



# Identifying *Verticillium dahliae* Resistance in Strawberry Through Disease Screening of Multiple Populations and Image Based Phenotyping

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### \*Correspondence:

Richard Jonathan Harrison  
Richard.Harrison@niab.com  
orcid.org/0000-0002-3307-3519  
† orcid.org/0000-0002-7375-1804

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Helen M. Cockerton<sup>1†</sup>, Bo Li<sup>1</sup>, Robert J. Vickerstaff<sup>1</sup>, Catherine A. Eyre<sup>2</sup>, Daniel J. Sargent<sup>2</sup>, Andrew D. Armitage<sup>1</sup>, Cesar Marina-Montes<sup>1</sup>, Ana Garcia-Cruz<sup>1</sup>, Andrew J. Passey<sup>1</sup>, David W. Simpson<sup>1</sup> and Richard Jonathan Harrison<sup>1\*</sup>

<sup>1</sup> NIAB EMR, East Malling, United Kingdom, <sup>2</sup> Driscoll's Genetics Ltd., East Malling Enterprise Centre, East Malling, United Kingdom

*Verticillium dahliae* is a highly detrimental pathogen of soil cultivated strawberry (*Fragaria x ananassa*). Breeding of *Verticillium* wilt resistance into commercially viable strawberry cultivars can help mitigate the impact of the disease. In this study we describe novel sources of resistance identified in multiple strawberry populations, creating a wealth of data for breeders to exploit. Pathogen-informed experiments have allowed the differentiation of subclade-specific resistance responses, through studying *V. dahliae* subclade II-1 specific resistance in the cultivar “Redgauntlet” and subclade II-2 specific resistance in “Fenella” and “Chandler.” A large-scale low-cost phenotyping platform was developed utilizing automated unmanned vehicles and near infrared imaging cameras to assess field-based disease trials. The images were used to calculate disease susceptibility for infected plants through the normalized difference vegetation index score. The automated disease scores showed a strong correlation with the manual scores. A co-dominant resistant QTL; *FaRVd3D*, present in both “Redgauntlet” and “Hapil” cultivars exhibited a major effect of 18.3% when the two resistance alleles were combined. Another allele, *FaRVd5D*, identified in the “Emily” cultivar was associated with an increase in *Verticillium* wilt susceptibility of 17.2%, though whether this allele truly represents a susceptibility factor requires further research, due to the nature of the F1 mapping population. Markers identified in populations were validated across a set of 92 accessions to determine whether they remained closely linked to resistance genes in the wider germplasm. The resistant markers *FaRVd2B* from “Redgauntlet” and *FaRVd6D* from “Chandler” were associated with resistance across the wider

germplasm. Furthermore, comparison of imaging versus manual phenotyping revealed the automated platform could identify three out of four disease resistance markers. As such, this automated wilt disease phenotyping platform is considered to be a good, time saving, substitute for manual assessment.

**Keywords:** disease resistance, *Fragaria x ananassa*, wilt, NDVI, NBS, breeding

## INTRODUCTION

*Verticillium dahliae* (Kleb.) is a soilborne plant pathogen which has a large detrimental impact on the yield of soil cultivated strawberry (*Fragaria x ananassa*) (Maas, 1998). This ascomycete fungi is particularly problematic due to the longevity of inoculum in the soil whereby the resting propagules, termed microsclerotia, persist for up to 14 years in the absence of a host plant (Schnathorst, 1981). Low inoculum densities of 2 cfu per gram of soil can result in complete strawberry crop losses (Harris and Yang, 1996), indicating that strawberry exhibits a very high susceptibility to *Verticillium* alongside the crops cotton (Paplomatas et al., 1992) and olive when artificially inoculated (López-Escudero and Blanco-López, 2007). *Verticillium* infects over 200 different dicotyledonous plant species including many horticultural crops and weeds (Woolliams, 1966; Bhat and Subbarao, 1999) meaning that crop rotation is an ineffective form of disease control (Atallah et al., 2011). Effective disease control is also hampered by the absence of curative fungicides and restriction of preventative chemical fumigants due to European regulations (e.g., 91/414/EEC; Colla et al., 2012). Disease resistant germplasm is therefore an essential resource required to combat the pathogen, particularly where countries rely predominantly on soil cultivation systems.

A pathogenesis related protein which catalyzes chitinase from wild tomato has been shown to be effective against *V. dahliae* when transformed into strawberry (Chalavi and Tabaeizadeh, 2003). This mechanism acts before infection therefore indicating very strong resistance as proven by the percentage infection of *Verticillium* in strawberry crowns. Complete resistance has not been observed in natural populations of octoploid strawberry to date. Tolerance, whereby the host is colonized by the fungus but does not exhibit infection symptoms, is frequently observed in strawberry alongside the crop species olive (López-Escudero et al., 2004) potato (Dan et al., 2001), cultivated tomato (Chen et al., 2004; Fradin et al., 2009) and cotton (Bolek et al., 2005; Zhang et al., 2011).

**Abbreviations:** AUDPC, area under the disease progression curve; cfu, colony forming units; ExF, “Emily” x “Fenella” mapping population; *FaRVd\*\**, *Fragaria x ananassa* resistance allele for *Verticillium dahliae* \*\* denotes chromosome (1–7) and subgenome (A–D); FxC, “Flamenco” x “Chandler” mapping population; GCA, general combining ability; HR, hypersensitive response; i35k, istraw35 affymetrix chip; i90k, istraw90 affymetrix chip; NB-ARC, domain name; NBS, nucleotide binding site; NDVI, normalized difference vegetation index; NIR, near infrared; QTL, quantitative trait loci; R, red; rAUDPC, relative area under the disease progression curve; RGB, red green blue; RH%, percentage relative humidity; RxH<sup>a</sup>, “Redgauntlet” x “Hapil” mapping population. Cross one ( $n = 169$ ), screened with mixed inoculum; RxH<sup>b</sup>, “Redgauntlet” x “Hapil” mapping population. Cross two ( $n = 80$ ), screened with isolate 12158; SCA, specific combined ability; SNP, single nucleotide polymorphism; SSR, simple sequence repeat; UAV, unmanned aerial vehicle.

High variation for *V. dahliae* resistance has been observed in Californian strawberry germplasm and empirical selection had led to an increase in resistance (Shaw and Gubler, 1996; Shaw et al., 1997). Studies investigating the GCA for *V. dahliae* resistance in strawberry, found that four out of ten cultivars had a significant GCA indicating a high transmission of resistance or susceptibility status from parent to progeny. This study suggests that *Verticillium* wilt resistance is controlled by additive quantitative genetic components (Masny et al., 2014). Furthermore, a significant SCA in two crosses indicated that some *Verticillium* resistance alleles are non-additive (Masny et al., 2014). Previous studies using *in vitro* strawberry have found *Verticillium* resistance to be controlled by additive genes and in one case a single partially dominant gene (Zebrowska et al., 2006). The study of the “Redgauntlet” x “Hapil” mapping population revealed that multiple small effect QTL control *V. dahliae* resistance (Antanaviciute et al., 2015).

Isolate and cultivar specific interactions complicate the description of resistance and must be considered for robust disease resistance breeding. Segregation of *V. dahliae* into six distinct races has been proposed based on the resistance status of different strawberry varieties (Govorova and Govorov, 1997) indicating a complex series of host-pathogen interactions. By contrast, a simpler dissection of *V. dahliae* isolate virulence has been proposed: two subclades of *V. dahliae* have been isolated from United Kingdom strawberry; II-1 and II-2, which exhibit different average levels of virulence on the susceptible strawberry cultivar “Hapil” (Jiménez-Díaz and Olivares-García, 2017; Fan et al., 2018).

Single major gene resistance to *V. dahliae* has been identified in tomato, lettuce and cotton; the *Ve1* host gene, which recognizes the avirulence pathogen effector *VdAve1*, leads to the separation of *V. dahliae* isolates into two races; those with and without *VdAve1* (Kawchuk et al., 2001; Hayes et al., 2011; Zhang et al., 2011; de Jonge et al., 2012). Fan et al. (2018) conclude that there is an absence of the *VdAve1* gene in *V. dahliae* isolated from United Kingdom strawberry. The exclusive infection of strawberry by “race 2” isolates in the United Kingdom, despite the presence of “race 1” isolates in other United Kingdom hosts, likely suggests a lack of dispersion of *VdAve1* isolates, rather than selection against *Ave1*, as *VdAve1* isolates were also able to infect strawberry. This reduces the relevance of harnessing *Ve1* mediated resistance in future strawberry breeding.

Platforms for strawberry genotyping have advanced substantially over the last decade (Bassil et al., 2015; Verma et al., 2017), however, the low throughput capacity of traditional large scale phenotyping is now the limiting factor restricting pre-breeding research (Mahlein, 2016). Currently, many breeders use manual assessments to quantify the disease resistance status

of plants, which is subjective and time consuming. Imaging techniques have been successfully applied to high-throughput plant phenotyping for the past decade (Barbedo, 2013) and with the development of lightweight UAV for precision agriculture, imaging techniques can be applied to screen large crop areas with centimeter level spatial resolution and accurate positional information (Candiago et al., 2015). Multispectral cameras are lighter than the majority of imaging sensors that can be attached to UAV (Sugiura et al., 2016) they also provide accurate quantification and are a cost effective strategy for disease severity quantification. The most common vegetation index derived from multispectral sensor is the NDVI where a positive NDVI value indicates healthy green vegetation whilst a negative value indicates the absence of vegetation (Candiago et al., 2015).

In this study, a low-cost UAV with global positioning system and multispectral imaging sensor was implemented as part of a phenotyping platform to measure *Verticillium* wilt resistance in strawberry.

