al., 2013) while F₂ genetic map could not be completed
168 due to delay in screening large number of markers and genotyping.

169 **Reproducing Genetic Maps and QTL Analysis**

170 The method of genetic map construction for both maps was given in detail by Qin et al.
171 (2012) and Wang et al. (2012). Here we made the nomenclature of both the genetic maps
172 uniform in consensus with the published consensus genetic maps (Gautami et al., 2012;
173 Shirasawa et al., 2013) where distinct linkage groups have been assigned to particular
174 genomes. MapChart 2.2 (Voorrips, 2002) was used for reproducing both the genetic maps
175 using uniform nomenclature with the genetic map information. The genotyping information
176 generated on both the generations (F_2 and F_5) was also used here for conducting QTL
177 analysis using software WinQTL Cartographer, version 2.5 (Wang et al., 2007). The
178 composite interval mapping (CIM) approach, which is based on a mixed linear model, was
179 used for detection of QTLs with LD more than 2.5. Parameters such as model 6, scanning
180 intervals of 1.0 cM between markers and putative QTLs with a window size of 10.0 cM were
181 used for conducting the CIM analysis.

182

183 **RESULTS**

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185 **Comparison of Both Genetic Maps with Reference Consensus Genetic Map**

186 Upon the comparison of the corresponding linkage groups (LGs) between these two (F_2 and
187 F_5) maps, 19 LGs of the F_2 map were found identical to 20 LGs of the F_5 map (Supplemental
188 Table S1). Of the total 22 LGs of the F_2 map and 26 LGs of the F_5 map, three LGs (AhIII,
189 AhXXI and AhXXII) and six LGs (LGT1, LGT12, LGT19, LGT22, LGT23 and LGT26)
190 could not correspond to each other due to less number of mapped loci as well as lack of
191 common loci, respectively. Two LGs of the F_5 map (LGT15 and LGT25) shared common
192 loci with one LG (AhVIII) of the F_2 map. Upon comparing these two genetic maps with
193 reference consensus genetic maps using the common marker loci, a total of 9 of the 10 LGs

194 from A-genome and 8 of the 10 LGs from B-genome could be assigned. In general the co-
195 linearity has been observed for these two maps with each other and also with the reference
196 consensus genetic map (Figure 1).

197 **QTL Analysis for Biotic Stresses**

198 The entire RILs with 248 individuals were phenotyped for thrips, TSWV and leaf spots (LS)
199 in multiple fields and planting dates from 2010 to 2012 in Georgia. Late leaf spot was
200 predominate pathogen in all three years. Therefore, the general term of leaf spots (LS) was
201 used in this study, including both early and late leaf spots. These phenotyping data were used
202 in combination with genotyping data based on F₂ and F₅ generation for identification of
203 QTLs associated with each trait. A total of 77 QTLs could be detected for these three diseases
204 using both the genetic maps. Of the 77 QTLs, 54 QTLs (two for thrips, 15 for TSWV and 37
205 for LS) were placed on the F₂ map (Supplemental Table S2, Figure 2) while 23 QTLs (one
206 for thrips, nine for TSWV and 13 for LS) on the F₅ map (Supplemental Table S3, Figure 3)
207 with phenotypic variance (PV) range of 5.86-19.43% (thrips), 4.40-34.92% (TSWV) and
208 5.20-21.45% (LS), respectively (Table 1). The LOD values ranged from 2.51 (TSWV, LS) to
209 5.92 (TSWV) in F₂ map while 2.50 (TSWV) to 6.38 (LS) in F₅ map.

210 **QTLs Identified for Thrips**

211 Total three QTLs could be identified for thrips using genetic mapping information of both the
212 populations and phenotyping data generated for one season at Dawson during 2010. Of the
213 three QTLs, two QTLs (*qF2TPS1* and *qF2TPS2*) were detected on F₂ map with PV ranging
214 from 12.14% to 19.43% while only one QTL (*qF5TPS1*) with 5.86% PV on F₅ map. Among
215 three QTLs, the *qF2TPS1* (IPAHM108-2 – AHGS0347) located on AhIX and *qF2TPS2*
216 (GM2337 – TC42A02) located on AhX are the two major QTLs detected for thrips with
217 12.14% PV and 19.43% PV, respectively (Table 2).

218

219 **QTLs Identified for Tomato Spotted Wilt Virus (TSWV)**

220 In the case of TSWV, a total of 24 QTLs were detected which include 15 QTLs from F₂ and
221 nine QTLs from F₅ map with PV ranging from 4.40% to 34.92% and 5.20% to 14.14%,
222 respectively (Table 1). All the 15 QTLs detected in F₂ map were located on eight genomic
223 regions of six LGs (AhI, AhII, AhIX, AhX, AhXI and AhXII) (Table 2). The same names
224 were given to all the QTLs if they were mapped with same genomic regions/marker interval.
225 So in this case, 15 QTLs were mapped on eight genomic regions as *qF2TSWV1* to
226 *qF2TSWV8* without referring to any season (Supplemental Table S2). The three genomic
227 regions named seq5D5 – GM2744 (*qF2TSWV3*) on AhII, TC42A02 – GM2337 (*qF2TSWV6*)
228 on AhX, and GNB2 – AHO116 (*qF2TSWV8*) on AhXII harbored three QTLs, while another
229 genomic region named IPAHM108-2 – AHGS0347 (*qF2TSWV4*) on AhIX possessed two
230 QTLs and these four genomic regions are referred as consistent QTLs across two or more
231 different environments. The phenotypic variance shown by consistent QTLs were higher in
232 general as compared to the non-consistent QTLs (which appeared in only one environment).
233 Among four consistent QTLs, *qF2TSWV3* had higher PV range (5.14 – 34.92%) followed by
234 *qF2TSWV8* (6.26 – 21.18% PV), *qF2TSWV4* (12.92 – 18.11% PV) and *qF2TSWV6* (10.78 –
235 16.56% PV) (Table 2). Among non-consistent QTLs, *qF2TSWV5* had the highest PV
236 (23.02%) followed by *qF2TSWV7* (15.75%), *qF2TSWV1* (9.41%) and *qF2TSWV2* (4.40%).

