Comparative transcript profiling by SuperSAGE identifies novel candidate genes for controlling potato quantitative resistance to late blight not compromised by late maturity

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INTRODUCTION

Plant resistance to microbial attacks manifests itself at three phenotypic levels. First, the entire plant species is immune against infection by all isolates of a microbial species, since pathogen entry into plant cells is blocked at a very early stage (non-host resistance) (Nürnberger and Lipka, 2005). Second, avirulent entry into plant cells is blocked at a very early stage (non-infection by all isolates of a microbial species, since pathogen resistance breeding can be enhanced by molecular diagnostic tools, which result from knowledge of the genetic basis of resistance and susceptibility. Breeding uses resistance conferred by single R genes and polygenic quantitative resistance. The latter is partial but considerable more durable. Molecular mechanisms of plant pathogen interactions are elucidated mainly in experimental systems involving single R genes, whereas most genes important for quantitative resistance in crops like potato are unknown. Quantitative resistance of potato to Phytophthora infestans causing late blight is often compromised by late plant maturity, a negative agronomic character. Our objective was to identify candidate genes for quantitative resistance to late blight not compromised by late plant maturity. We used diagnostic DNA-markers to select plants with different field levels of maturity corrected resistance (MCR) to late blight and compared their leaf transcriptomes before and after infection with P. infestans using SuperSAGE (serial analysis of gene expression) technology and next generation sequencing. We identified 2034 transcripts up or down regulated upon infection, including a homolog of the kiwi fruit allergen kiwellin. 806 transcripts showed differential expression between groups of genotypes with contrasting MCR levels. The observed expression patterns suggest that MCR is in part controlled by differential transcript levels in uninfected plants. Functional annotation suggests that, besides biotic and abiotic stress responses, general cellular processes such as photosynthesis, protein biosynthesis, and degradation play a role in MCR.

Keywords: potato, late blight, transcript profiling, SAGE, marker-assisted selection
Eulgem, 2005; Bent and Mackey, 2007; Koornneef and Pieterse, 2008). In contrast, the understanding of quantitative resistance at the molecular level is very limited. Genetic evidence and plausibility suggest that genes functional in pathogen recognition, defense signaling and defense responses are, at least in part, involved in quantitative resistance (Poland et al., 2009; Kou and Wang, 2010).

The improvement of crop plant genetic resistance to pests and pathogens is a major objective of plant breeding. In this respect, both $R$ gene mediated resistance and quantitative resistance are important. Resistance factors are incorporated in advanced cultivars by introgression breeding, mostly from landraces or closely related wild species. However, in a number of pathosystems of worldwide relevance, $R$ gene mediated resistance often proved not durable under field conditions (Wastie, 1991; Skamnioti and Gurr, 2009). $R$ genes introgressed into varieties were soon defeated by new, virulent races of the pathogen, due to the fact that single mutations in the pathogen can overcome $R$ gene mediated resistance. Such mutations are favored by strong selection pressure, when highly resistant varieties carrying single $R$ genes are widely cultivated. Quantitative resistance is considered more durable due to its partial and polygenic nature, which exerts less selection pressure on the pathogen and requires more than one mutation to overcome resistance. However, by the same reasons, quantitative resistance is more difficult to analyze and to handle in breeding programs (Parlevliet, 2002; Poland et al., 2009).

The interaction of the oomycete Phytophthora infestans with the potato (Solanum tuberosum) causes late blight, the world wide most important disease in potato cultivation (Kamoun and Smart, 2005; Forbes, 2012). $R$ genes introgressed during the 20th century from the wild potato species Solanum demissum proved not durable in the field, a fact that shifted the interest of potato breeders to quantitative resistance (Ross, 1986; Wastie, 1991; Colon et al., 1995). The phenotypic selection of cultivars with quantitative resistance requires multi-year and location trials and is complicated by the fact that quantitative resistance is often accompanied by other, negative agricultural attributes such as late plant maturity (Visker et al., 2004; Forbes, 2012). Plant maturity is a quantitative developmental trait influenced by day length (Kloosterman et al., 2013). It refers to the time the potato plant requires to complete its annual life cycle, beginning with sprouting, shoot/leaf growth, flowering, tuber initiation, and ending with tuber maturation and foliage senescence. Knowledge of the loci and their natural allelic variants that control quantitative resistance not compromised by late maturity will facilitate the selection of improved cultivars by means of DNA-markers, which are diagnostic for superior resistance or inferior susceptibility alleles.