We also report a reanalysis of historical data using the “Redgauntlet” and Hapil mapping populations infected with a mixed inoculum of *V. dahliae*, reported by Antanaviciute et al. (2015) using newly generated SNP data and also test additional progeny of “Redgauntlet” and Hapil against a single isolate from subclade II-1. Furthermore, two additional mapping populations are studied to identify putative resistance loci toward a highly virulent subclade II-2 isolate of *V. dahliae*.

## MATERIALS AND METHODS

### Study Area and Experimental Design

Field phenotyping for *V. dahliae* resistance was conducted on three strawberry mapping populations. Mapping populations were produced through crosses between the cultivars “Emily” x “Fenella” (ExF, 181 genotypes), “Flamenco” x “Chandler” (FxC, 140 genotypes) and “Redgauntlet” x “Hapil” (RxH<sup>b</sup>, 160 genotypes). The RxH<sup>b</sup> cross differs from the population described in previous research, as it is a different set of individuals (Sargent et al., 2012; Antanaviciute et al., 2015). The analysis reported in Antanaviciute et al. (2015) used the original “Redgauntlet” x “Hapil” (RxH<sup>a</sup>) cross and SSR markers. This study integrates the Antanaviciute et al. (2015) phenotypic data where, in contrast to the previous analysis, the AUDPC and Best Linear Unbiased Estimate (BLUE) scores were calculated to represent the disease score of each genotype across 3 years of phenotyping. Use of SNP marker genotyping allowed a more powerful analysis and comparison of resistance markers across populations. The validation experiment utilized 92 accessions selected from across the wider germplasm, where 97.7% of the SNPs were polymorphic in at least one individual out of a total 22,296 SNPs. Parent and progeny stock plants were maintained in a polytunnel and runners were pinned down into 9 cm pots before planting in “Calves Leys,” Aylesford, Kent United Kingdom field in autumn 2015 (ExF & FxC) or “Rocks Farm,” East Malling, Kent, United Kingdom in 2016 (Validation & RxH<sup>b</sup>). Plants were arranged East to West with 64 plants per row at 0.6 m intervals in a randomized block design with 5–10 replicate plants

per genotype or accession and parental lines. Black MyPex<sup>®</sup> was used for weed growth suppression and allowed segregation of plant foliage for image analysis. Plants were rainfed with additional overhead irrigation supplied if required. The pre-existing microsclerotia level was quantified using the Harris method (Harris et al., 1993) and found to be 4.2 cfu g<sup>-1</sup> soil in “Calves Leys” and 0.9 cfu g<sup>-1</sup> in “Rocks Farm.” To ensure robust disease symptom expression, plants were inoculated with 10 ml of 4 x 10<sup>6</sup> conidia ml<sup>-1</sup> into the crown and immediate surrounding soil of each strawberry plant. A single, highly virulent isolate of *V. dahliae* (12008) was used as inoculum in March 2016 for ExF, FxC and March 2017 for germplasm experiments. The isolate, 12008, has been used extensively in work conducted by Soares (2004) and Fan et al. (2018) and represents an isolate from *V. dahliae* subclade II-2, the “high virulence subclade” when inoculated onto strawberry. Plants in the RxH<sup>b</sup> phenotyping event were inoculated with isolate 12158 from subclade II-1. All RxH<sup>b</sup> phenotyping events were conducted through a trial plot originally inoculated with a large mixture of *V. dahliae* isolates (Antanaviciute et al., 2015). Weather conditions were 12.2 (±3.7)°C; 76.7 (±8.6) RH% spring 2016, 18.5 (±2.3)°C; 77.4 (±6.3) RH% summer 2016, 13.34 (±0.49)°C; 74.5 (±0.76) RH% spring 2017 and 18.36 (±0.62)°C; 75.76 (±1.12) RH% summer 2017.

### Visual Assessment of *Verticillium* Wilt

Disease scores were recorded five times from June to September at 3-week intervals, plants were scored for percentage wilting disease symptoms on a score of 1–9 depending on severity of leaf wilting where a score of 1 denoted a completely healthy plant; 3 denoted 25% necrotic leaves; 5 denoted 50% necrotic leaves; 7 denoted 75% necrotic leaves and 9 denoted 100% necrosis, a dead plant (Antanaviciute et al., 2015). The AUDPC was calculated across each phenotyping event using the R package “agricolae” (Felipe de, 2017) to predict scores for QTL analysis. AUDPC was calculated as below (Shaner and Finney, 1977).

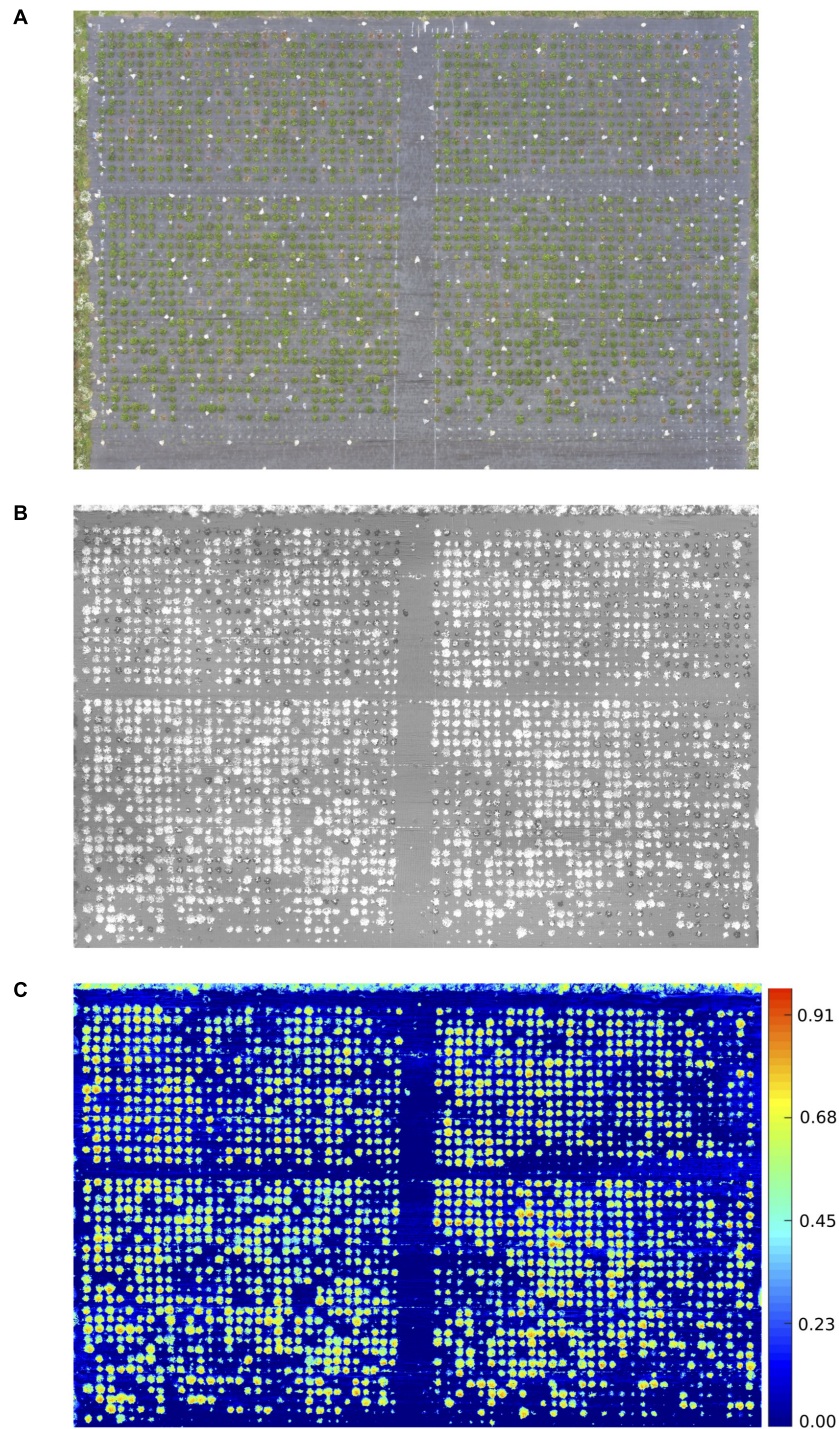
$$AUDPC = \left\{ \sum_{i=1}^{n-1} [(y_{i+1} + y_i) / 2] * [X_{i+1} - X_i] \right\} \quad (1)$$

Where y is the disease score, for score i and X represents time in days and n is the number of scoring events. Relative AUDPC (rAUDPC) was calculated by dividing the AUDPC by the number of days after inoculation.

