

Supplementary Information for

Arabidopsis cell wall composition determines disease resistance specificity and fitness

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Other supplementary materials for this manuscript include the following:

Dataset S1 (glycomics data) Dataset S2 (Phyton scripts for CRT)

SI *Appendix*

Supplementary Material and Methods

Determination of Ca+2 burst upon treatment with cell wall fractions

Eight-day-old liquid-grown apoaequorin-expressing seedlings (Col-0^{AEQ} and *agb1-2AEQ*) were used for Ca²⁺_{cyt} upon cell wall fractions treatment. Col-0^{AEQ} and *agb1-2*^{AEQ} plants (were grown in 24-well plates (~10 seedlings per well) under long day conditions (16 h of light) at 20-22ºC in liquid MS medium. Then, they were placed individually in 96-well plates in coelenterazine (PJK GmbH) and water and were incubated overnight in the dark. Luminescence was recorded with a Varioskan Flash Multimode Reader (Thermo Scientific) as described, after treatment with cell wall extracts (PEC1 and PEC2) from wild-type plants and cell wall mutants (1). *agb1-2^{AEQ}* lines were generated by crossing *agb1-2* with Col-0^{AEQ} plants, and homozygous *agb1-2*AEQ plants were selected by measuring the luminescence emitted after the addition of 2 M CaCl₂ to leaf discs from three-week-old plants and by allelespecific PCR amplification to confirm *agb1-2* background (2, 3).

Immunoblot analysis of MAPK activation

Twelve-day-old seedlings (n = 10) grown on liquid MS medium in 24-well plates were treated with water (mock) or PEC1 fractions (heat-treated at 121 ºC for 20 min; 50 ng/µl) from wildtype plants (Col-0) or cell wall mutants. Seedlings were harvested in liquid nitrogen at 0 (before treatment), 10 and 20 minutes. Protein extraction and detection of activated MAPKs using the Phospho-p44/42 MAPK (Erk1/2) (Thr202/Tyr204) antibody (Cell Signaling Technology) were performed as described (1). Phosphorylated MAPKs bands were detected with the iBright CL1000/FL1000 imaging system (Thermo Fisher Scientific) and MAPKs bands intensity was determined using the software of the equipment and following manufacturer instructions.

Determination of Ca+2 in cell wall fractions

Total free $Ca²⁺$ in PEC1 and PEC2 cell wall fractions was determined by atomic absorption spectrophotometry (4) in the Analytical Chemistry Unit of the Escuela Técnica Superior de Ingeniería Agronómica, Alimentaria y de Biosistemas (Universidad Politécnica de Madrid). Equipment was calibrated with 1, 5 and 3 ppm standards. PEC1 and PEC2 fractions were diluted to 1 mg fraction/mL and centrifuged for 5 minutes at full speed. Aqueous supernatants were directly measured, whereas pellets containing materials in suspension total $Ca²⁺$ were first acid-digested with 14 cm³/g 15 M HNO₃ and 6 cm³/g 27 M HCl followed by a digestion at 120 ºC for 2 h in Teflon bombs in an oven.

Plant tolerance to desiccation assays

Three-week-old soil-grown plants (n=8) were restricted completely from irrigation for 21 days and then wilted plants were re-watered for 7 days and the number of plants that survived to the stress and recovered their developmental phenotype was scored. These experiments were performed four times.

Mathematical Modelling

The description of the mathematical analyses performed is shown in the schema of *SI Appendix*, Fig. S16, and can be divided into two interrelated tasks that share a set of analytical steps. These common analyses were initiated with the experimental data on susceptibility to pathogens, fitness and tolerance to desiccation of *cwm* and wild-type plants (Fig. 1 and Fig. 2A, B, and *SI Appendix* Fig. S3 and Fig. S7A). For each of these 6 response variables (resistance to 3 pathogens, 2 fitness parameters and plant desiccation tolerance) a different two-way ANOVA model was fit for each ecotype (Col-0, Ws-0, or La-*er*), considering the genotype as the factor/effect of interest, but also introducing a second factor/effect to allow collecting together the data from a number of similar experiments (2 independent experiments in the case of pathogen resistance, 3 for fitness and 3 for tolerance to desiccation). No interaction was initially considered between both factors/effects, rendering the experiment effect as a block. These initial analyses aim to confirm the known significance of the genotype factor/effect, and to estimate the least squares means (or LS means) of each response in each genotype. These LS means provide a single estimation of the average response level (e.g. mean disease rating for both *Pc* and *Rp*, mean conidiospores/mg fresh weight for *Ha*, mean seed yield in mg/plant, mean rosette fresh weight/biomass in g, and mean survival rate (%) after desiccation) for each genotype under consideration, controlling the effect of the experimental differences due to the unbalanced nature of the designs.