Genetic dissection of quantitative resistance to late blight via molecular linkage mapping has identified a minimum of 20 quantitative trait loci (QTL) on the 12 potato chromosomes (Gebhardt and Valkonen, 2001; Simko et al., 2007; Danan et al., 2011). Most QTL for resistance to late blight detected so far were linked with QTL for plant maturity (Collins et al., 1999; Oberhagemann et al., 1999; Bormann et al., 2004; Visker et al., 2005; Danan et al., 2011). Some resistance QTL co-localized with $R$ genes for resistance to $P. infestans$ and other pathogens or with genes of unknown function that share high sequence similarity with $R$ genes ($RGL$, resistance gene like), with genes functional in defense signaling and with genes involved in defense responses (Leonards-Schippers et al., 1994; Leister et al., 1996; Gebhardt and Valkonen, 2001; Trognotz et al., 2002; Pajerowska et al., 2005). The genetic linkage studies collectively support the idea that natural variation in genes that control $R$ gene mediated resistance is causal for quantitative resistance to $P. infestans$ and other pathogens. They formed the basis of a candidate gene approach toward the identification of DNA polymorphisms associated with quantitative resistance to late blight in multi-parental populations of potato varieties and breeding clones (Gebhardt et al., 2004; Malosetti et al., 2007; Pajerowska-Mukhtar et al., 2009).

In particular, association mapping discovered single nucleotide polymorphisms (SNPs) in two potato genes coding for allene oxide synthase ($StAOS2$ and $StAOS1$), which were associated with resistance to late blight not compromised by late maturity (Pajerowska-Mukhtar et al., 2009). Allene oxide synthase is a key enzyme in the biosynthesis of jasmonates, a class of lipid-derived plant hormones involved in defense and stress responses as well as developmental processes (Kombrink, 2012). Two neighboring SNP markers, $StAOS2$-snp691 and $StAOS2$-snp692 at the $StAOS2$ locus on potato chromosome XI explained between thirty and forty percent of the genetic variation of maturity-corrected resistance to late blight (MCR) in a population of 184 breeding clones (Pajerowska-Mukhtar et al., 2009). Silencing the expression of $StAOS2$ resulted in increased susceptibility to $P. infestans$. This result together with quantitative functional complementation analysis of five natural $StAOS2$ alleles in an $aos$ knock-out mutant of Arabidopsis thaliana provided evidence that DNA variation at the $StAOS2$ locus is responsible, for a portion of the quantitative resistance to $P. infestans$ (Pajerowska-Mukhtar et al., 2008). The role of $AOS$ in quantitative resistance was not restricted to the pathosystem $S. tuberosum/P. infestans$, as overexpression of a rice $AOS$ increased resistance of transgenic plants to the rice blast fungus Magnaporthe grisea (Mei et al., 2006).

Although successful, the candidate gene approach is necessarily biased toward genes with known function in pathogenesis and defense. To discover additional, novel genes involved in quantitative resistance, unbiased approaches such as comparative transcript, protein, or metabolite profiling are required. Few comparative transcript profiling studies addressed aspects of quantitative resistance to late blight in potato using subtractive hybridization approaches (Ros et al., 2004; Henriquez and Daayf, 2010) or cDNA microarrays (Restrepo et al., 2005; Wang et al., 2005; Lindqvist-Kreuze et al., 2010). In these studies, known and novel transcripts were identified that were up or down regulated upon infection with $P. infestans$ in a compatible interaction. In most cases two different genotypes with different levels of resistance to late blight were compared. Such a comparison does not allow to distinguish whether transcripts are differentially expressed due to different resistance levels or due to unspecific differences between individual genetic backgrounds that are unrelated with the resistance phenotype. SuperSAGE (serial analysis of gene expression) combined with next-generation sequencing is a novel platform for capturing
and quantifying in depth and de novo the transcriptome in a biological sample (Matsumura et al., 2003, 2008; Gilardoni et al., 2010). SuperSAGE generates 26 base pair sequence tags from each transcript in the sample. The tag frequency is proportional to the amount of the corresponding transcript in the sample. Comparisons of tag frequencies between pairs of samples reveal differences in expression level. Mapping the tag sequences to collections of expressed sequences and to annotated genome sequences allows annotation and thereby the identification of putative functions of the corresponding transcripts.

Here we describe the results of a comprehensive transcript profiling experiment using SuperSAGE technology. The plants analyzed resulted from marker-assisted selection for increased vs. decreased quantitative resistance to *P. infestans* not compromised by late plant maturity. Our objectives were (i) the validation of the diagnostic power of SNP markers at the *StAOS2* locus that were previously shown to be associated with maturity-corrected resistance (MCR) to late blight and (ii) the identification of novel potato candidate genes for having a functional role in quantitative resistance to *P. infestans*.