### Image Acquisition and Processing

Aerial imaging was conducted, in addition to manual scoring, for 2017 field trials. The 2017 trials were of the RxH<sup>b</sup> population and the validation set, the experimental field was 45 m x 30 m in size containing approximately 2500 plants (Figure 1A). The UAV platform was a 1.6 kg DJI Flamewheel F450 quadcopter. RGB images were captured using a Canon SX240 HS, 12 MP digital camera. Multi-spectral images with resolution of 1280 × 960 pixels were captured using a MicaSense RedEdge narrow-band multispectral camera (MicaSense, Seattle, DC, United States). Images were captured at altitude of 30 m at 5 bands including Blue (B: 475 nm center wavelength, 20 nm bandwidth), Green



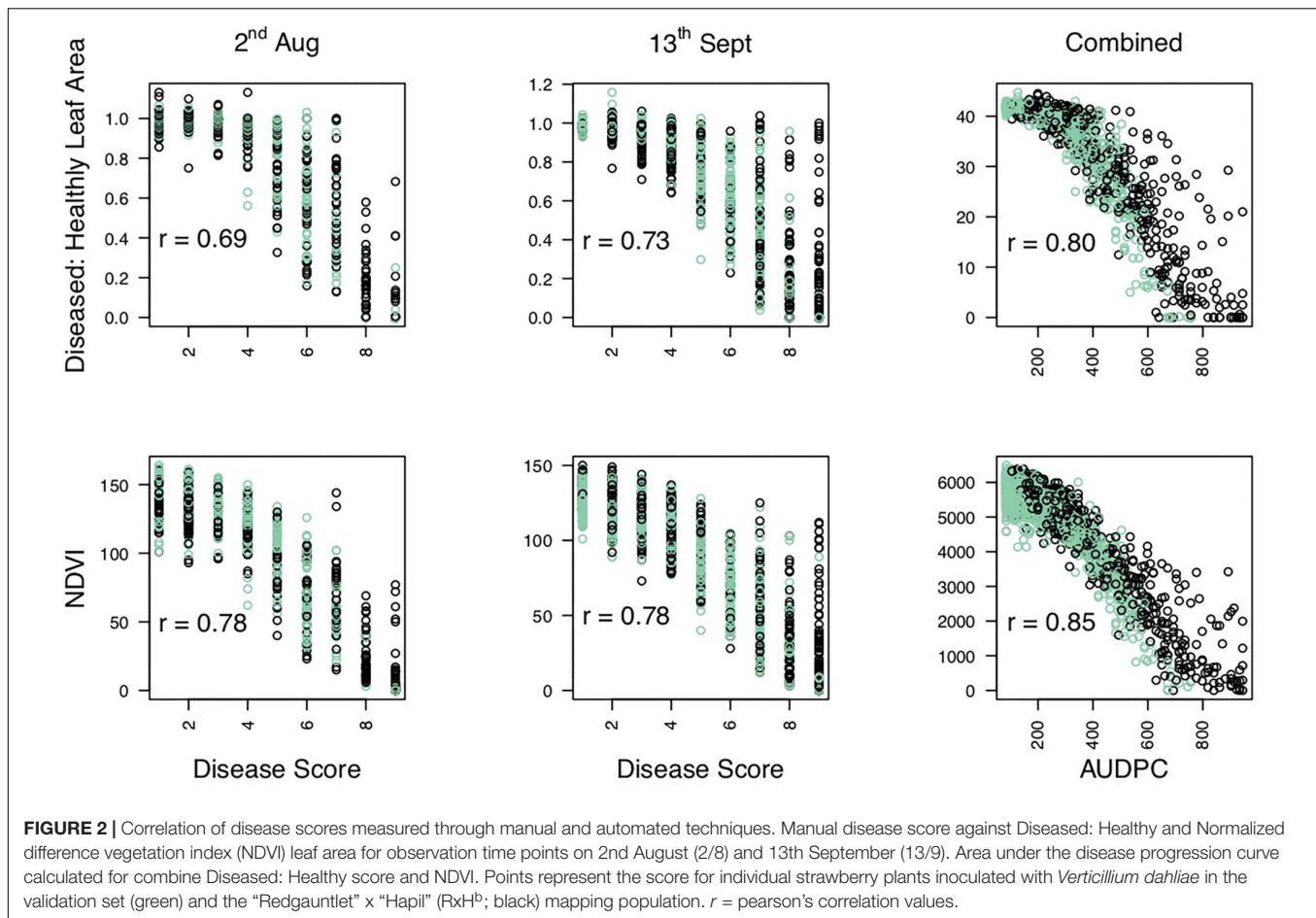


**FIGURE 1** | Aerial image taken using UAV of 2017 Verticillium disease field experiments containing the RxH<sup>b</sup> population and the validation set **(A)** RGB image **(B)** Green: Red ratio mask image of the canopy for each strawberry plant **(C)** Normalised difference vegetation index (NDVI) image with false color of the validation set and the “Redgauntlet” x “Hapii” (RxH<sup>b</sup>) mapping population.

(G: 560 nm, 20 nm), Red (R: 668 nm, 10 nm), Red Edge (RE: 717 nm, 10 nm) and Near Infrared (NIR: 840 nm, 40 nm) were captured simultaneously with the format of 16-bit raw GeoTIFF. Ortho-mosaic images were produced by processing

UAV images in Pix4Dmapper Pro software (Pix4D SA, 1015 Lausanne, Switzerland). Two surveys were undertaken of the experimental plot on the August 2, 2017 at 11:00 and the September 13, 2017 at 12:00. The final image resolution was





1.27 cm<sup>2</sup> pixel<sup>-1</sup>, the resolution of **Figure 1** has been lowered to reduce file size. NDVI was calculated as the normalized ratio between near infrared (NIR) and red (R) bands (Potgieter et al., 2017), which is shown in Eq. (2). The diseased:healthy leaf area was calculated based on the green and total plant pixels, which is shown in Eq. (3).

$$\text{NDVI} = (\text{NIR} - \text{R}) / (\text{NIR} + \text{R}) \quad (2)$$

$$\text{Diseased:healthy leaf area} = \text{green pixels} / \text{total plant pixels} \quad (3)$$

Bandpass thresholding was applied to obtain the mask image of the whole canopy for each strawberry plant, the green:red band ratio image was found to provide a good contrast between the plant canopy and background (**Figure 1B**). A semi-automated image analysis software was developed to calculate the average NDVI value for each plant (**Figure 1C**). Manual selection of a plant on the masked image allows the software to automatically calculate the ratio of total NDVI: total canopy pixel number.

## Linkage Map Generation

The Qiagen DNeasy plant mini extraction kit (Qiagen Ltd., Manchester, United Kingdom) was used to extract DNA from the studied genotypes and accessions according

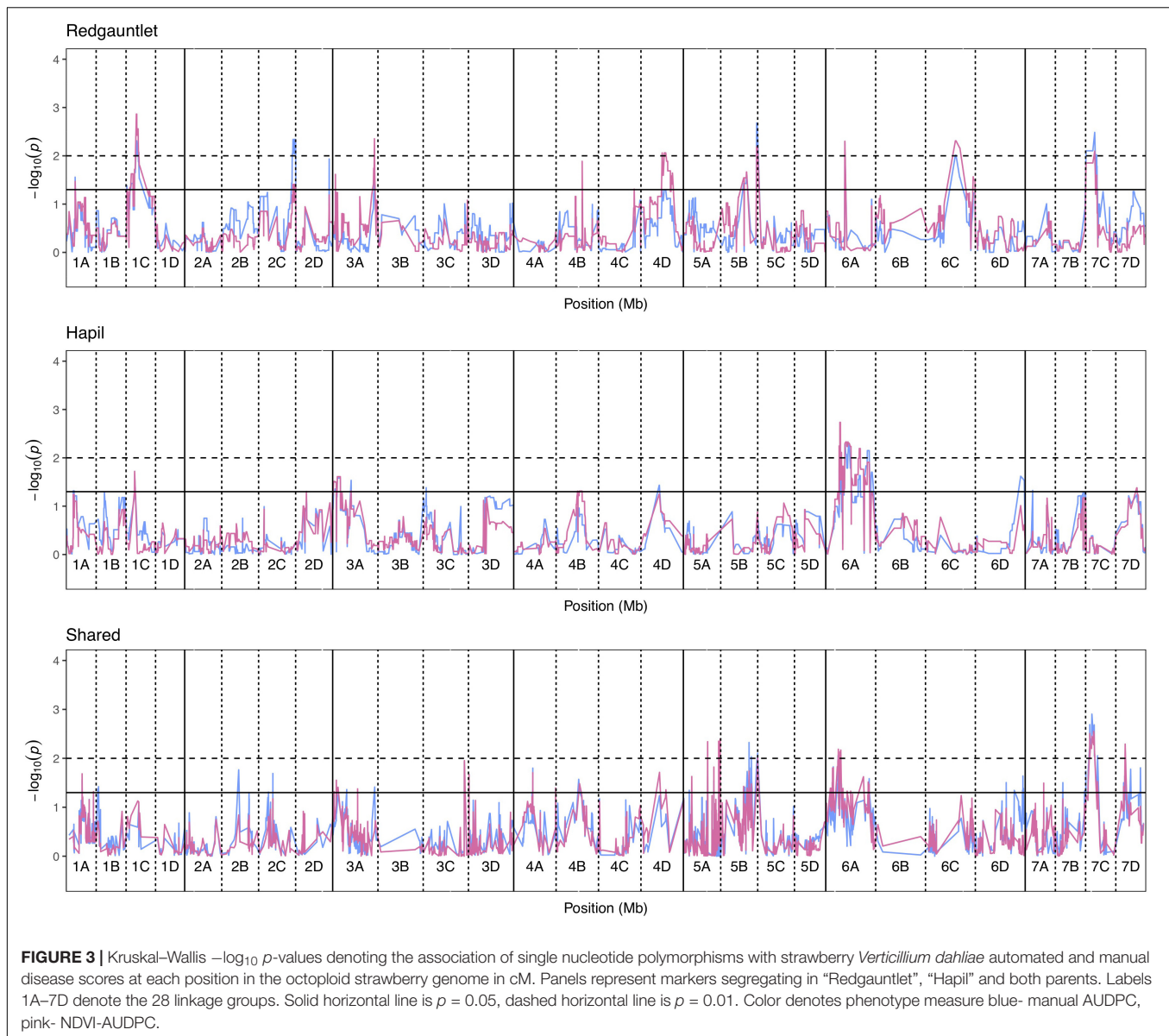
to the manufacturer’s instructions. F1 mapping populations RxH<sup>a</sup>, ExF and FxC were genotyped using the Affymetrix IStraw90 Axiom® array (i90k; Bassil et al., 2015) whereas the population RxH<sup>b</sup> and validation accessions were genotyped on the streamlined Axiom® IStraw35 384HT array (i35k; Verma et al., 2017). Crosslink was used to generate linkage maps<sup>1</sup> a program developed specifically for polyploid plant species (Vickerstaff and Harrison, 2017). *Fragaria x ananassa* chromosome number is denoted by 1–7 and sub-genome number is represented by A–D as specified in van Dijk et al. (2014) and Sargent et al. (2015).