The correlation analysis between mean biotic stress resistance, on one side, and mean fitness and desiccation tolerance, on the other side, was then performed on this intermediate set of 21 genotypes. To this aim, the percentage ratio of each genotype LS mean to that of the corresponding ecotype wt was obtained for each response variable. These susceptibility ratios allowed expressing the average response information of each genotype in relation to that of the reference wt, as well as bringing the information of different ecotypes to a similar scale. A linear model was then fit for each combination of the logtransformed biotic susceptibility ratios with the fitness and abiotic susceptibility ratios to analyze their correlations (see Fig. 2 C, D and *SI Appendix*, Fig. S7; for the fitted equations, *R*-squares and *p*-values; *SI Appendix*, Figure S16). As a consequence of the logarithmic transformation of the biotic susceptibility ratios, the *x*-axes in these figures range between 0 (smaller susceptibility) to 5 (greater susceptibility). The SAS software has been used to implement the previous analyses (*glm* and *corr* procedures).

Next, a paired comparisons analysis on the mentioned LS means was done, and experimental data on glycomic response of cell wall mutants' fractions (Fig. 3C and Dataset S1) was then incorporated to obtain predictive classification models (correlating wall composition with disease resistance and fitness phenotypes. More specifically, on the basis of the *p*-values of the one-sided Dunnett tests (α = 0.05), each genotype was assigned a class (e.g. a categorical valuation) for each response, which represent its status in such feature compared to the wt: either a plant has a similar performance to the wt (class *E*), or the plant performs significantly better than the wt (class *B*) or the plant performs significantly worse than the wt (class *W*). A supervised classification methodology was then applied in order to correlate wall composition with biotic stress resistance, fitness, and desiccation tolerance phenotypes as represented by the mentioned classes *E*, *B* and *W*. The available data to describe wall composition was given by 32 glycomic measurements of a set of 155 antibodies (for each cell-wall fraction) performed for 11 different genotypes of the Col-0 ecotype (Fig. 3) Thus, the objective of the present analysis was to uncover and generalize the potential relationships between the set of glycomic responses (that therefore act as independent or explanatory variables) and the classes that describe the performance of the genotypes (acting as dependent variables), for each of the studied features, to explain the different phenotypical status of the genotypes in the core set, and to generalize such patterns in order to provide a mathematical model exhibiting an adequate predictive accuracy.

Many statistical and machine learning techniques are available to fit supervised classification models, as for instance linear discriminant analysis, logistic regression, random forest or classification trees. In the present analysis we have chosen this last technique (classification trees) since it constitutes a well-known, standard classification methodology with almost no statistical assumptions that provides interpretability of the resulting models, automatic independent variables selection, and an adequate predictive capability for the purposes of the present study. More specifically, the particular classification tree algorithm applied, known as CRT (or CART, from Classification and Regression Tree), is a nonparametric technique that makes no assumptions on the distribution of the data and the variability and balance between classes. The bases of the CRT algorithm are both the identification of independent variables (in this case, cell-wall antibodies) and the definition of cut-points on these variables' values that allow separating in different branches of a tree the instances belonging to different classes. In this way, departing from an initial *root* node containing all the instances of the training sample, CRT tries to obtain purer or more homogenous nodes (in terms of their class composition) by computing the reduction of the node's impurity allowed by each of the available independent variables. Given a set of *n* classes, each appearing in a given node with a relative frequency f_i , $i = 1, \ldots, n$, CRT computes the node's impurity through the Gini Index