**MATERIALS AND METHODS**

**PLANT MATERIAL AND MARKER ASSISTED SELECTION (MAS)**

Based on prescreening tetraploid, heterozygous parental clones with the SNP markers *StAOS2-snp691* and *StAOS2-snp692*, two half-sib families derived from the crosses *Phy20 × Phy14* and *Phy20 × Phy16* were chosen for MAS. The parents *Phy14*, *Phy16*, and *Phy20* were breeding clones from the commercial breeding program of SaKa Pflanzenzucht GmbH & Co. KG (Windeby, Germany). The common seed parent *Phy20* (genotype: *StAOS2-snp691: AAGG; StAOS2-snp692: CCGG*) was maturing mid-early (score 5) and was more susceptible to late blight than the pollen parents *Phy14* and *Phy16* (MCR = 64.2). The pollen parents *Phy14* (genotype: *StAOS2-snp691: AAAAA; StAOS2-snp692: CCCCC*) and *Phy16* (genotype: *StAOS2-snp691: AAGG; StAOS2-snp692: CCGG*) were maturing mid-early (score 5) and mid-early to late (score 6), respectively, and expressed good maturity corrected resistance (MCR) to late blight (*Phy14: MCR = −57.4; Phy16: MCR = −39.2*). One tuber per *F1* seedling was harvested in 2007 and planted in the field at Windeby (Germany) in 2008. Approximately 100 mg fresh leaf tissue (one young leaflet) per genotype were collected in 96 raised collection tubes (Qiagen, Hilden, Germany) and freeze dried. Genomic DNA was extracted from freeze dried leaf tissue using a BioSprint 96 robotic workstation (Qiagen, Hilden, Germany) according to the manufacturer’s protocol. Plants were genotyped in 2008 for the SNP markers *StAOS2-snp691* and *StAOS2-snp692* by amplicon sequencing as described previously (Pajerowska-Mukhtar et al., 2009). Three genotype groups A1, A2, and B2 were selected. Marker-selected plants (SL genotypes) were tuber propagated in 2009 under standard phytosanitary regimes and re-genotyped using newly isolated DNA from field grown plants. Only genotypes with identical marker scores in both years were used for preparing SuperSAGE samples.

**FIELD EVALUATION OF MATURITY CORRECTED RESISTANCE (MCR) TO LATE BLIGHT**

Marker selected SL genotypes were evaluated in 2010 in separate field trials for plant maturity and resistance to late blight as described (Bormann et al., 2004; Pajerowska-Mukhtar et al., 2009). Briefly, plant maturity was scored from 1 to 9 (1 very early, 9 very late maturing) in comparison to standard varieties of known maturity type. Two replications, each with 12 plants per genotype (lattice design with partial blocks), were infected on 5th July 2010 with a mixture of complex *P. infestans* field isolates (provided by Julius Kühn Institutes Braunschweig and Groß-Lüsewitz, Germany). Using mixtures of *P. infestans* isolates insured that all major genes for late blight resistance potentially present in the plants were overcome. Disease progression was monitored by scoring the plants every 3 to 4 days from 12th July to 19th August, using a 0–9 scale (0: no symptoms, 9: plants are dead). Maturity corrected resistance (MCR) was calculated from the regression curve of the “area under disease progress curve” (AUDPC) against plant maturity. MCR corresponded to the vertical distance of each AUDPC value to the regression curve. Negative MCR values indicate higher levels of resistance, whereas positive MCR values indicate higher susceptibility (Bormann et al., 2004).

**PREPARATION OF TISSUE SAMPLES FOR SuperSAGE AND qRT-PCR**

Twenty nine SL genotypes were multiplied by shoot cuttings. Six to seven weeks old plants (three plants per SL genotype) were transferred to a growth chamber (Snijders Scientific B.V., Tilburg, Netherlands) and acclimated for few days to 16 h light at 22°C and 8 h dark at 20°C before infection with *P. infestans*. The same mixture of *P. infestans* isolates as used for field inoculations (see above), which included all 11 race specificities were used for infection experiments. The race composition was confirmed by a detached leaflet test (Goth and Keane, 1997) using the race 1–11 differential set of potato cultivars from the James Hutton Institute (former Scottish Crop Research Institute, Invergowrie, UK). *P. infestans* was propagated on detached leaflets of 6–10 weeks old susceptible plants (cvs. Désirée or Grata) or on tuber slices of the susceptible variety Grata. Inoculum was prepared as described (Gyetvai et al., 2012). Three compound leaves per plant (3rd, 4th, and 5th from the top) were spray inoculated on the abaxial side each with 200–400 µl sporangia suspension (20,000 sporangia/ml, 65,000–130,000 zoospores/ml) using a pump glass flask. The plants were then covered with clear plastic bags to insure high humidity for optimal growth of *P. infestans*. Inoculations and tissue sampling were performed between 8:00 and 11:30 A.M. Individual leaflets of similar size were collected from the 4th and 5th compound leaf just before inoculation (*T0*), one (*T1*) and two (*T2*) days post inoculation. With these infection time points we captured two time slots from the mid to late biotrophic growth phase of *P. infestans* (Avrova et al., 2003).
different plant to avoid wounding and priming effects. The infection experiment was repeated five times, each with new batches of plants and inoculum. The best three infection experiments (experiment 1, 2, and 3) were chosen for qRT-PCR experiments (see below). Experiment 1 was used for constructing SuperSAGE libraries.