## Statistical Analysis

For the RxH<sup>a</sup> historical data the BLUE was calculated using the relative AUDPC for QTL analysis (R package “nlme,” Pinheiro et al., 2017).

For populations phenotyped with both manual and UAV imaging, the Pearson’s correlation coefficient was calculated between the ratio of healthy:diseased leaf area, NDVI and the raw phenotypic disease score at each time point. A combined analysis used the NDVI-AUDPC and healthy:diseased leaf area-AUDPC alongside the AUDPC disease score to determine the efficacy of the drone phenotyping method. Transgressive

<sup>1</sup><https://github.com/eastmallresearch/crosslink>



segregation where progeny wilt phenotype varied more than expected based on parental phenotypes was assessed using a Dunnett’s test.

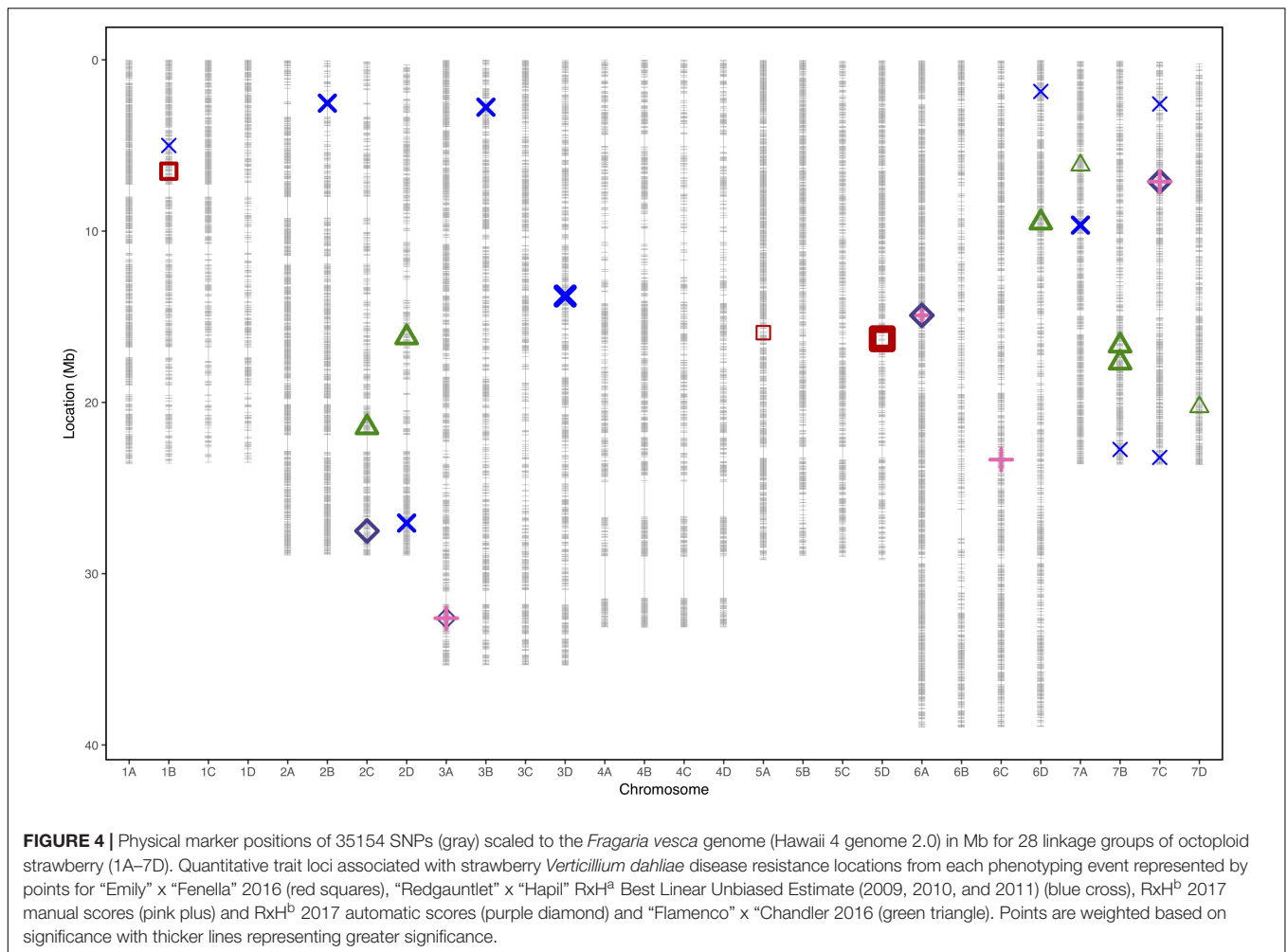
Disease resistance markers were identified and validated as outlined in Cockerton et al. (2018b). Furthermore, inference of whether resistant markers were present across multiple populations and the targeted marker association study was conducted as outlined in Cockerton et al. (2018b). Candidate resistance genes were identified in the *Fragaria vesca* genome (assembly v1.1; Shulaev et al., 2011) and screened for the presence of NB-LRR, TM-CC, RLP, RLK (S-type and general) domains and candidate Rosaceous MLO genes (Pessina et al., 2014) following published pipelines (Li et al., 2016). Resistance genes were identified within 100 kb of the significant resistance marker using BEDtools (Quinlan and Hall, 2010). Characterisation of homologous genes in the NCBI database was undertaken

through tblastx (Karlin and Altschul, 1993). NB-ARC domains were identified from *F. vesca ab initio* and hybrid gene models using InterProScan (Quevillon et al., 2005). Significant association of NBS and NB-ARC domains with focal markers was tested through assessing their occurrence within 100 kb of 25 randomly sampled markers from across the four populations over 10,000 permutations.

## RESULTS

### Resistance to Isolates Varies Between Cultivars

The cultivar “Redgauntlet” exhibits tolerance to the subclade II-1 isolate, 12158 and moderate tolerance to the subclade II-2 isolate 12008, however, the cultivar “Hapil” is susceptible to



the isolates from both subclades (**Supplementary Figure S1**). The cultivars “Fenella,” “Flamenco,” and “Chandler” are highly tolerant to the *V. dahliae* subclade II-2 isolate whereas “Emily” is highly susceptible (**Supplementary Figure S2**).

### The ‘Flamenco’ x ‘Chandler’ Linkage Map

The newly generated FxC linkage map (**Supplementary Tables S1, S2**) has an average genetic distance between markers of 0.3 cM which is a lower average gap than ExF and RxH<sup>a</sup> (Cockerton et al., 2018b), however, there are 10 gaps greater than 20 cM and linkage groups 2C, 3C, 6C, and 6D each resolved into two linkage groups. FxC linkage information was used as one of the five populations to construct the consensus map. All reported marker positions listed in this study are based on the position in the consensus map.

### Comparison of Automated and Manual Phenotyping Methods

The proportion of diseased: healthy leaf area and the NDVI values were assessed at discrete time points and over time (**Figure 1**). A strong negative correlation was observed between the manual

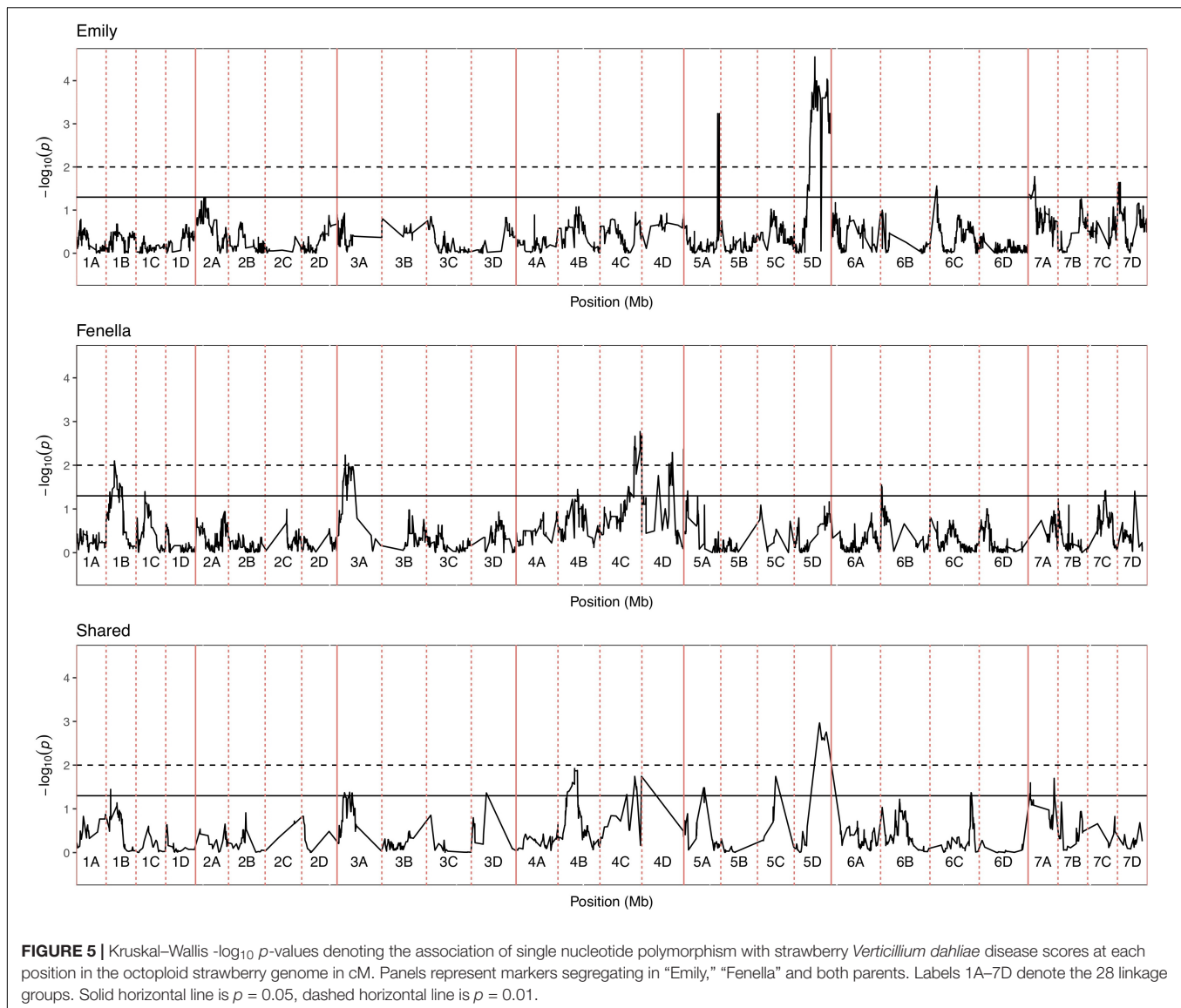
disease scores and diseased: healthy leaf area (**Figure 2**), with a stronger relationship observed between the manual disease score and NDVI for validation and RxH<sup>b</sup> phenotyping events August 2, 2017 ( $r = 0.78$ ,  $p < 0.001$ ) and September 13, 2017 ( $r = 0.78$ ,  $p < 0.001$ ) and also between the manual AUDPC and NDVI-AUDPC ( $r = 0.85$ ,  $p > 0.001$ ; **Figure 2**).