237 Similarly, all the nine QTLs (5.20 – 14.14% PV) identified in F₅ map were located on
238 seven genomic regions on seven different LGs named LGT1 (TC3H02-410 – seq14A7-300),
239 LGT6 (TC11A02-300 – GNB523-500), LGT7 (GNB519-205 – GM1076-460), LGT9
240 (AC3C07-70 – RN35H04-1500), LGT11 (GNB619-340 – GM2607-90), LGT12 (seq14G03-
241 500 – GM2808-400), and LGT25 (IPAHM167-130 – GM1555-1000). These genomic regions
242 were named as *qF5TSWV1* to *qF5TSWV7*, respectively (Supplemental Table S3). Two

243 genomic regions i.e, *qF5TSWV4* (AC3C07-70 – RN35H04-1500) and *qF5TSWV7*
244 (IPAHM167-130 – GM1555-1000) were consistent as both harbored two QTLs for TSWV
245 which were located on LGT9 and LGT25 with PV range of 11.45 – 14.14% and 7.25 –
246 7.62%, respectively (Table 3). Among the five non-consistent QTLs, *qF5TSWV5* had high
247 PV (10.80%) followed by *qF5TSWV6* (10.64%), *qF5TSWV1* (9.31%), *qF5TSWV2* (7.71%)
248 and *qF5TSWV3* (5.20%).

249 Of the 15 QTLs detected in the F₂ map, 11 QTLs were contributed by Tifrunner while
250 four QTLs were contributed by GT-C20 with additive effects, ranging from -0.443
251 (*qF2TSWV8*) to -1.250 (*qF2TSWV6*) and 0.797 (*qF2TSWV7*) to 1.347 (*qF2TSWV4*),
252 respectively (Supplemental Table S2). Similarly in the case of F₅ map, five QTLs were
253 contributed by Tifrunner while four QTLs were contributed by GT-C20 with additive effects,
254 ranged from -0.235 (*qF5TSWV3*) to -3.860 (*qF5TSWV1*) and 0.332 (*qF5TSWV6*) to 0.401
255 (*qF5TSWV4*), respectively (Supplemental Table S3).

256 **QTLs Identified for Leaf Spot (LS)**

257 QTL analyses for ten different phenotyping data of LS led to identification of a total of 50
258 QTLs, which include 37 QTLs for F₂ map and 13 QTLs for F₅ map with PV ranging from
259 6.61% to 27.35% and 5.95% to 21.45%, respectively (Table 1). All the 37 QTLs detected in
260 F₂ map were located on 12 genomic regions of nine LGs (AhII, AhV, AhVI, AhVIII, AhIX,
261 AhX, AhXI, AhXII and AhXVIII). The same names were given to the QTLs if they are
262 mapped with same genomic regions/marker interval. Thus, 37 QTLs mapped on 12 genomic
263 regions on F₂ map were named as *qF2LS1* to *qF2LS12* without referring to any season
264 (Supplemental Table S2).

265 The seven genomic regions namely GM2744 – seq5D5 (*qF2LS1*) on AhII, IPAHM108-2
266 – AHGS0347 (*qF2LS5*), TC5A07 – TC7G10 (*qF2LS6*) and TC42A02 – GM2337 (*qF2LS7*)

267 on AhIX, seq2G4 – PM499 (*qF2LS8*) and PM200 – AC2C05 (*qF2LS9*) on AhXI, and GNB2
 268 – AHO116 (*qF2LS10*) harbored five, four, four, nine, four, two and four QTLs, respectively,
 269 thus, these seven genomic regions are referred as consistent QTLs (Table 2). The phenotypic
 270 variances explained by the consistent QTLs for LS were higher in general as compared to the
 271 non-consistent QTLs. Among seven consistent QTLs, three consistent QTLs namely *qF2LS5*,
 272 *qF2LS6* and *qF2LS7* contributed more or less equally as their PV ranged from 11.27 –
 273 24.45%, 10.8 – 24.19% and 13.48 – 24.85%, respectively, followed by *qF2LS10* (15.30 –
 274 21.19%), *qF2LS8* (6.61 – 18.97%), *qF2LS1* (7.80 – 13.11%) and *qF2LS9* (10.29 – 11.51%)
 275 (Table 2). Similarly among the five non-consistent QTLs, *qF2LS11* had the highest PV
 276 (27.35%) followed by *qF2LS3* (12.56%), *qF2LS12* (11.59%), *qF2LS2* (8.22%), and *qF2LS4*
 277 (8.11%).