$$
I_G = \sum_{i=1}^n f_i (1 - f_i).
$$

Thus, from a given node, CRT selects both the variable and its cut-point that allow the greatest impurity reduction from the "father" node to the "child" nodes. CRT is based on binary branching, e.g. only two child nodes are produced from each father node being split. The main inconvenient of the CRT methodology is its tendency to produce overfit, e.g. to provide classification models that generalize not only the right, essential patterns or relationships between the explanatory variables and the classes, but also the potential noise that the specific data used to construct (or train) the model may contain. When overfit is present, the obtained model tends to provide excellent results on the training sample, but quite worse results on unseen instances not used in the construction of the trees, what is usually called a validation or test sample. To avoid overfit, we force the tree growth process to stop after the first split of the root node, and at the same time we carry out a demanding cross-validation process in order to get a precise estimation of the actual model's accuracy. Therefore, the obtained tree is composed of just the root node and two leaves, defined through a cut-point value for a single explanatory variable. In this way, we focus on uncovering just the main, most discriminant antibody for separating genotypes exhibiting a different phenotypical performance (see Fig. 4, SI *Appendix*, Fig. S10-S12). Once this was done, we focused on estimating the actual accuracy of the obtained model by conducting a 10-fold cross-validation process, e.g. the available data is randomly divided in 10 similar-size parts, and then similar trees are trained using 9 of these parts, leaving the remaining part to be used as a test sample (as these instances are not used in the tree fitting step). This step is carried out 10 times, each time using a different part as test sample. The proportion of testsample instances correctly classified throughout this 10-steps process provides an estimation of the original model's accuracy. To obtain a more robust estimation and average out the dependence on the random division step, the whole 10-fold cross-validation process was replicated 100 times, each time using a different random division of the available data in 10 parts. The mean and standard deviation of these 100 10-fold cross-validation estimations at each addressed classification task are reported in Fig. S10*B* and Table S1. Both the CRT models fit and their cross-validation were performed using Python *scikit-learn* library.

Glycome analysis by Enzyme-Linked ImmunoSorbent Assay (ELISA)

mAbs were obtained as hybridoma cell culture supernatants either from laboratory stocks (CCRC series, JIM series, MAC series; available from CarboSource [http://www.carbosource/net]) or from Plant Probes (LM series, PAM1 [http://www.plantprobes.net/]) unless otherwise indicated. A detailed list of all mAbs included in this study showing the immunogens used to develop them, their isotype, and the cell wall

polysaccharide class they primarily recognize is provided in Dataset S1. The experimental protocol previously described (5) was performed with minor modifications. In brief, flat-bottom 96-well ELISA plates (Nunc 439454 from Thermo Fisher Scientific or 2507 Costar from Corning Life Sciences), we used to apply cell wall mutant fraction polysaccharides (50 μL of 10 μg mL−1 in deionized water per well, or deionized water for controls) and then plates were dried to the well surfaces by evaporation overnight at 37°C. The plates were blocked with 200 μL of 1% (w/v) instant nonfat dry milk (Carnation) in Tris-buffered saline (50 mm Tris-HCl, pH 7.6, containing 100 mm sodium chloride) for 1 h. All subsequent aspiration and wash steps were performed using an ELx405 microplate washer (Bio-Tek Instruments). Blocking agent was removed by aspiration, and 50 μL of undiluted hybridoma supernatant of each antibody were added to the well and incubated for 1 h at room temperature. Supernatant was removed and wells were washed three times with 300 μ L of 0.1% (w/v) instant nonfat dry milk in Tris-buffered saline (wash buffer). Peroxidase-conjugated goat anti-mouse IgG or goat anti-rat IgG antibodies (Sigma-Aldrich), depending on the primary antibody used, was diluted 1:5,000 in wash buffer, and 50 μL were added to each well and incubated for 1 h. Note that the secondary antibodies used in this study are generated against whole immunoglobin molecules and thus bind to several isotypes of primary antibodies, including IgGs, IgMs, and IgAs, according to the manufacturers. Wells were then washed five times with 300 μL of wash buffer. 3,3',5,5'-Tetramethylbenzidine substrate solution (Vector Laboratories) was freshly prepared according to the manufacturer's instructions, and 50 μL were added to each well. After 20 min, the reaction was stopped by adding 50 μL of 0.5 n sulfuric acid to each well. The OD of each well was read as the difference in A450 and A655 using a model 680 microplate reader (Bio-Rad). The reading from each test well was subtracted from that of a control well on the same plate that contained the same primary and secondary antibodies but no immobilized polysaccharide. This experiment was repeated 3 times with 2 independent biological replicates.

Figure S1. *Arabidopsis thaliana m***utants used in the disease resistance and fitness analyses.** Gene locus, mutant allele tested and type of mutation (loss of function mutation (LFM) or hypomorphic mutation (HM) alleles) are described. In bold are indicated mutant alleles analyzed for fitness parameters (biomass and seed yield; Fig. 2). Additional mutant alleles tested in the disease resistance against *Pc* (Fig S4) are underlined. Reference of the mutant lines used are showed (T-DNA insertion lines or EMS-derived mutants with the indication the amino acid change or exon splicing changes). RT-PCR of those mutants characterized in this work are shown in Fig. S2. References describing previously a function of the indicated gene/protein in cell wall biogenesis/composition are indicated in green, in disease resistance are marked in red and if both phenotypes (cell wall/disease resistance) are indicated in blue.