Significance values for focal SNPs predicted by either automated and manual phenotyping follow the same patterns across the strawberry genome (**Figure 3**) and three out of four resistance markers were successfully identified in the automated phenotyping QTL analysis (**Figure 4**).

### QTL Mapping in Four F1 Full-Sib Mapping Populations

In total, four populations were assessed for resistance to *V. dahliae*. Twenty-five focal markers for *V. dahliae* resistance were identified in the RxH<sup>a</sup>, RxH<sup>b</sup>, ExF, and FxC populations of strawberry (**Figures 3–7** and **Table 1**). Twelve of these focal markers were considered to have a moderate effect with greater than 10 percent impact on disease score across the population. When comparing the observed versus expected disease scores the coefficients of determination, the focal markers explain between





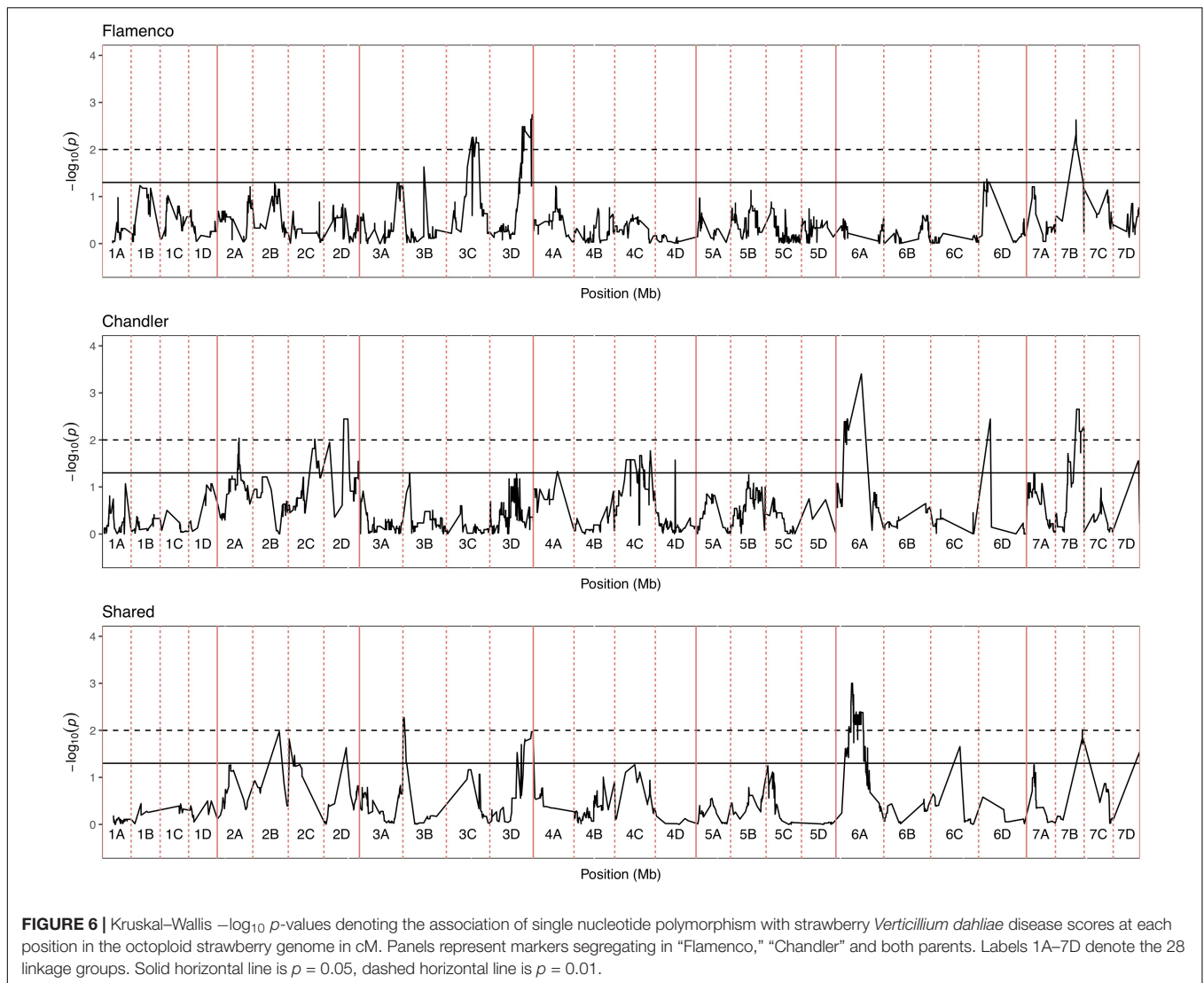
25 and 68% of the observed mean disease scores between progeny members (Table 2).

### Resistance Markers Found Close to Neighboring Resistance Genes

A total of 14 out of 25 markers identified were located within 100 kb of a putative resistance gene found in *F. vesca* (Table 1), each of which indicating a potential target for further study. Twelve resistance markers were found to be within 100 kb of a putative resistance gene containing a NBS. NBS containing genes were more frequently associated with resistance focal SNPs than markers selected at random ( $p = 0.0073$ – $0.0017$ ,  $n = 10,000$ ; Supplementary Figure S3). Nine of the twelve NBS genes contained an NB-ARC domain, however, NB-ARC domain containing genes were not more frequently associated with resistance focal SNPs than markers selected at random ( $p = 0.066$ – $0.024$ ,  $n = 10,000$ ; Supplementary Figure S4).

### Improved QTL Identification With SNP Data

Newly generated SNP data has allowed the analysis of the RxH<sup>a</sup> mapping population wilt phenotypic data reported in Antanavičiute et al. (2015). Our new analyses identified four different loci represented by SNP markers which are located on the same linkage groups as four of the QTL previously reported using the SSR marker analysis (Antanavičiute et al., 2015) and six novel resistance QTL. The original SSR markers associated with wilt resistance all mapped to the same chromosome as originally reported, however, different sub-genomes were assigned when following the linkage group nomenclature stipulated by van Dijk et al. (2014) and Sargent et al. (2015). The QTL *RVd1* maps to linkage group 3B and is located 2.7 Mb from the *FaRVd3B* SNP marker. *RVd3* maps to linkage group 7A and is 6.4 Mb from *FaVd7A2*. *RVd7* maps to linkage group 2D and is 1.2 Mb away from