278 The 13 QTLs (5.95 – 21.45% PV) identified on F₅ map were located on 11 genomic
 279 regions of eight different LGs, LGT3, LGT5, LGT6, LGT7, LGT14, LGT16, LGT17 and
 280 LGT18 (Supplemental Table S3). These genomic regions were named as *qF5LS1* to
 281 *qF5LS11*, respectively. Two genomic regions i.e, *qF5LS5* (TC7C06-170 – seq15D3-500) and
 282 *qF5LS10* (GM1254-160 – seq15C10-205) were consistent as both harbored two QTLs for
 283 LS, which were located on LGT6 and LGT10 with PV range of 7.61 – 11.20% and 7.50 –
 284 9.08%, respectively. Among the nine non-consistent QTLs, *qF5LS1* had a PV of 21.45%,
 285 while the remaining eight QTLs (*qF5LS2*, *qF5LS3*, *qF5LS4*, *qF5LS6*, *qF5LS7*, *qF5LS8*,
 286 *qF5LS9* and *qF5LS11*) had a low PV range of 5.95% (*qF5LS8*) to 8.98% (*qF5LS3*) (Table 3).

287 Of the 37 QTLs detected for LS in F₂ map, 20 QTLs were contributed by Tifrunner while
 288 17 QTLs were contributed by GT-C20 with additive effects, ranged from -0.861 (*qF2LS1* at
 289 season TF11E4) to -2.921 (*qF2LS10* at season TF12E8) and 0.720 (*qF2LS8* at season
 290 TF11E6) to 4.629 (*qF2LS11* at season TF12E9), respectively (Supplemental Table S2).

291 Similarly in F₅ map, of the 13 QTLs detected for LS, five QTLs were contributed by
292 Tifrunner while six QTLs were contributed by GT-C20 with additive effects, ranged from -
293 0.1739 (*qF5LS4* at season TF11E6) to -2.430 (*qF5LS1* at season TF11E4) and 0.1311
294 (*qF5LS2* at season TF11E4) to 0.2733 (*qF5LS7* at season DW10E2), respectively
295 (Supplemental Table S3).

296 **Common QTLs Identified Among the Traits**

297 Two common regions were identified in F₂ map for all the three diseases. The first common
298 genomic region (GM2337 – TC42A02) was located on AhX which harbored one QTL for
299 thrips (*qF2TPS2*), three QTLs for TSWV (*qF2TSWV6* for 3 seasons) and nine QTLs for LS
300 (*qF2LS7* for 9 of the total 10 seasons). This genomic region is contributing 19.43% PV for
301 thrips, 10.78 – 16.56% PV for TSWV and 13.48 – 24.85% PV for LS. In all the three traits,
302 the phenotypic contribution came from the resistant parent, Tifrunner. The second common
303 region (IPAHM108-2 – AHGS0347) located on AhIX harbored one QTL for thrips
304 (*qF2TPS1*), two QTLs for TSWV (*qF2TSWV4* for 2 seasons) and four QTLs for LS (*qF2LS5*
305 for 4 of the total 10 seasons). This genomic region is contributing 12.14% PV for thrips,
306 12.92 – 18.11% PV for TSWV and 11.27 – 24.45% PV for LS. Interestingly, for all the three
307 diseases, the phenotypic contribution came from the susceptible parent, GT-C20, for this
308 second common region.

309 Further, other four genomic regions harbored QTLs for both TSWV and LS. These four
310 genomic regions are located on four different LGs of F₂ map i.e., on AhII (GM2744 –
311 seq5D5), AhIX (TC5A07 – TC7G10), AhXI (seq2G4 – PM499), and AhXII (GNB2 –
312 AHO116). The first genomic region (GM2744 – seq5D5) harbored three QTLs for TSWV
313 (5.14 – 34.92% PV) and five QTLs for LS (7.80 – 13.11% PV) with the contribution from the
314 resistant parent, Tifrunner. Similarly, the second genomic region (TC5A07 – TC7G10)

315 harbored a single QTL for TSWV (23.02% PV) and four QTLs for LS (10.08 – 24.19% PV)
316 with the contribution coming from the susceptible parent, GT-C20. The third genomic region
317 (seq2G4 – PM499) harbored a single QTL for TSWV (15.75% PV) and four QTLs for LS
318 (6.61 – 18.97% PV) contributed by the susceptible parent, GT-C20. The fourth genomic
319 region (GNB2 – AHO116) harbored three QTL for TSWV (6.26 – 21.18% PV) and four
320 QTLs for LS (15.30 – 21.19% PV) contributed by the resistant parent, Tifrunner.

321 In contrast to the F₂ map, there was no common QTL for all three traits in the F₅ map.
322 There was only one common genomic region located on LGT6 (TC11A02-300 – GNB523-
323 500) harboring one QTL for TSWV (*qF5TSWV2*) with 7.71% PV and LS (*qF5LS4*) with
324 8.02% PV.

325 **Common QTLs Identified Between Two Maps**

326 There was one QTL controlling LS in F₂ map (AhXVIII) and one QTL controlling TSWV in
327 F₅ map (LGT7) flanked by same markers i.e., GNB159 – GNB335. In the other case, even
328 though the flanking markers were not same but the QTLs were found on the same linkage
329 group. Such QTLs have been observed between corresponding LGs of both genetic maps, for
330 example between AhII and LGT17, AhV and LGT16, AhVI and LGT11, and AhX and
331 LGT6.