Figure S2. RT-PCR expression of genes in *Arabidopsis* **mutants used in the disease resistance analyses.** Expression of the indicated genes were determined in four-week old wild-type plants (Col-0 and Ws) and in the indicated mutants by extracting total RNA and performing RT-PCR using the oligonucleotides indicated in Table S3. PCR amplification bands were quantified using Fiji-J and the expression of the genes in Col-0 and Ws was set to 100% and the values of RT-PCR amplifications (n=2) in the mutants determined and represented.

Susceptible

Wild type

Resistant

Figure S3. Disease resistance analysis of *Arabidopsis* **cell wall mutants**. Average disease resistance values (±SD) of wild-type plants and cell wall mutants in different backgrounds (black-Col-0; green-La-*er*; purple-Ws) to *P. cucumerina* BMM (DR from 1 to 5), *R. pseudosolanacearum* (GMI1000 strain for Col-0 and La-*er*, and RD15 for Ws; DR from 1 to 4), and *H. arabidopsidi*s (Noco2 strain for Col-0, Ewma1 for Ws and Cala for La-*er;* conidiospore/mg plant fresh weight). Color-code of the corresponding columns indicates the level of the resistance phenotype, from susceptible (blue) to resistant (red) whose differences with wild-type plants (white) are statistically significant (Dunnett's test p≤0.05). White color of the columns values of the mutants means not statistically significant from wt plants. Genotypes used as control of resistance (*cr*) or control of susceptibility (*cs*) for the different pathogens are indicated with shaded cell. For *Hpa* La*-er* and Col-0 wild-type ecotypes were included as cr for Col-0 and La-er/Ws mutants backgrounds, respectively
(conidiospores/mg fresh weight is equal to 0), and NahG plants (Col-0), eds1-1 (Ws) and eds1-2 (Col-0) alleles were used as *cs* for Col-0, Ws and La-*er* background, respectively.

Figure S4. Disease rating (DR) of *cwm* second alleles inoculated with *P. cucumerina*. DR average (±SD) of wild-type (wt) plants (Col-0 background) and mutants at 7 days post inoculation (dpi) with the necrotrophic fungus *P. cucumerina*. DR varies Colored columns indicate significant differences compared with wt values (ANOVA non-balanced analysis and Dunnett's test p≤0.05). This is one representative experiment of the three performed that gave similar results ($n = 10$).

Figure S5. Macroscopic disease symptoms in representative genotypes inoculated with *H. arabidopsidis***.** Pictures of Col-0, La-*er* and Ws wild-type plants, control of resistance (*cr*) and control of susceptibility (*cs*) genotypes, and some representative cell wall mutants at 7 days post inoculation with strains Noco2 (Col-0), Emwa1 (Ws) and Cala (La-*er*). La*-er* and Col-0 wild-type ecotypes were included as *cr* for those mutants in backgrounds Col-0 and La-*er*/Ws, respectively. *NahG* plants (Col-0), and *eds1-1* (Ws) and *eds1-2* (Col-0) mutant alleles were used as *cs* for Col-0, Ws and La-*er* background, respectively. In the picture magnifications, *Hpa* sporangiophores can be observed on surface of the inoculated leave. For additional details see Fig. S3.

Figure S6. Developmental phenotypes of three-week old plants of *cwm* **and wildtype ecotypes (Col-0, Ws and La-***er***).** Plants were grown under short day conditions as described in Material and Methods.

Figure S7. Desiccation tolerance of wild-type plants and cell wall mutants. *(A)* Percentage of plant survival after drought stress application for 21 days followed by plant re-watering. Data are the average of 10 plants. Columns color indicate significant differences (% of plant survival) compared with wild-type plants (wt) values (ANOVA non-balanced analysis, Dunnett´s test p≤0.05), with higher and lower values than wt indicated in red and blue, respectively. Mutant genotypes with lower values of survival than wt (blue color) were not identified. This is one representative experiment of the three performed that gave similar results. *(B)* Correlation analysis between desiccation tolerance and resistance to pathogens of 18 *cwm* mutants and wt plants. Average response information of each genotype (dot in the graph) is expressed in relation to that of the reference wild-type plant (black dot, value of 100% at the *y*-axes). Disease resistance ratios were logtransformed, and accordingly *x*-axes range from 0 (lower susceptibility) to 5 (greater susceptibility), with the wild-type plants situated at 4.72 = *ln* (1 + 100). A linear model was fitted for each combination and correlations determined. Fitted equations, R-squares and *p*-values are indicated in the insets of graphs. The *x*-axes of the figures involving *Pc* are enlarged in the 4-5 range for better visualization.