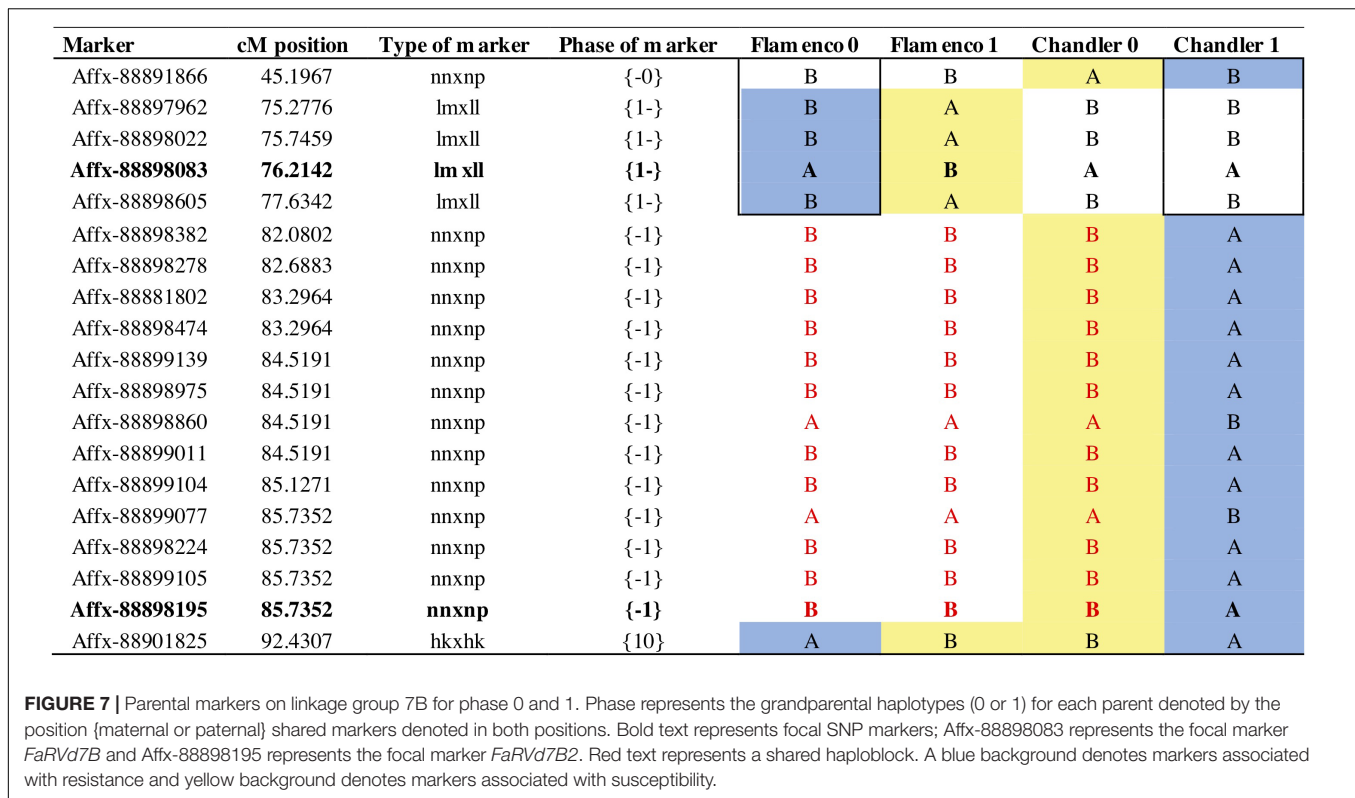
*FaRVd2D2*. *RVd4-M1* mapped to linkage group 2B, however, it is not considered to represent the same QTL as *FaRVd2B* as it was mapped 12.6 Mb away. The RxH<sup>a</sup> SNP strawberry map has a greater density of segregating loci (3451) than the SSR map (1133) therefore the SNP data allows greater accuracy of QTL mapping which, when combined with the consensus map, assists the comparison of alleles to other phenotyped populations. Discrepancies between the two analyses can be explained by the removal of 13 rogue individuals, the use of AUDPC phenotyping measure and BLUE calculated across multiple years of phenotyping, in comparison the original analysis used data from the single most heritable scoring event for each year.

## Overlap of Resistance Markers Between Cultivars

The resistance marker *FaRVd7B* identified in ‘Flamenco’ and *FaRVd7B2* in ‘Chandler’ are 9.5 cM and 1 Mb apart, however,

the analysis of haploblocks revealed that these two markers represent discrete resistance loci present on different haplotypes (Figure 7). Shared markers indicate that the resistance marker from “Flamenco” and “Chandler” contribute to resistance in an additive fashion. The haploblock representing the marker *FaRVd7B* is associated with resistance in “Flamenco” is also present in “Chandler,” however, it is associated with susceptibility. This low transferability indicates that marker tagging the resistance haplotype *FaRVd7B* is not in linkage disequilibrium with the resistance gene.

The resistance markers *FaRVd1B* and *FaRVd1B2* from population RxH<sup>a</sup> and ExF, respectively, are positioned 1.5 Mb apart on linkage group 1B (Table 1). Comparison of haploblocks across the two-populations allowed us to determine whether the two markers represent the same resistance allele. Although the focal markers are reciprocally monomorphic, analysis of shared polymorphic neighboring markers indicated that the resistant markers are present on different haplotypes and therefore represent discrete resistance loci.



No overlap in markers was observed between resistance loci identified between the RxH<sup>a</sup> combined analysis and the RxH<sup>b</sup> populations screened with mixed inoculum and subclade II-2 inoculum, respectively (Table 1). By contrast, the marker Affx-88837276 identified on linkage group 7C was 0.72 Mb from the focal marker identified in the RxH<sup>a</sup> 2011 phenotyping event indicating the possibility of an overlapping QTL associated with resistance to different isolates.

The aforementioned co-dominant shared marker, *FaRVd3D*, was identified in both “Redgauntlet” and “Hapil” cultivars; this is a shared resistance QTL between the two cultivars. The shared marker *FaRVd5A* identified a resistance allele present in both “Emily” and “Fenella” cultivars whereby homozygous genotypes containing two resistance alleles are required to observe significant levels of resistance.

## Validation of Two Resistance Markers Across the Wider Germplasm

Identification of substitute i35k SNPs co-localizing with focal SNPs identified in the i90k F1 mapping population analysis allowed focal markers to be screened across the wider germplasm (Supplementary Figure S5 and Supplementary Table S3). Two of the focal SNPs identified in the F1 cross studies maintained a strong association with resistance across the validation accessions. *FaRVd2B* identified in “Redgauntlet” ( $X^2_{(4,5;1)} = 5.72$ ;  $p = 0.017$ ) and *FaRVd6D* identified in “Chandler” ( $X^2_{(4,6;2)} = 7.47$ ;  $p = 0.024$ ) explained 13.4 and 2.2% of the variation in disease scores observed in the validation germplasm, respectively.

## DISCUSSION

### Description of Resistance

Multiple sources of resistance to Verticillium wilt were observed across six strawberry cultivars indicating a wealth of genetic resources that can be exploited by breeders. Similar studies have also found multiple sources of resistance to *V. dahliae* in strawberry, and all alleles were found to be dominant (Govorova and Govorov, 1997). We observe that the dominantly inherited allele *FaRV5D* (Affx-88867578) is associated with an increase in susceptibility (Figure 8). It is impossible to know from this work whether *FaRV5D* is a susceptibility factor or whether it represents an additive resistance allele which is in repulsion to the identified marker. Further work, through selfing “Emily,” or through crossing heterozygous ExF progeny would generate the missing homozygous class and reveal the inheritance of this resistance allele more clearly. Should *FaRV5D* be found to represent a recessive resistance allele it could prove a valuable tool for strawberry breeders. Resistant homologs of susceptibility factors have been shown to be highly robust, with exploitation lasting for 50 years in the field (Kang et al., 2005; Pavan et al., 2010), therefore utilization of such a resistance incidence could prove a highly robust strategy to prevent against Verticillium infection.

A resistance QTL was identified in both “Redgauntlet” and “Hapil” cultivars on linkage group 3D. Analysis of parental and shared markers in this region indicated resistance alleles from both parents co-localized to the same location and thus represent the same QTL. This QTL was termed *FaRVd3D* and could be best represented by haploblocks phase 1 “Hapil” and phase 0



**TABLE 1** | Focal SNPs linked with each quantitative trait loci associated with strawberry *Verticillium dahliae* disease resistance identified through the Kruskal–Wallis analysis.

QTL Name	Linkage Group	Closest SNP_i90k	Position (Mb)	P-value	k	Sig	Parent	Percentage effect	Closest R/S gene (bp)	Type of gene	Gene name	Number R/S genes 100 kb	Population
<i>FaRVd5D</i>	5D	Affx-88867578	16.3	$2.8 \times 10^{-5}$	17.5	****	Emily	17.2	NA	NA	NA	0	ExF
<i>FaRVd1B</i>	1B	Affx-88814091	6.5	$8.0 \times 10^{-3}$	7.0	**	Fenella	10.9	NA	NA	NA	0	ExF
<i>FaRVd5A</i>	5A	Affx-88865854	1.6	$3.3 \times 10^{-2}$	8.7	*	Emily and Fenella	-10.1	NA	NA	NA	0	ExF
<i>FaRVd7B</i>	7B	Affx-88898083	16.7	$2.4 \times 10^{-3}$	9.3	**	Flamenco	9.3	9853	NBS	maker-LG7-exonerate_protein2genome-gene-152.135-mRNA-1	6	FxC
<i>FaRVd2C</i>	2C	Affx-88826960	21.5	$3.3 \times 10^{-3}$	6.7	**	Chandler	10.9	34428	NBS	maker-LG2-augustus-gene-170.171-mRNA-1	3	FxC
<i>FaRVd2D</i>	2D	Affx-88821311	16.2	$3.3 \times 10^{-3}$	8.5	**	Chandler	9.5	NA	NA	NA	0	FxC
<i>FaRVd7B2</i>	7B	Affx-88898195	17.7	$3.3 \times 10^{-3}$	9.4	**	Chandler	-9.8	25189	RLK	mrna26334.1-v1.0-hybrid	1	FxC
<i>FaRVd7A</i>	7A	Affx-88894946	6.2	$3.3 \times 10^{-2}$	3.8	*	Chandler	9.4	16557	NBS	maker-LG7-exonerate_protein2genome-gene-88.76-mRNA-1	2	FxC
<i>FaRVd6D</i>	6D	Affx-88877237	9.5	$3.3 \times 10^{-3}$	8.5	**	Chandler	-13.2	9934	NBS	mrna09586.1-v1.0-hybrid	3	FxC
<i>FaRVd7D</i>	7D	Affx-88900355	20.3	$3.3 \times 10^{-2}$	4.8	*	Chandler	10.7	8101	NBS	mrna34123.1-v1.0-hybrid	1	FxC
<i>FaRVd6A</i>	6A	Affx-88902601	14.9	$3.3 \times 10^{-3}$	7.8	**	Redgauntlet	15.2 (-5.0)	NA	NA	NA	0	RxH <sup>b</sup> Manual and Automated
<i>FaRVd2C2</i>	2C	Affx-88828094	27.5	$4.6 \times 10^{-3}$	8.0	**	Redgauntlet	-15.5	13390	NBS	genemark-LG2-processed-gene-188.73-mRNA-1	2	RxH <sup>b</sup> Manual

(Continued)