332

333 **DISCUSSION**

334 Due to the increased uniformity in marker nomenclature, the corresponding linkage groups
335 between these two maps have been identified. Further, a total of 9 of the 10 LGs from A-
336 genome and 8 of the 10 LGs from B-genome could be assigned after comparing these two
337 genetic maps with the reference consensus genetic maps using the common marker loci
338 (Gautami et al., 2012). In general, a good co-linearity has been observed for these two genetic

339 maps and with the reference consensus genetic map (Figure 1). This population has shown
340 great potential not only for genetic mapping but also for identification of QTLs to several
341 economically important traits such as morphological descriptors, oil quality, and disease
342 resistance. Here, successful attempt was made to make use of both the genetic maps and the
343 identified QTLs for the three resistance traits to thrips, Tomato spotted wilt virus (TSWV)
344 and leaf spots (LS).

345 RIL population is a set of genotypes of highly inbred F_2 lines. RILs approach
346 complete homozygosity for all loci as the number of generations of inbreeding approaches
347 infinity. In practice, the convention is to use six to eight generations of inbreeding, resulting
348 in ~99.84 to 99.96% homozygosity respectively. A major advantage of RILs is that the
349 descendents of any one RIL are genetically identical, hence “immortal”, allowing RILs to be
350 marker-genotyped once and phenotyped repeatedly in multiple labs and experiments
351 (Elnaccash and Tonsor, 2010). It is well understood that RIL-based QTL analysis is more
352 reliable than the F_2 -based mapping populations for identification of QTLs. Majority of the
353 studies showed identification of large number of QTLs with overestimated phenotypic effect.
354 However, none of the study was conducted at both the stages (F_2 and RIL) using the same
355 population and thus, this study was focused on using genotyping data generated at F_2 and F_5
356 generation and phenotyping data generated at F_8 generation onwards on the same population.
357 Phenotyping data generated on this population after F_8 generation was used for both the
358 genetic maps to identify QTLs for the three traits, thrips, Tomato spotted wilt virus (TSWV)
359 and leaf spots (LS). Therefore, a total of 77 QTLs were identified in these two maps, 54
360 QTLs in F_2 map (Figure 2) while 23 QTLs in F_5 map (Figure 3) with PV up-to 19.43%
361 (thrips), 34.92% (TSWV) and 21.45% (LS), respectively.

362 We should therefore expect that the F_2 and the RIL populations might show high
363 phenotypic variance and this effect will be exaggerated in RIL compared to the F_2 because

364 all individuals are homozygous at virtually all loci, and large sample size in RIL reducing the
365 variance of the mean and transgressive segregation and homozygosity increasing the mean's
366 variance (Beavers, 1998). As expected, the phenotypic variance explained by QTLs detected
367 in F₂ map showed relatively higher phenotypic variance as compared to F₅ map. Occurrence
368 of more QTLs with relatively higher estimation of phenotypic effect in F₂ map than the F₅
369 map was due to presence of higher level of heterozygosity in F₂ generation. Nevertheless,
370 this study has provided comparative QTL analysis using genotyping data generated at F₂ and
371 F₅ generation on the same population and confirms the assumption established based on
372 studies on different populations. Because of above two technical deficiencies (higher number
373 of QTLs and high estimation of phenotypic variance) of using F₂ population for conducting
374 QTL analysis, earlier studies support the use of RIL populations such as double haploids and
375 RILs. These RIL populations have additional advantage of being useful for phenotyping the
376 population for multiple season/location in order to identify consistent (across seasons) and
377 stable (across locations) QTLs.

378 It was interesting to note that not only alleles of the resistant parent have contributed
379 towards the total phenotypic variance but the susceptible parent also made significant
380 contribution through favorable alleles. For thrips no study so far has been conducted while
381 for TSWV, earlier using the same population, Qin et al. (2012) reported one QTL with 12.9%
382 PV (*qtsww1*). Beside above QTL, no other QTL for TSWV has been reported so far in peanut.
383 Therefore, all the QTLs identified in current study for thrips and TSWV are novel in nature
384 and are of great importance for further study and their deployment in molecular breeding.

385 The highest PV explained by any QTL for leaf spot was 27.35% (*qF2LS11*) in present
386 study, while earlier QTL analysis using extensive phenotyping data on two RIL populations
387 (TAG 24 × GPBD 4 and TG 26 × GPBD 4) for 7–8 seasons and genotyping data (207 marker
388 loci each) resulted in identification of a total of 28 QTLs for late leaf spot (LLS; 10.1 to

389 67.8% PV) (Khedikar et al., 2010; Sujay et al., 2012). These QTLs include a major QTL for
390 LLS with upto 62.34% PV flanked by GM1573/GM1009 and seq8D09.

391 Plants possess a strong immune system and defense mechanism to prevent themselves
392 from the pathogens. Thus common genomic regions controlling more than one disease may
393 be even more important in order to improve plant resilience. Considering the above
394 hypothesis, two common genomic regions (GM2337 – TC42A02 and IPAHM108-2 –
395 AHGS0347) were identified in F₂ map for all the three diseases, while four common genomic
396 regions (GM2744 – seq5D5, TC5A07 – TC7G10, seq2G4 – PM499 and GNB2 – AHO116)
397 in F₂ map and one common genomic region (TC11A02-300 – GNB523-500) in F₅ map were
398 identified for LS and TSWV. The presence of common QTLs has also been reported by Sujay
399 et al., (2012) where in three genomic regions harbored QTLs from two populations for both
400 leaf rust and late leaf spot. Thus, these common genomic regions may harbor genes which
401 play major role in plant defense against several pathogens and hence can be used for
402 improving resistance for more than one disease through increasing resistance.