Figure S8. Cell wall biochemical composition of *Arabidopsis cwm* **plants.** (*A*) Total sugars quantification (% μg per mg of dry weight) in the non-cellulosic carbohydrate fraction from the cell walls of the mutants and their corresponding background plants (Col-0, La-*er* or Ws). (*B*) Total uronic acid (UA) content (% µg per mg of dry weight) from the cell walls of the indicated genotypes. (*C*) Cellulose content (% µg of total sugars per mg of dry weight). Data represent average values (± SE) of three independent experiments. Asterisks indicate mean values significantly different from wild-type plants (Student's *t*-test. *p* < 0.1; n > 10).

Figure S9. Total lignin content of *Arabidopsis* **wild-type and** *cwm* **plants**. Quantification of total lignin (µg acid insoluble lignin per mg cell wall) in the indicated mutants and their corresponding background plant (Col-0). Data represent average values (± SE) of three independent experiments. Asterisks indicate mean values significantly different from wild-type plants (Student's *t*-test. *p* < 0.1; n > 10).

B

Figure S10. Predictive CRT model correlating wall composition resistance/fitness/desiccation phenotypes of *Arabidopsis* **cell wall mutants**. (*A***)** Scheme of the CRT model obtained in the analysis of the correlation between resistance to *P. cucumerina* and cell wall epitopes. The tree model obtained with antibody CCRC-M106 (fucosylated xyloglucan) in PEC2 fraction of *cwm* and wild-type plants is shown. (*B*) Summary of the most relevant CRT models obtained for the different variables (resistance to pathogens, fitness and desiccation tolerance). For each variable, the CRT-selected antibodies detecting epitopes of some cell wall extracts as well as their absorbance cut-points are indicated (see A). The number of observations (*cwm* and wild-type plants) of each disease resistance/fitness/desiccation phenotype (resistant/higher (R/H), equal (E) or susceptible/lower (S/L) than wild-type plants) verifying either side of the cut-points is also reported. The predictive accuracy (% accuracy) of each model is estimated as the percentage of correct classifications in 100 replications of a 10 fold cross-validation process.

Figure S11. Predictive CRT model correlating wall composition and disease resistance/fitness phenotypes of *Arabidopsis* **cell wall mutants to** *H. arabidopsidis and R. pseudosolanacearum*. *(A)* Scheme of the CRT models obtained in the analysis of the correlation between resistance to *H. arabidopsidis* or *R. pseudosolanacearum* and cell wall epitopes. The tree models obtained with antibodies in cell wall fractions of cell wall mutants and wild-type plants are shown. *(B)* Biological validation of CRT results with cell wall mutants from 6 clusters analyzed*.* The absolute value (average ±SD) of the epitope signal detected by the antibody are shown (*n* = 3). The color code of the column indicates the resistance level of the corresponding mutant, from red (resistant) to blue (susceptible) in comparison with wild-type (wt) disease resistance level (in white) (see Fig. 1). The absorbance cut-point value for considering a mutant as resistant, as determined by CRT, is indicated by the dotted lines. *Arabidopsis* cell wall mutants that fulfill the CRT model are marked with an asterisk.

Figure S12. Predictive CRT model correlating wall composition and fitness phenotypes of *Arabidopsis* **cell wall mutants**. *(A)* Predictive CRT model correlating wall composition and fitness phenotypes of *Arabidopsis* cell wall mutants. *(A)* Scheme of CRT models obtained in the analysis of the correlation between fitness and cell wall epitopes. The trees models obtained with antibodies in cell wall fractions of cell wall mutants and wild-type plants are shown. *(B)* Biological validation of CRT results with the cell wall mutants analyzed. The absolute value (average ±SD) of the epitope signal detected by the antibody are shown (n = 3). The color code of the column indicates the level of the fitness value corresponding to the cell wall mutant, from red (enhanced values) to blue (reduced values) in comparison with wild-type (wt) values (in white) (see Fig. 2A,B). The absorbance cut-point value for considering a cell wall mutant as matching the CRT model is indicated by the dotted lines. *Arabidopsis* cell wall mutants that fulfill the CRT model are marked with an asterisk.