TABLE 1 | Continued

QTL Name	Linkage Group	Closest SNP_i90k	Position (Mb)	P-value	k	Sig	Parent	Percentage effect	Closest R/S gene (bp)	Type of gene	Gene name	Number R/S genes 100 kb	Population
<i>FaRVd3A</i>	3A	Affx-88845095	32.6	$1.9 \times 10^{-2}$	5.5	*	Redgauntlet	13.1 (-5.0)	NA	NA	NA	0	RxH <sup>b</sup> Manual and Automated
<i>FaRVd7C</i>	7C	Affx-88895402	7.1	$3.3 \times 10^{-3}$	8.6	**	Redgauntlet	-14.7 (5.9)	40246	NBS	maker-LG7-augustus-gene-98.128-mRNA-1	1	RxH <sup>b</sup> Manual and Automated
<i>FaRVd6C</i>	6C	Affx-88883592	23.3	$4.9 \times 10^{-3}$	7.9	**	Redgauntlet	(-5.3)	NA	NA	NA	0	RxH <sup>b</sup> Automated
<i>FaRVd6D2</i>	6D	Affx-88818910	1.8	$1.1 \times 10^{-2}$	6.5	*	Redgauntlet	6.0	13492*	NBS*	maker-LG1-snap-gene-197.206-mRNA-1*	2*	RxH <sup>a</sup>
<i>FaRVd2B</i>	2B	Affx-88822931	2.5	$2.7 \times 10^{-3}$	9.0	**	Redgauntlet	6.6	NA	NA	NA	0	RxH <sup>a</sup>
<i>FaRVd2D2</i>	2D	Affx-88828415	27.0	$4.2 \times 10^{-3}$	8.2	**	Redgauntlet	-10.0	NA	NA	NA	0	RxH <sup>a</sup>
<i>FaRVd3D</i>	3D	Affx-88836863	13.8	$8.8 \times 10^{-4}$	16.5	***	Redgauntlet and Hapil	-18.3-both alleles	82711	MLO-homolog	mrna31264.1-v1.0-hybrid	16 (S)	RxH <sup>a</sup>
<i>FaRVd3B</i>	3B	Affx-88833107	2.8	$7.9 \times 10^{-3}$	7.1	**	Redgauntlet	8.4	31622	NBS	genemark-LG3-processed-gene-23.50-mRNA-1	2	RxH <sup>a</sup>
<i>FaRVd7B3</i>	7B	Affx-88900983	22.7	$1.8 \times 10^{-2}$	5.6	*	Redgauntlet	7.2	26652	NBS	maker-LG7-augustus-gene-201.144-mRNA-1	4	RxH <sup>a</sup>
<i>FaRVd7C2</i>	7C	Affx-88900732	23.2	$3.6 \times 10^{-2}$	4.4	*	Redgauntlet	6.2	17146	NBS	mrna12407.1-v1.0-hybrid	2	RxH <sup>a</sup>
<i>FaRVd1B2</i>	1B	Affx-88813017	5.0	$3.4 \times 10^{-2}$	4.5	*	Hapil	-9.4	NA	NA	NA	0	RxH <sup>a</sup>
<i>FaRVd7C4</i>	7C	Affx-88894332	2.6	$3.4 \times 10^{-2}$	4.5	*	Hapil	-7.7	NA	NA	NA	0	RxH <sup>a</sup>
<i>FaRVd7A2</i>	7A	Affx-88895680	9.7	$4.2 \times 10^{-3}$	8.2	**	Hapil	5.8	52659	NBS	mrna04795.1-v1.0-hybrid	2	RxH <sup>a</sup>

Closest resistance gene reported within 100 kb if applicable. Gene name represents the nearest putative resistance (R) or susceptible (S) gene within 100 kb of the SNP based on v1.0 *Fragaria vesca* gene models.

**TABLE 2** | Model parameters for the predictive linear model for each phenotyping event.

Mapping population	Year	Isolate	R <sup>2</sup>	df	F-value	p-value	RSE	H <sup>2</sup>	Markers in Model
ExF	2016	12008	0.25	2,129	14.17	4.8 × 10 <sup>-8</sup>	69.48	0.08	Affx-88865854, Affx-88867578, Affx-88814091
FxC	2016	12008	0.52	7,71	10.97	2.6 × 10 <sup>-9</sup>	63.50	0.18	Affx-88898083, Affx-88826960, Affx-88821311, Affx-88898195, Affx-88894946, Affx-88877237, Affx-88900355
RxH <sup>a</sup>	2009	Mixed	0.69	15,110	16.25	2.2 × 10 <sup>-16</sup>	103.80	0.45	Affx.88852211, Affx.88818910, Affx.88833107, Affx.88894332, Affx.88866219, Affx.88848759, Affx.88878850, Affx.88836872, Affx.88822931, Affx.88818695, Affx.88812713, Affx.88857747, Affx.88895680, Affx.88868119, Affx.88850610
RxH <sup>a</sup>	2010	Mixed	0.46	9,111	51.61	1.4 × 10 <sup>-11</sup>	51.61	0.17	Affx.88872845, Affx.88855744, Affx.88824293, Affx.88839291, Affx.88822931, Affx.88877823, Affx.88848601, Affx.88851418, Affx.88900407
RxH <sup>a</sup>	2011	Mixed	0.48	8,118	13.64	7.4 × 10 <sup>-14</sup>	86.56	0.27	Affx.88868119, Affx.88837276, Affx.88897702, Affx.88874994, Affx.88876011, Affx.88848763, Affx.88856954, Affx.88901109
RxH <sup>a</sup>	BLUE	Mixed	0.53	11,127	13.1	2.3 × 10 <sup>-16</sup>	79.13	NA	Affx-88818910, Affx-88822931, Affx-88828415, Affx-88836863, Affx-88833107, Affx-88900983, Affx-88900732, Affx-88813017, Affx-88894332, Affx-88895680
RxH <sup>b</sup>	2017	12158	0.43	4,58	11.15	8.9 × 10 <sup>-7</sup>	50.70	0.10	Affx-88828094, Affx-88828094, Affx-88845095, Affx-88895402
RxH <sup>b</sup> (A)	2017	12158	0.42	4,58	10.58	1.7 × 10 <sup>-6</sup>	389.80	0.12	Affx-88902601, Affx-88845095, Affx-88895402, Affx-88883592

Predicted versus observed disease scores or each genotype within the population. R<sup>2</sup> is the coefficient of determination, df are the degrees of freedom associated with the F statistic, the numerator is associated with model parameter number. H<sup>2</sup> is the broad sense heritability. (A) denotes the automated disease assessment.

“Redgauntlet.” Both alleles are required in order to observe the greatest combined resistance effect thus indicating that this QTL was inherited in a co-dominant fashion.

Genotypes exhibit large variation in the disease response when compared to variation across genotypes. The variation is represented by large standard error values (**Supplementary Figure S2**) and the corresponding low broad sense heritability values (**Table 2**). The high correlation between automated and manual phenotyping values, validates the manual phenotypic scores. We can therefore conclude that the large variation associated with disease score reflects the truly variable nature of the *Verticillium* disease responses in strawberry. This within-genotype variation has been observed previously and as such, high replication of genotypes in *Verticillium* trials ( $n = 10$ ) mitigates this large variation and results in a greater phenotyping accuracy.

## Transgressive Segregation

Transgressive segregation toward susceptibility was observed in the FxC population (**Supplementary Figure S2**) with 7.5% of progeny exhibiting a significantly higher disease symptoms than that of the parents. The parental cultivars “Flamenco” and

“Chandler” are related, namely “Chandler” is the grandparent of “Flamenco.” Reports that inbreeding results in increased susceptibility to plant diseases (Watt et al., 2013) alongside negative implications on other traits (Maas and Galleta, 1996) support the observation of increased susceptibility after crossing two related parents. Transgressive segregation toward susceptibility indicates that the two parental lines contain different resistance alleles (Geiger and Heun, 1989), indeed we do not identify any shared markers or loci between the two parents, however, only one significant marker was identified in “Flamenco.” Nonetheless, this cross allowed the identification of a number of focal SNPs for further investigation.

## Limitations of the i35k Phenotyping Platform

Phenotypic data was re-analyzed using the subset of i90k markers represented on the i35 SNP chip. The subset of i35k markers were associated with a slightly reduced power to detect resistance markers (**Supplementary Figure S5**). A surrogate i35k SNP marker could not be elucidated for *FaRVd6D*. The i35k focal SNP representing the *FaRVd5D* shifted 9.2 Mb and *FaRVd1B2* had shifted 2.0 Mb. However, the remaining focal



Marker Name	cM	Type	Phase	Emily (0)	Emily (1)	Fenella (0)	Fenella (1)	$p$	$k$	sig	ll	lm	hh	hk	kh	kk
Affx-88863919	68.6	<hkxhk>	{00}	B	A	B	A	0.25	4.07	*	NA	NA	292.3	295.9	264.4	262.8
Affx-88867578	90.1	<lmxl>	{1-}	B	A	B	B	0.00	17.54	****	248.5	308.2	NA	NA	NA	NA
Affx-88869286	100.0	<hkxhk>	{10}	A	B	B	A	0.00	16.09	**	NA	NA	245.5	255.2	300.3	311.0

**FIGURE 8 |** Phasing and marker effect sizes for the *FaRVd5D* focal SNP and neighboring shared markers. Parental phased haploblocks for linkage group 5D represented in “Emily 0”, “Emily 1”, “Fenella 0” and “Fenella 1” columns, Red haplotype associated with susceptibility. Grand phenotype means for each marker class represented under marker classes denoted “ll”, “lm”, “hh”, “hk”, “kh” and “kk”; ll/lm represents markers that segregate in the maternal parent, hh, hk, kh and kk represents markers that segregates in both parents. cM – centimorgan distance along linkage group 5D,  $p$  – probability for the  $k$  – Kruskal–Wallace test statistic testing differences between marker classes.

markers were detected within 0.8 Mb or less of the i90k focal SNP. The validation set of 92 cultivars was phenotyped using the streamlined i35k SNP chip. A targeted marker association analysis using the validation phenotyping event did not pull out any resistance markers, typically genome wide analysis requires a greater genotype number. Either a greater density of markers or a greater number of genotypes may allow the identification of resistance QTL present across the wider germplasm.