403 In summary, through screening more than 5000 markers, genetic maps upto 329
404 marker loci have been developed. High DNA polymorphism and high phenotypic variability
405 between parental genotypes have made the T-population a very good genetic material for
406 identification of linked markers through QTL analysis to thrips, TSWV and LS. Common
407 genomic regions controlling more than one disease has also been identified with significant
408 contribution towards disease resistance. Thus, this population has shown great potential for
409 dense genetic mapping and identification of QTLs controlling several disease and agronomic
410 traits in peanut. In addition it was evident that the number of QTLs and the estimates of
411 phenotypic variance were reduced in F₅ map. The identified QTLs, consistent or not, will be
412 studied further through fine mapping for potential use in breeding for genetic improvement of
413 disease resistance in peanut.

414

415 **Acknowledgements**

416 We thank Billy Wilson, Jake Fountain, Stephanie Lee, Lucero Gutierrez and Sara Beth
417 Pelham for technical assistance in the field and the laboratory. This research was partially
418 supported by funds provided by the USDA Agricultural Research Service, the Georgia
419 Agricultural Commodity Commission for Peanuts, Peanut Foundation and National Peanut
420 Board. We are also thankful to National Science Foundation to DRC (NSF, DBI-0605251)
421 and USDA/CSREES/Capacity Building Program to GH (#2006-38814-17489) for supporting
422 this research. Mention of trade names or commercial products in this publication is solely for
423 the purpose of providing specific information and does not imply recommendation or
424 endorsement by the U.S. Department of Agriculture.

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511 **Figure Legends**

512 **Figure 1.** Co-linearity between two genetic maps of T-population and reference consensus
513 genetic map.

514 **Figure 2.** QTL locations for thrips, Tomato spotted wilt virus (TSWV) and leaf spots (LS) on
515 F₂ generation-based genetic map of T-population.
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517 **Figure 3.** QTL locations for thrips, Tomato spotted wilt virus (TSWV) and leaf spots (LS) on
518 F₅ generation-based genetic map of T-population.
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524 **Table 1. Summary on QTL analysis based on the F₂ and F₅ population for disease**
 525 **resistance**

Traits/QTLs	Number of QTLs identified	LOD value range	Phenotypic variance range (%)	Additive effect (a0) range
Based on F₂ population				
Thrips	2	2.69 - 3.27	12.14 - 19.43	0.482 -0.608
Tomato spotted wilt virus (TSWV)	15	2.51- 5.92	4.40 - 34.92	1.347 - 0.526
Leaf spot (LS)	37	2.51 - 5.68	6.61 - 27.35	4.629 - 0.720
Based on F₅ population				
Thrips	1	2.51	5.86	0.0518
Tomato spotted wilt virus (TSWV)	9	2.50 - 4.61	5.20 - 14.14	0.400 - 0.249
Leaf spot (LS)	13	2.51 - 6.38	5.95 - 21.45	0.273 - 0.174

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533 **Table 2. Consistent QTLs detected for thrips, TSWV and leaf spots in F₂ T-population**

Traits/QTL name	Linkage group	Marker Interval	Seasons appeared	LOD value	Phenotypic variance (PV) %	Additive effect (a0)
Thrips						
<i>qF2TPS1</i>	AhIX	IPAHM108-2 - AHGS0347	DW10	2.69	12.14	0.482
<i>qF2TPS2</i>	AhX	GM2337 - TC42A02	DW10	3.27	19.43	-0.608
Tomato spotted wilt virus (TSWV)						
<i>qF2TSWV1</i>	AhI	GNB629 - TC31G11	TF10E2	2.62	9.41	-1.153
<i>qF2TSWV2</i>	AhI	GA161 - GNB154	TF10E3	2.51	4.40	-0.685
<i>qF2TSWV3</i>	AhII	seq5D5 - GM2744	TF10E2, TF10E3, TF11E4	2.79 - 5.92	5.14 - 34.92	(-)3.539 to (-) 0.526
<i>qF2TSWV4</i>	AhIX	IPAHM108-2 - AHGS0347	TF10E2, TF10E3	3.99 - 4.84	12.92 - 18.11	1.024 to 1.347
<i>qF2TSWV5</i>	AhIX	TC5A07 - TC7G10	TF11E4	4.42	23.02	1.120
<i>qF2TSWV6</i>	AhX	TC42A02 - GM2337	TF10E2, TF10E3, TF11E4	3.01 - 3.28	10.78 - 16.56	(-)1.250 to (-) 0.743
<i>qF2TSWV7</i>	AhXI	seq2G4 - PM499	TF11E4	2.93	15.75	0.797
<i>qTSWV8</i>	AhXII	GNB2 - AHO116	DW10E1, TF10E2, TF10E3	2.61 - 4.16	6.26 - 21.18	(-) 1.374 to (-) 0.443
Leaf spots (LS)						
<i>qF2LS1</i>	AhII	GM2744 - seq5D5	TF10E3, TF11E4, TF11E5, TF11E6, TF11E7	2.69 - 3.59	7.80 - 13.11	(-) 1.422 to (-) 0.861
<i>qF2LS2</i>	AhV	TC1B02 - TC4A02	TF12E9	2.54	8.22	1.399
<i>qF2LS3</i>	AhVI	GM2724 - GNB619	TF11E6	2.68	12.56	1.064
<i>qF2LS4</i>	AhVIII	PM36 - GM2137	TF12E9	2.78	8.11	-1.771
<i>qF2LS5</i>	AhIX	IPAHM108-2 - AHGS0347	DW10E1, DW10E2, TF11E6, TF11E7	2.51 - 5.68	11.27 - 24.45	1.188 - 2.262
<i>qF2LS6</i>	AhIX	TC5A07 - TC7G10	TF10E3, TF11E4, TF11E5, TF12E10	3.33 - 5.01	10.8 - 24.19	1.253 - 1.834