**SuperSAGE LIBRARY CONSTRUCTION AND SEQUENCING**

SuperSAGE libraries were constructed from total RNA of nine leaf samples from infection experiment 1. For each time point (T0, T1 and T2) one leaflet each of similar size from 14, 6, and 9 SL genotypes in genotypic groups A1, A2, and B2 (Table 1), respectively, were pooled. Frozen pooled leaf tissue was powdered in a CryoMill (Retsch, Haan, Germany). Total RNA was extracted from powdered tissue using the Plant RNA Isolation Kit of Invitrogen (Karlsruhe, Germany) following the manufacturer’s protocol. RNA purity, removal of genomic DNA contamination, concentration measurements and quality tests were performed as described (Draffehn et al., 2010). RNA integrity was additionally assessed by electrophoresis using an Agilent 2100 Bioanalyzer (Agilent Technologies, Santa Clara, USA). SuperSAGE libraries were prepared at GenXPro GmbH (Frankfurt, Germany) essentially as described (Matsumura et al., 2010). To prevent amplification biases the TrueQuant technology was applied as described by B. Rotter (Patent application Nr. WO2009152928). Two libraries were constructed for each RNA sample, one using NlaIII (recognition site: 5′-CATG-3′) and the other using DpnII (recognition site: 5′-GATC-3′). Transcripts that do not have a NlaIII restriction site close to the 3′ end cannot be detected as no tag is generated. Those transcripts may be rescued by preparing libraries using a second anchoring enzyme. Therefore, the 26 bp tags carry either CATG or GATC at their 5′ end. The 18 libraries (3 genotype pools, 3 time points, 2 restriction enzymes) were pooled and sequenced by Solexa/Illumina technology on an Illumina GAII instrument (Illumina, Inc., USA). The sequence data analyzed in this paper have been deposited in NCBI’s Gene Expression Omnibus (Edgar et al., 2002; Barrett et al, 2013) and are accessible through GEO Series accession number GSE48071 (http://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE48071).

**Table 1 | Summary of marker-assisted selection for MCR.**

| Cross combination | Phy20 × Phy14 | Phy20 × Phy16 |
|-------------------|---------------|---------------|
| No. of F1 plants genotyped | 309 | 142 |
| No. of F1 plants selected | 14 | 6 | 9 |
| Genotype StAOS2-snp689 | AAAA | AAAA | AGGG |
| Genotype StAOS2-snp682 | CCC | CCC | CGGG |
| Genotype group name | A1 | A2 | B2 |
| No. of genotypes evaluated in the field | 11C | 4 | 8 |
| Mean MCR | 26.2b | −69.8a | 5.9b |

*a,b* Post-hoc test (LSD) for significant differences between groups.

**DATA ANALYSIS**

The primary sequence data were purified as described (Matsumura et al., 2010). Tag counts were normalized (tags per million = tpm). The 26 bp tags were annotated by BLAST analysis (Altschul et al., 1990) using the databases STGI.042210 for potato transcripts (Ronning et al., 2003) and /NCBI/ENTREZ/NCBI_P.infestans_mRNA+REFSEQAUG10/NCBI_P.infestans_mRNA+REFSEQ.Fasta, BROAD_Ins/phytophthora_infestans_t30-4_1_genest.fasta, and BROAD_Ins/phytophthora_infestans_t30-4_1_transcripts.fasta for transcripts of *Phytophthora infestans*. For tag annotation, at least 20 bp of 26 bp total had to be identical with the sequence of the target transcript. Different tags matching to the same transcript were totaled to obtain absolute and normalized “hit” counts (hits per million = hpm). Likelihoods for different tag and hit counts in pair wise comparisons were calculated according to (Audic and Claverie, 1997). For mapping tags to the potato genome sequence (Xu et al., 2011), 266,361 unique tags (unitags) with more than five counts across all samples were extracted from the SuperSAGE output files and converted into FASTA format. Tag mapping was performed in forward and reverse orientation with the BWA (Burrows-Wheeler Alignment) software (Li and Durbin, 2009) (http://bio-bwa.sourceforge.net/). Tags were mapped against the annotated genome of *Solanum phureja* Group tuberosum (PGSC_DM_v3_2.1.11_pseudomolecule_annotation.gff) allowing either no mismatch or up to three base pairs mismatches. BWA output files were filtered for tags with alignment quality scores larger than 20. Venn diagrams were constructed using the package VennDiagram v.1.5.1 of the statistical software R v.2.15.1 (Venables et al., 2008). Differentially expressed tags (*p < 10<sup>−4</sup>* of fifteen pair wise comparisons were visualized.

**In silico MAPPING OF DNA-MARKERS AND TRANSCRIPTS**

Restriction fragment length polymorphism (RFLP) markers linked to QTL for resistance to late blight and other pathogens were collected from the literature (Leonards-Schippers et al., 1994; Zimnoch-Guzowska et al., 2000; Ballvora et al., 2011; Danan et al., 2011). Potato marker sequences were obtained from the PoMaMo (Potato Maps and More) database at http://www.gabipd.org/ (Meyer et al., 2005) and tomato marker sequences from the SOL genomics network (SGN) at http://solgenomics.net/. Sequences of markers and potato transcripts were matched by BLAST (Altschul et al., 1990) to the potato pseudomolecules (version v2.1.11) (Xu et al., 2011) at http://solanaceae.plantbiology.msu.edu/integratedsearches.shtml to retrieve genomic positions and the corresponding loci. At least 90% sequence identity had to be reached between query and target sequence.