Figure S13. Clustering of the expression pattern of canonical defensive genes and disease resistance phenotypes in wild-type plants and cell wall mutants and their disease resistance phenotypes. (*A*) Clustering of disease resistance phenotypes and expression levels (determined by qRT-PCR) of defensive and MAMP-induced genes in non-inoculated three-week old plants from the indicated genotypes. (*B*) Relative expression (PcBMM/Mock at 1-day post inoculation, dpi). Clusters were computed using Euclidean distances for absolute gene expression levels and disease indexes, and Z-scores were calculated across rows for normalization. Genes expression levels relative to the *UBC21* gene and to mock-treatment (S13A and S13B, respectively) are shown. Values are means ± SE (*n* = 3). Experiments were performed three times with similar results.

Figure S14. Cell wall derived $frac$ **fractions (PEC1 and PEC2) of** *cwm* **plants trigger Ca2+ elevations in Col-0 AEQ plants.** Increases in cytoplasmic
calcium concentrations ([Ca²⁺]_{cyt}) in seedlings of Col-0AEQ upon treatment with heat treated (121 ºC for ²⁰ min) cell walls extracts (PEC1 and PEC2, ⁵⁰ ng/µl) from Col-0 wild-type plants and the indicated cell wall mutants. PEC1 corresponds to the weakly bound pectin fraction whereas PEC2 is the highly bound pectin fraction. Data are representative of three independent experiments that gave similar results (means ± SD, n = 16).
Asterisks indicate statistically significant differences (*p* < 0.05) with the corresponding fraction from Col-0 plants (ANOVA, Dunnett´s multiple comparisons test correction). Total free Ca²⁺ in PEC1 and PEC2 cell wall fractions used in these analyses was determined by atomic absorption spectrophotometry, but not detectable amounts of Ca2+ were found.

Figure S15. Cell wall derived extracts from PEC1 of cell wall mutant plants trigger immune responses in *agb1-2* **and** *bak1-5* **mutants impaired in the immune regulators AGB1 and BAK1, respectively.** (*A*) Increases in cytoplasmic calcium concentrations ($[Ca^{2+1}]_{\text{cyl}}$) in seedlings of Col-0^{AEQ} and *agb1*-2^{AEQ} lines upon
treatment with heat-treated (121 °C for 20 min) cell wall extracts (50 ng/µl PEC1) from Col-0 and plants (means ± SD, n = 8). Flagellin 22 peptide MAMP (flg22; 1 µM) was used as positive control. Asterisks indicate statistically significant differences between Col-0 and *agb1-2* mutant plants treated with the same fraction or flg22 (*p* < 0.1; ANOVA, Dunnett multiple comparisons test correction). (*B*) Mitogen-activated protein kinases (MAPK) phosphorylation in 12-days old *Arabidopsis* seedlings of Col-0 and *bak1-5* upon treatment with heat-treated (121 ºC for 20 min) cell wall extracts (0.25 ng/µl PEC1) from Col-0 and the indicated *cwm* plants. Western Blot using anti-pTEpY antibody for phosphorylated MAPK moieties at different time points: 0 (seedlings harvested before treatment with PEC1 or flg22), and 10 and 20 minutes after treatment. Arrows indicate the position of MPK6, MPK3 and MPK4/11 proteins. Ponceau red-stained membranes show equal loading. Flg22 (1 µM) and distilled water (mock) were used as controls. Intensity of the bands (relative units) were quantified for each sample with iBright
Analysis Software and values are indicated. Additional details of the methodology used for Western analyses explained in SI Material and Methods.

Figure S16. Schema of the mathematical analyses performed to generate the data of Figures 2C, Fig. 4, Fig. 4,

Table S1: Summary of the results from the performed Classification and Regression Tree (CRT) analysis. The observed phenotypes and the cell wall fractions used for the analyses are indicated. The monoclonal antibodies which provided a better classification of the observed phenotypes were selected, and the epitope they recognise are shown. In dark green and pale green are indicated the antibodies providing the top 1 and 2 highest CV_ accuracy for the phenotypes analysed, respectvely. Cut-off values are absorbance values that determine the two classification categories. Raw and corrected accuracy values (CV_) are shown. For additional details see SI.

Table S2: Oligonucleotides used for T-DNA insertional mutant characterization.

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