<i>qF2LS7</i>	AhX	TC42A02 - GM2337	DW10E1, DW10E2, TF10E3 TF11E4, TF11E5, TF11E6, TF11E7, TF12E8, TF12E10	2.51 - 4.82	13.48 - 24.85	(-)2.519 to (-) 0.978
<i>qF2LS8</i>	AhXI	seq2G4 - PM499	TF10E3, TF11E5, TF11E7, TF12E8	2.55 - 3.52	6.61 - 18.97	0.720 - 1.399
<i>qF2LS9</i>	AhXI	PM200 - AC2C05	TF11E4, TF11E5	2.51 - 2.70	10.29 - 11.51	0.738 - 1.347
<i>qF2LS10</i>	AhXII	GNB2 - AHO116	TF10E3,TF11E4, TF11E5, TF12E8	2.65 - 2.90	15.30 - 21.19	(-)1.208 - (-2.921)
<i>qF2LS11</i>	AhXVIII	GNB904 - GNB625	TF12E9	3.54	27.35	4.629
<i>qF2LS12</i>	AhXVIII	GNB159 - GNB335	TF12E9	3.11	11.59	-2.497

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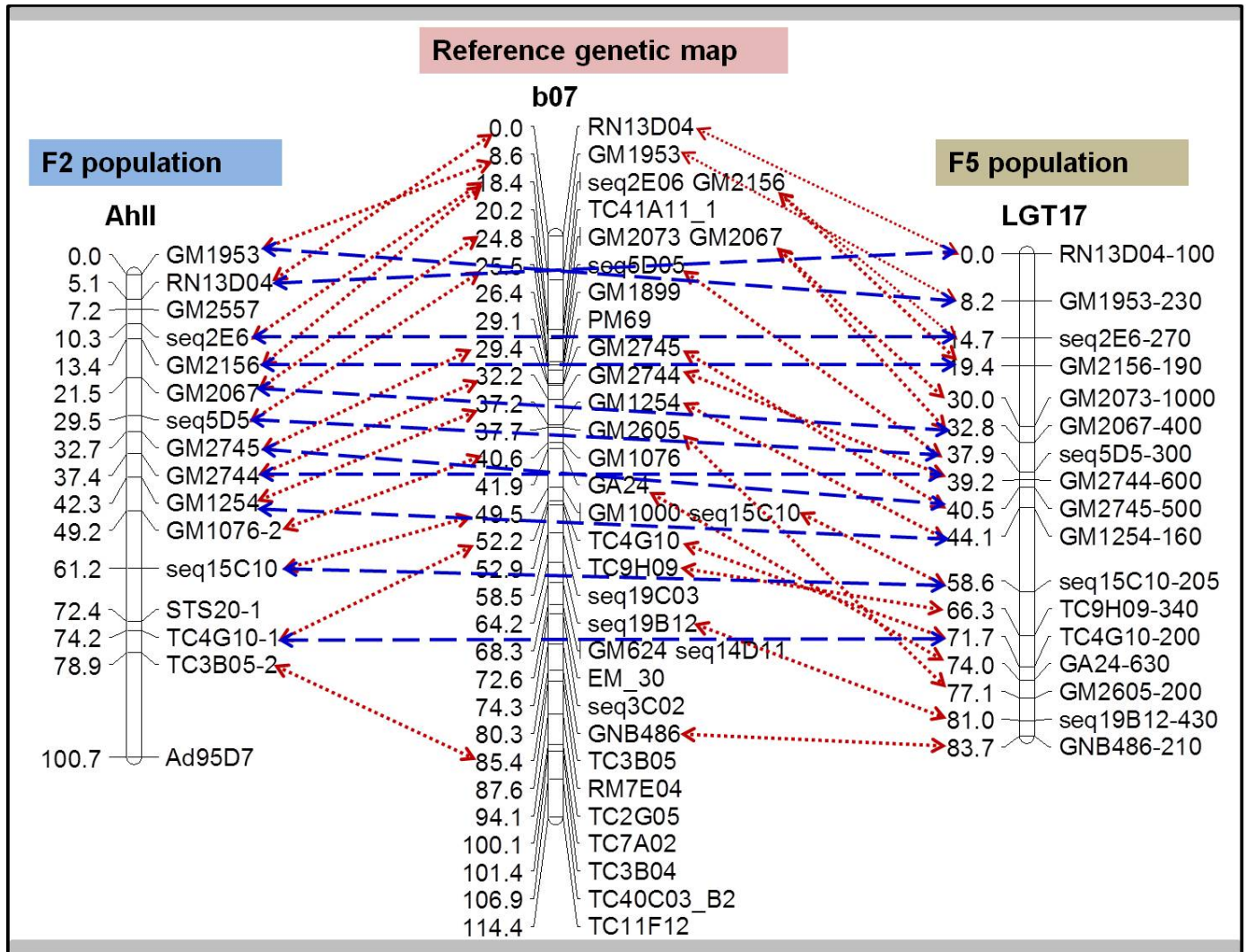
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538 **Table 3. Summary of QTLs detected for thrips, TSWV and leaf spots in F₅ T-population**