**QUANTITATIVE REAL TIME qRT-PCR**

cDNA synthesis and qRT-PCR were performed as described (Draffehn et al., 2010, 2012). Primer sequences, amplicon lengths, and annealing temperatures are listed in Table 2. Whenever possible, one primer was designed such that it included all or part of the differential SuperSAGE tag with the best match to the target transcript, to promote locus specific amplification. PCR conditions for all primer pairs were: Initial denaturation at 95°C...
### Table 2 | Primers used for semiquantitative qRT-PCR.

| Transcript | Annotation | Locus PGSC0003DMG | Chromosome | Forward (F) and reverse (R) primers (5′ to 3′) | Amplicon length (bp) | $T_a$ (°C) | Differential tag included in primer |
|------------|------------|-------------------|------------|-----------------------------------------------|----------------------|----------|-----------------------------------|
| TC194752   | Putative membrane protein | 400020757         | V          | F: GTTGGCGTTGCTGGATGGG R: GTGGTCTGCGCTTGGCCCTCA    | 150                  | 65       | no                                |
| TC194940   | Osmotin-like protein | 4000003057        | VIII       | F: AGAGGTGGGTTGCGAGACCG R: ACAAGAAAAGAGGTGCTCACCTCATG  | 550                  | 62       | yes                               |
| TC195330   | Ethylene response factor 5 | 400010753        | XII        | F: AGGTGCCGCACCTGTCAC R: AATGAAATTCAAATTGGAAGATC | 160                  | 57       | yes                               |
| TC195401   | CCR4 associated factor 1-related protein | 400005899      | VI         | F: TCTGTAGACCGGGCGGT R: CTAATCCATACAGAGCGCCAG     | 140                  | 55       | yes                               |
| TC197025   | Kiwellin    | 4000008101        | XI         | F: CGATGACCCTGACGTTGG R: GACCACCCCCGTCCTCAC       | 130                  | 60       | no                                |
| TC199390   | 1,3-beta-glucan glucanohydrolase | 400020017       | I          | F: GCAATCGGTTAAGCTGGTTGGG R: ACGGACAAAGGGTGCCCGG | 170                  | 65       | no                                |
| TC200652   | Caffeoyl-CoA O-methyltransferase | 400006214      | X          | F: ATCGGTAGTGCGACACCTGATG R: GCCGCGACACAGGGTGATGC     | 150                  | 65       | yes                               |
| TC204989   | Lipoxigenase | 400010859        | VIII       | F: CTCCCACCGCCCAACG R: TCACCATTCTACGTGATAATTTGATCC     | 300                  | 58       | yes                               |
| TC206389   | ZPT2-13     | 400015533         | XI         | F: CATGAGGCGTCATCGCCG R: CAATTTGGTGCGTGC           | 150                  | 60       | yes                               |
| TC208021   | Jasmonate ZIM-domain protein | 400002930       | XII        | F: GAGCTGTGGAAAGGTGCTAGAGC R: CATAGTTAGTCAAAAGTGCTCACTC   | 160                  | 58       | yes                               |
| TC209317   | Non-specific lipid-transfer protein | 400012481       | VIII       | F: GCCACCCCCAGTCACG R: GTGAAACAAACAAATACCATGTA       | 167                  | 55       | no                                |
| TC212020   | 2-oxoglutarate-dependent dioxygenase | 400011751      | VII        | F: CACGCACTATTGGGAGCCC R: TTAGGAAAGAAGATCTTATCTCATG  | 176                  | 59       | yes                               |
| TC218213   | Lipid desaturase | 400035878 or 400036004 | ? | F: CACCACACTCTCCTTCATTG R: GCTTTTTGGCTGCTCAACTGCG     | 190                  | 58       | no                                |
| TC221793   | Xyloglucan-endotransglucosylase | 400000408       | XI         | F: GTCTAGTTCTACCTGCCC R: CAAATTGCTCCGGGATTG       | 90                   | 55       | yes                               |
| TC225046   | Conserved gene of unknown function | 400027566      | IV         | F: AATACTACGCGCGGTCGCG R: CATATATAATAATTCAAAAGACATG    | 150                  | 57       | yes                               |

The TC identifiers according to The Solanum tuberosum Gene Index (http://compbio.dfci.harvard.edu/cgi-bin/tgi/tc_ann.pl?gudb=potato), annotation, locus identifier according to PGSC (http://potatogenomics.plantbiology.msu.edu/index.html), chromosome, primers, amplicon length, annealing temperature ($T_a$) and information about the integration of the significant SuperSAGE tag in one of the primer of the primer pair are given.