## Environmental Factors

Variation in weather conditions can lead to variation in disease severity between years (Talboys and Bennett, 1969), the variation in disease development may explain differences observed in RxH<sup>a</sup> phenotyping events (**Supplementary Figures S6, S7**) (Keyworth and Bennett, 1951). Where variation in disease susceptibility was observed in cultivars of “Earliglow”, “Howard 17” and “Bounty” across different publications (Vining et al., 2015) this may be due to environmental variation or variation in the isolates subclade used for inoculations. Pre-inoculation of strawberry with low virulence *V. dahliae* isolates has been shown impact the virulence of pathogenic *V. dahliae* strawberry isolates (Diehl et al., 2013). Thus, the mix of isolates for RxH<sup>a</sup> trials and the presence of existing *V. dahliae* microsclerotia at trial sites may have influenced disease expression.

## NBS Genes May Contribute to Verticillium Resistance

A high proportion of the resistance focal SNPs were associated with NBS resistance genes indicating that NB-LRR mediated signaling may play a role in strawberry Verticillium resistance. NBS genes have been implicated in Verticillium resistance in other host systems. Seven TIR-NBS-LRR resistance genes were observed to be up-regulated in *Arabidopsis thaliana* 24 h after Verticillium co-culturing (Scholz et al., 2018) again indicating NBS-LRRs may play a role in Verticillium resistance. The NBS resistance gene *GbaNA1* was found to control disease resistance to Verticillium in cotton and also confer resistance when transformed into *A. thaliana* (Li et al., 2018a,b). A positive correlation between the number of Verticillium and Fusarium wilt resistance QTL and NBS genes was observed on subgenome A of Cotton (Zhang et al., 2015). The most convincing evidence for the existence of Verticillium specific nuclear interactions can be observed through the pathogen effector *VdSCP7* which was found to localize at the host nucleus and modulate effector triggered immunity in cotton (Zhang et al., 2017). Of the 12 identified NBS resistance genes, nine were

found to contain a NB-ARC domain. NB-ARCs have been demonstrated to trigger HR leading to localized plant cell death and thus containment of the pathogen (Hammond-Kosack and Jones, 1996; van der Biezen and Jones, 1998). HR occurs in response to pathogen derived molecules (Avr genes) with trigger specificity controlled by LRR domains of the resistance gene (van der Biezen and Jones, 1998). A high frequency of NB-ARC association with Verticillium resistance focal SNPs suggests that HR may play a large role in *V. dahliae* resistance response of strawberry. HR resistance is typically considered to be race specific and also have a lower durability within the field (Lindhout, 2002). Previous studies have highlighted the importance of the HR in roots: *Phytophthora sojae* resistance was partially induced in soybean through the use of lesion mutant lines which triggered root cell death in response to pathogen invasion (Kosslak et al., 1996). This also resulted in a trade-off where lesion mutants exhibited an inability to form symbiotic nodules with nitrogen-fixing bacteria (Kosslak et al., 1996). Further evidence that HR may be an important factor of a resistance response to *V. dahliae* infection can be seen where the effector *PevD1* identified in *V. dahliae* isolated from cotton resulted in HR when infiltrated onto tobacco (Wang et al., 2012) and similarly with the Verticillium effector *Ave1* in tobacco (Fan et al., 2018). Of particular interest was the marker *FaRVd7B3* where the closest resistance gene shows 90% identity to 47% of the resistance gene *muRdr1* controlling gene-for-gene specific resistance to *Diplocarpon rosae* a foliar fungal disease in tetraploid rose (*Rosa multiflora*) (Terefe-Ayana et al., 2011).

## Automated Phenotyping as a Tool for Breeders

The UAV and imaging have allowed the development of a high-throughput phenotyping system to assess the disease resistance status of plants. A substantial labor-saving cost could be achieved through implementation of the phenotyping platform as the manual assessment of 2500 plants five times over the season took a total of 37.5 h. In contrast switching to a UAV-based phenotyping approach cut the time down to 2.5 h. There was a strong association between the manual disease scores (AUDPC) and the automated disease scores (NDVI-AUDPC) of *V. dahliae* inoculated plants. Furthermore, the use of automated phenotypic scores resulted in successful identification of resistance markers. In a similar study NDVI was found to be a good measure for Verticillium wilt structural damage in olive (Calderón et al., 2013), which suggests the transferability of this NDVI disease

score across different crop hosts. In future work, the semi-automated image analysis will be improved to fully automated canopy segmentation.

## Deploying the Identified Resistance

Most of the alleles identified in this study are of moderate effect with two out of 25 consistently performing over the wider germplasm. Studying the *Verticillium* resistance present within pertinent cultivars related to breeding populations will ensure greater relevance of future resistance markers. In the absence of robust markers associated with the moderate resistance incidences seen here, and in the complete absence of major single gene resistance, we believe that genomic selection may provide a better strategy to breed *Verticillium* disease resistance into strawberry. Nonetheless, recent advances in strawberry research including recent advances in genome sequencing (unpublished observation) and successful CRISPR/CAS9 transformation (Wilson et al., 2018), could be used to identify putative resistance genes and allow functional characterisation, respectively. These tools may allow the development of robust functional markers which perfectly tag the causative resistance genes associated with *FaRVd5D* and *FaRVd3D*.

## CONCLUSION

Marker-assisted breeding and more likely genomic selection will result in a higher probability of developing a successful cultivar containing *Verticillium* wilt resistance and provides plant breeders with a competitive advantage in comparison to those implementing empirical breeding strategies. Here we report multiple loci of interest for breeders, two of which are associated with resistance across the wider strawberry germplasm. Furthermore, we highlight the potential for a HR resistance mechanism to play a large role in resistance to *Verticillium* in strawberry. The automated phenotyping platform could provide a valuable tool for breeders and pre-breeding research work.

## AUTHOR CONTRIBUTIONS

HC, DS, CE, and RH conceived and designed the experiments. HC, CM-M, and AG-C performed all pathogenicity tests. RV analyzed SNP data and made linkage map. AP propagated plant material. HC analyzed pathogen data and conducted quantitative genetics analysis. BL analyzed imaging data. DWS provided plant material and phenotyping advice. AA provided gene annotations and bioinformatics support. HC, BL, and RH wrote the manuscript with contributions from all authors.

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## SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fpls.2019.00924/full#supplementary-material>

**FIGURE S1** | Area under the disease progression curve for “Hapil,” “Redgauntlet,” and “Chandler” cultivars. Light gray bars represent plants inoculated with *Verticillium dahliae* isolate 12008 from subclade II-2, white bars 12158 from subclade II-1 and dark gray represents mock inoculated plants.

**FIGURE S2** | Area under the disease progression curve for each genotype from the “Flamenco” x “Chandler” and “Emily” x “Fenella” populations.

**FIGURE S3** | Permutation test results showing the frequency of twenty-five randomly sampled markers, from those present in the F1 populations, that fell within 100 kb of a NBS containing resistance gene over 10,000 iterations. The red vertical line represents the number of markers within 100 kb of an NBS observed in three F1 populations.

**FIGURE S4** | Permutation test results showing the frequency of twenty-five randomly sampled markers, from those present in the F1 populations, that fell within 100 kb of an NB-ARC containing resistance gene over 10,000 iterations. The red vertical line represents the number of markers within 100 kb of an NB-ARC observed in three F1 populations.

**FIGURE S5** | Physical marker positions of SNP markers (gray) scaled to the *Fragaria vesca* genome (Hawaii 4 version 2.0) in Mb for 28 linkage groups of octoploid strawberry (1A–7D) marker positions scaled to *F. vesca* genome. Resistance marker locations from the IStraw90 Affymetrix chip (red; +) and IStraw35 Affymetrix chip validation SNPs (green; ◊). Overlap of symbols indicates focal SNPs identified at the same location.

**FIGURE S6** | Kruskal–Wallis  $-\log_{10} p$ -values denoting the association of SNPs with strawberry *Verticillium dahliae* disease scores at each position in the octoploid strawberry genome in cM. Panels represent markers segregating in “Redgauntlet,” “Hapil” and both parents. Labels 1A–7D denote the 28 linkage groups. Solid horizontal line is  $p = 0.05$ , dashed horizontal line is  $p = 0.01$ . Color denotes phenotyping event blue- 2009, lime- 2010, green- 2011, orange- 2017.

**FIGURE S7** | Relative Area Under the Disease Progression Curve (AUDPC) for each of the seven phenotyping events illustrating the phenotypic range of disease symptoms. “Emily” x “Fenella” (ExF), “Flamenco” x “Chandler” (FxC), Validation set (Val), “Redgauntlet” x “Hapil” in population 1 (RxH<sup>a</sup>) over 3 years (2009, 2010, and 2011), and population 2 (RxH<sup>b</sup>). A relative AUDPC of 80 indicates an a-symptomatic plant and 900 indicates a plant that died at the first timepoint.

**TABLE S1** | Genetic map for “Flamenco” x “Chandler” F1 population for Axiom® IStraw90 markers generated using Crosslink (Vickerstaff and Harrison, 2017).

**TABLE S2** | “Flamenco” x “Chandler” Axiom® IStraw90 marker order based upon *Fragaria vesca* Hawaii 4 genome version 2.0 22.

**TABLE S3** | List of individuals in the validation set.

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**Conflict of Interest Statement:** DS and CE were employed by company Driscoll's Genetics Ltd.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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