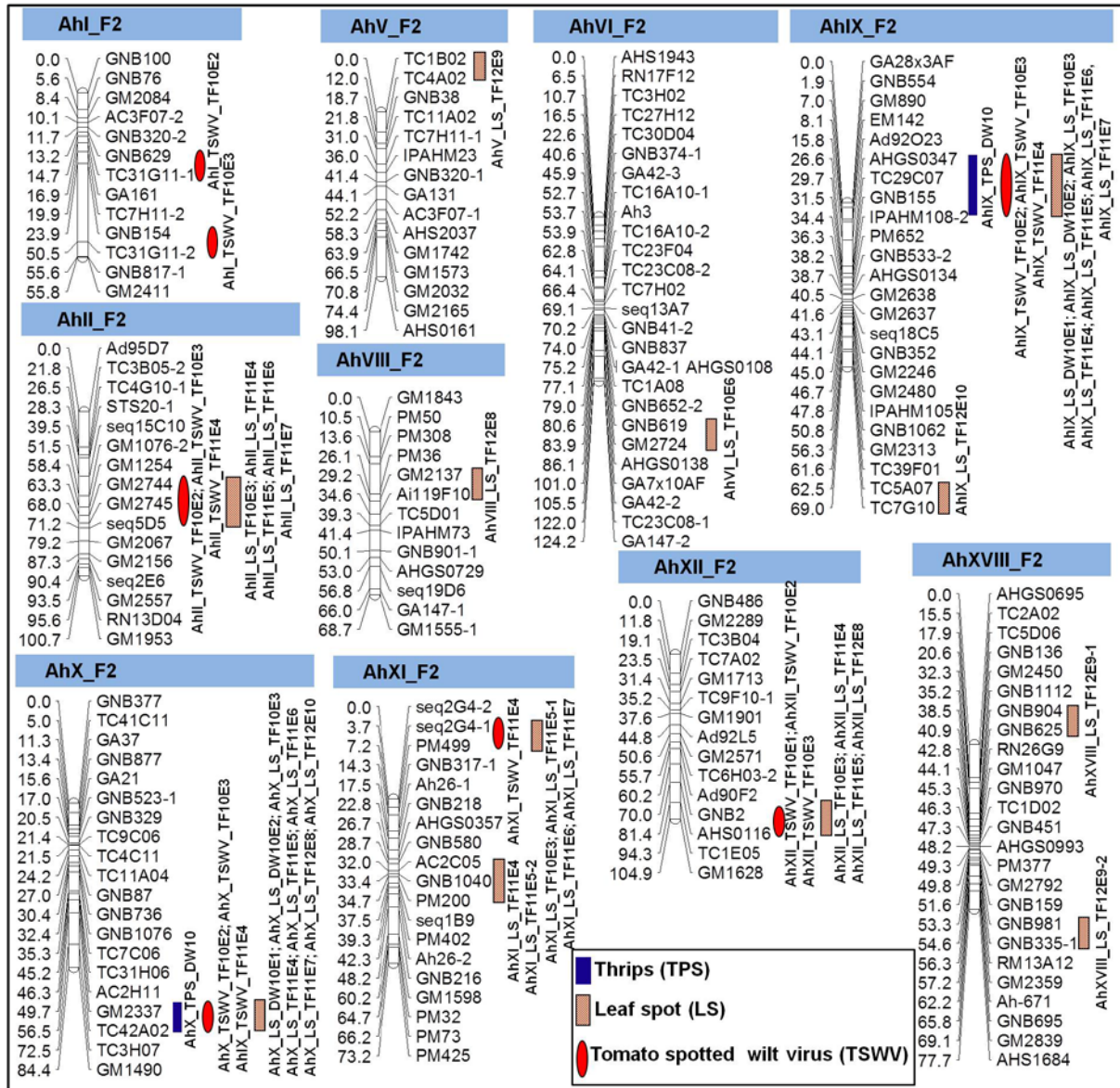
Trait/QTLs	Linkage group	Marker Interval	Seasons appeared	LOD value	Phenotypic variance (PV) %	Additive effect (a0)
Thrips						
<i>qF5TPS1</i>	LGT2	RI1F06-410 - Ah26-195	DW10	2.51	5.86	0.0518
Tomato spotted wilt virus (TSWV) disease						
<i>qF5TSWV1</i>	LGT1	TC3H02-410 - seq14A7-300	TF10E2	3.56	9.31	-0.3860
<i>qF5TSWV2</i>	LGT6	TC11A02-300 - GNB523-500	TF11E4	3.50	7.71	-0.2486
<i>qF5TSWV3</i>	LGT7	GNB519-205 - GM1076-460	TF10E3	2.50	5.20	-0.2357
<i>qF5TSWV4</i>	LGT9	AC3C07-70 - RN35H04-1500	DW10E1, TF11E4	3.90 - 4.61	11.45 -14.14	0.335 - 0.401
<i>qF5TSWV5</i>	LGT11	GNB619-340 - GM2607-90	DW10E1	3.50	10.80	0.3453
<i>qF5TSWV6</i>	LGT12	seq14G03-500 - GM2808-400	TF10E3	3.40	10.64	0.3318
<i>qF5TSWV7</i>	LGT25	IPAHM167-130 - GM1555-1000	TF10E2, TF10E3	2.52 - 2.60	7.25 - 7.62	(-)0.347 to (-)0.274
Leaf spots (LS) disease						
<i>qF5LS1</i>	LGT3	TC1E06-370 - PM238-150	TF11E4	6.38	21.45	-0.2430
<i>qF5LS2</i>	LGT3	seq19G7-150 - GNB649-250	TF11E4	2.65	6.10	0.1311
<i>qF5LS3</i>	LGT5	GM1878 - GM637-240	TF11E6	3.25	8.98	-0.1835
<i>qF5LS4</i>	LGT6	TC11A02-300 - GNB523-500	TF11E6	3.35	8.02	-0.1739
<i>qF5LS5</i>	LGT6	TC7C06-170 - seq15D3-500	TF11E6, TF11E7	2.94 - 3.41	7.61 - 11.20	(-)0.169 to (-)0.199
<i>qF5LS6</i>	LGT6	TC3H07-500 - TC3H07-280	TF10E3	2.53	8.15	-0.1939
<i>qF5LS7</i>	LGT7	seq3B8-400 - GM1880-2000	DW10E2	2.69	7.35	0.2733
<i>qF5LS8</i>	LGT14	seq14D11-180 - IPAHM451-300	TF11E5	2.65	5.95	0.2218
<i>qF5LS9</i>	LGT16	GM678-300 - GM1742-1300	TF10E3	2.74	7.04	-0.1818
<i>qF5LS10</i>	LGT17	GM1254-160 - seq15C10-205	TF10E3, TF11E6	2.51 - 2.95	7.5 - 9.08	0.172 - 0.212
<i>qF5LS11</i>	LGT18	IPAHM229-170 - IPAHM219-155	TF11E7	3.70	8.71	0.1762