? indicates that no genomic position is available for this locus.
for 50 cycles of 95°C for 15 s, annealing at the temperatures listed in Table 2 for 30 s and extension at 72°C for 45 s. At the end of the PCR, reactions were tested for undesired primer dimer formation by melting curve analysis (55–95°C with a heating rate of 0.1°Cs−1 and continuous fluorescence measurement). In preliminary experiments the observed variation between technical replicates was small compared to the variation between different infection experiments. Average transcript levels were calculated from two technical replicates of each sample (three genotype pools at T0, T1, and T2) of three infection experiments. Expression levels were normalized against transcripts of the SAND gene (Czechowski et al., 2005; Exposito-Rodriguez et al., 2008) using the primer combination 5′-CTTGGAGGAACAGACG-3′ and 5′-GCAACAGGCCCTGATAC-3′. PCR was carried out at 58°C annealing temperature and using otherwise the same conditions as above. Amplification of SAND resulted in a 164 bp fragment. Genotype pools were tested for successful infection using as internal marker the P. infestans transcript for the 60 s ribosomal protein L23a (PTTG_02694) (Table 2). The P. infestans transcript values were not normalized against SAND.

RESULTS

MARKER-ASSISTED SELECTION (MAS) FOR MATURITY-CORRECTED RESISTANCE (MCR) TO LATE BLIGHT

Four hundred and fifty one tetraploid, heterozygous F1 plants from two half-sib families were subjected to genotypic selection with the two neighboring SNP markers StAOS2-snp691 and StAOS2-snp692. The haplotype StAOS2_A691G692 was associated with increased resistance (negative MCR values) in a dosage dependent manner, whereas the haplotype StAOS2_G691G692 was associated with increased susceptibility (positive MCR values) (Pajerowska-Mukhtar et al., 2009). Fourteen and six genotypes were selected from 309 F1 plants of the family Phy20 × Phy14 and 142 F1 plants of the family Phy20 × Phy16, respectively, which were homozygous for haplotype StAOS2_A691G692 (Table 1). Plants homozygous for the haplotype StAOS2_G691G692, were not present in the F1 families due to the parental genotypes. As contrasting genotype group we selected nine genotypes of the family Phy20 × Phy16, which carried three copies of haplotype StAOS2_G691G692 and one copy of haplotype StAOS2_A691G692 (Table 1). Subsequently we refer to the three selected genotype groups as A1, A2, and B2. Groups A1 and A2 were identical with respect to the StAOS2_A691G692 haplotype but differed in genetic background as the plants originated from different cross combinations, whereas groups A2 and B2 originated from the same cross but differed with respect to the genotype at the StAOS2 locus. The field evaluation of MCR of plants in groups A1, A2, and B2 showed that plants in group A2 were on average more resistant than plants in group B2. However, plants in group A1 were on average as susceptible as plants in group B2, irrespective of the fact that the A1 plants possessed the StAOS2 SNP haplotype associated with higher resistance (Table 1).

COMPARATIVE TRANSCRIPT PROFILING BY SuperSAGE

The plants in genotype groups A1, A2, and B2 and their average MCR values provided the basis for comparative transcript profiling by SuperSAGE. Figure 1 shows a flow chart of the experimental design and the data analysis. Nine leaf tissue samples were prepared from plants artificially inoculated with P. infestans in a growth chamber. In order to average transcript frequencies across individual, heterozygous genetic backgrounds, SuperSAGE libraries were constructed from total leaf RNA of the pooled 14, 6, and 9 genotypes in groups A1, A2, and B2, respectively, prior to infection (T0), one (T1), and 2 days (T2) after infection (see Materials and Methods). The combined tag counts of both NlaIII and DpnII libraries resulted in 1.1 to 6.2 million tags per sample (Supplementary Table S1). Of total 266,361 unique tags (unitags), 52.6% mapped to the potato genome sequence (Xu et al., 2011) when allowing up to three mismatches per 26 base pairs, and 23.3% mapped without mismatch. Allowing three mismatches took into account the high level of intra-specific DNA polymorphism in Solanum tuberosum Group tuberosum genotypes (Rickert et al., 2003) and the fact that tags from S. tuberosum Group tuberosum genotypes were matched against the genome of a related, homozygous diploid genotype “DM” of Solanum phureja Group tuberosum (Xu et al., 2011). Moreover, allowing mismatches enables the detection of allelic tags (two slightly different tags matching to the same position in the genome). 67.7% and 73.5% of the unitags with up to three and zero mismatches, respectively, were located in annotated genes, while the rest matched to putative intergenic regions (Table 3, Supplementary Tables S2, S3).
Table 3 | Summary of tags matching to the potato genome sequence.

|                          | Total no. of unitags | No. of matching unitags | No. of unitags matching to gene | No. of unitags matching to putative intergenic regions |
|--------------------------|----------------------|-------------------------|---------------------------------|--------------------------------------------------------|
| Up to 3 mismatches       | 266 361              | 140 046                 | 94 452                          | 45 595                                                 |
| No mismatch              | 266 361              | 61 970                  | 45 567                          | 16 403                                                 |

**Pair wise comparisons between SuperSAGE samples**

Fifteen pair wise comparisons were performed between the nine samples in order to identify transcripts differentially expressed in response to infection (six comparisons: three genotype pools, two time points) or between the three genotype pools before (T0), one (T1), and 2 days (T2) after infection (nine comparisons: three genotype pools, three time points). The number of unitags per comparison ranged from 127,000 to 182,000 (Supplementary Table S4). When matched against expressed sequence databases of potato (DFCI potato gene index) and *P. infestans* (Materials and Methods), on average 57% of the unitags did not match any expressed sequence (“no hits”). This was in part due to tags consisting mainly of A stretches derived from polyadenylated RNA. The occurrence of non-matching tags without obvious A stretches indicated that a portion of transcribed potato genes was not represented in the DFCI database. On the other hand, many tags matched to more than one expressed sequence (transcript) with similar annotation (see Supplementary Tables S8 and S10 below). In most cases, these transcripts matched with the top score to the same locus in the potato genome, suggesting that they corresponded to allelic variants. Ambiguities arose when transcripts matched with similar high scores to different members of multigene families or to transcripts with inconsistent annotations (see below).

The numbers of differentially expressed tags obtained in the pair wise comparisons when using arbitrary cut-off values of $p \leq 10^{-6}$, $p \leq 10^{-5}$ and $p \leq 10^{-4}$ for the likelihood of a difference, are summarized in Supplementary Table S4. For subsequent analyses the cut-off value $p < 10^{-4}$ was used throughout. With this threshold, an average of 5% (between 3207 and 11,771) of all tags in a sample showed differential expression, two orders of magnitude more than expected by chance alone. The highest number of differences was observed for the comparison between genotype pools A1 and A2 one day after infection (A1-T1 vs. A2-T1), and the lowest between genotype pools A2 and B2 2 days after infection (B2-T2 vs. A2-T2) (Supplementary Table S4). The number of differences in response to infection and between genotype pools even before infection (T0), was in the same order of magnitude. In all comparisons, approximately one half of the significant tags was differentially regulated in the opposite direction to the other half (up and down compared to T0, higher in pool A compared to B and vice versa).

Forty three percent of the tags (average 67,800) matched to approximately 30,000 transcripts per sample, of which an average of 11% were differentially expressed (Supplementary Table S4).

Venn-diagrams of significant tags in the comparisons over the infection time course and between genotype pools (Figure 2) showed that up to 50% of the differential tags were comparison specific. This category of tags likely included the largest portion of false positive differential tags due to random biological variation of the leaf transcriptome. On the other hand, the 2476 (T2), 3741 (T1), and 4011 (T0) differential tags specific for the comparisons between pools A1 and A2 included tags differentially expressed due to the fact that pools A1 and A2 originated from different crosses and/or differed in MCR values. Among the 290 (T2), 873 (T1), and 4682 (T0) differential tags specific for the comparisons between pools A2 and B2 should be those tags that are the consequence of the contrasting genotype at the *StAOS2* locus and/or the different MCR levels. Between 600 and 4400 tags were differentially expressed in two or three comparisons. Tags differentially expressed in multiple comparisons are more likely to be functionally significant. We therefore, combined pair wise comparisons according to various rationales (Figure 1 and below), in order to filter the data for transcripts that possibly play a functional role in the compatible interaction with *P. infestans* and in MCR.

**P. infestans tags for monitoring infection**

The six comparisons of pools A1, A2, and B2 before (T0) and after infection (T1, T2) (Figure 1) were filtered for tags matching *P. infestans* transcripts, which were absent at T0 and induced upon infection in all three genotype pools. Tags matching to *P. infestans* transcripts, which were already present at T0 were not considered authentic. They likely corresponded to transcripts of potato or non-pathogenic microorganisms colonizing potato leaves, which share high sequence similarity with transcripts of *P. infestans*. At T1 the frequency of *P. infestans* tags was negligible (not shown). At T2 sixty six tags fulfilled the selection criteria (Supplementary Table S5). The majority of these tags matched *P. infestans* ribosomal genes and other components of the translation machinery (translation elongation...
factor 1-alpha, nascent polypeptide-associated complex subunit alpha-like). The second most frequent class corresponded to hypothetical proteins. Among the rest, tags matching to the elicitors NPP1 (necrosis-inducing Phytophthora protein 1 (Kanneganti et al., 2006) (tag 9 in Supplementary Table S5), surface glycoprotein elicitor (Kamoun et al., 1997) (tag 63 in Supplementary Table S5) and elicitin-like INF7 (Kamoun et al., 1999) (tag 27 in Supplementary Table S5) were observed. Most importantly, the normalized tag counts (tpm, tags per million) of 90% of the P. infestans tags were lower in pool A2 as compared to both pools A1 and B2 (Figure 3, Supplementary Table S5), consistent with the resistance ranking of pools A1, A2, and B2, with pool A2 being the most resistant (Table 1).