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Figure 1



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Figure 2

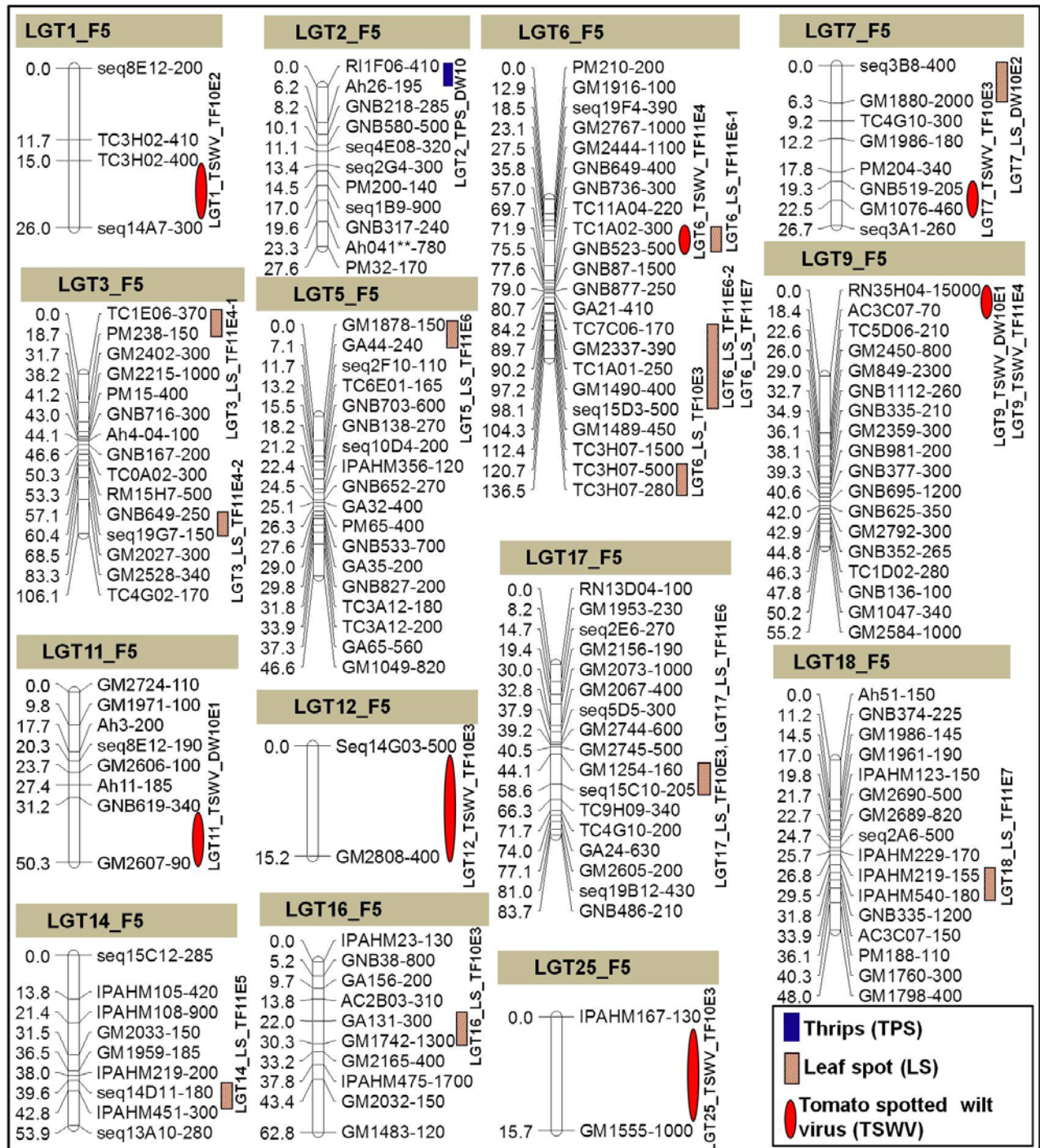


Figure 3

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