**Differential expression of alleles at the StAOS2 locus**

We tested whether SuperSAGE tags corresponding to the two StAOS2 haplotypes used for constructing the genotype pools A1, A2, and B2 were detectable and whether expression differences could be observed. Filtering Supplementary Table S2 for tags matching to the StAOS2 locus on potato chromosome X1 (PGSC0003DMG400001149) yielded seven tags, four of which originated from secondary restriction sites, resulting from incomplete digestion of the primary restriction site upstream of the poly(A) tail during SAGE library construction (Figure 4C). The frequencies of the secondary tags were very low and no significant differences between samples were detected (not shown). One of the three tags derived from primary restriction sites (Figure 4, Supplementary Table S6) was detected at low frequency exclusively in pool B2 (Figure 4B). This tag therefore, corresponded to haplotype StAOS2\_G691C692, which was present in triple dosage in B2 plants but absent in A1 and A2 plants. The second and third tag were allelic variants that differed by only one nucleotide at position 23 (Supplementary Table S6, Figure 4C). The tag with the highest frequency was present in all three genotype pools as was the haplotype StAOS2\_A691C692 (Figure 4A), whereas the other was inconsistently detected at very low frequency (tpm < 3) only in pools A1 and B2 (Supplementary Table S6). The frequency of the tag corresponding to haplotype StAOS2\_A691C692 was significantly higher in pool A2 as compared to both pools A1 and B2 prior to infection (T0), consistent with the observed MCR ranking of the three pools. At T1 the frequency was significantly lower in pool B2 compared with both pools A1 and A2, corresponding to the expected ranking of the three pools according to the StAOS2 genotype. At T2 the differences between genotype pools had diminished (Figure 4A, Supplementary Table S6). The expression patterns of these tags demonstrated that (i) SuperSAGE was able to detect the expression of different alleles of the same locus and (ii) the two StAOS2 haplotypes used for MAS were indeed differentially expressed prior to infection and one day after infection.

**Candidate genes for different MCR levels resulting from the contrasting genotype at the StAOS2 locus**

Tags differentially expressed specifically between pools A2 and B2 but not between pools A2 and A1 can result from the pool’s contrasting genotypes at the StAOS2 locus and/or the different MCR values. The differential tags specific for the three comparisons between pools A2 and B2 at T0, T1 and T2 (Figure 1) were filtered for tags differentially expressed at two or three time points. This resulted in 93 tags, mostly with rather low expression levels (Supplementary Table S7), of which 25 could not be annotated, neither by matching to expressed sequences nor to annotated genes in the potato genome sequence nor by comparing matching transcripts to the NCBI nucleotide collection. Nine tags had ambiguous annotations. Eighty five of the 93 tags were differentially expressed prior to infection, indicating that the genotypic differences between pools A2 and B2 led to constitutive changes in the leaf transcriptome. Thirty eight tags were expressed at higher level in the more resistant A2 pool compared to the more susceptible B2 pool. Among those were putative components of hormone signaling pathways, for example “salicylic acid-binding protein 2” (SABP2, TC201627) (Kumar and Klessig, 2003), the polypeptide hormone “rapid alkalinization factor 2” (RALF2, TC205054) (Ryan et al., 2002), “protein phosphatase 2C” regulating ABA signaling (PP2C, TC212535) (Hubbard et al., 2010), proteins located in chloroplasts, for example “fibrillin 8” (TC207935) (Singh and McNellis, 2011) and protein “chloroplast import apparatus 2” (CIA2, PGSC0003DMG4000022983, tag 15 in Supplementary Table S7) (Sun et al., 2009), and of cellular redox homeostasis such as “glutaredoxin-C6” (Grx, TC214704) (Meyer et al., 2009). Thirty six tags were expressed at higher level in the B2 pool. These included “chloroplast epoxide hydrolase” (TC218915) functional in lipid metabolism or signaling (Guo et al., 1998), stay green protein (TC206798) involved in senescence (Hörtensteiner, 2009).
and tag 26 (Supplementary Table S7) matching to the locus PGSC0003DMG400015546 annotated in the potato genome as “Endoplasmic reticulum-type Calcium ATPase” (Wimmers et al., 1992). This tag was present at a very low level in pool B2 at all three time points but was completely absent in pool A2. The remaining 19 tags showed expression with inverted direction between time points, for example the tag matching TC208517 [putative plastid-lipid-associated protein (Langenkämper et al., 2001)]. Expression of this transcript strongly decreased from T0 to T1 in pool A2 but increased in B2.

Potato transcripts differentially regulated upon infection with compatible P. infestans isolates

Up and down regulation of transcripts in response to pathogen attack has been widely studied in plant host-pathogen interactions including potato. Filtering the SuperSAGE tags for such transcripts should therefore, expose the known up regulation of pathogenesis-related (PR) transcripts that could serve as internal controls for an activated defense response. In addition, the dynamic range and depth of digital transcript profiling should enable the discovery of novel defense related genes. The genotype pools A1, A2, and B2 represented the allelic diversity of 14, 6, and 9 half-sib (A1, A2) and full-sib (A2, B2) individuals, respectively, which originated from three heterozygous, tetraploid parental genotypes representing 12 genome wide potato haplotypes. To select transcripts up or down regulated in response to infection with P. infestans irrespective of genetic background, the comparisons over the infection time course (Figure 1) were filtered for potato transcripts that fulfilled the following criteria: the normalized hit counts (hpm, hits per million; hit counts are the sum of the counts of all tags matching to the same transcript) showed a difference between T0 and T1 and/or between T0 and T2 in the same direction (up or down) in all three genotype pools. Hit counts instead of tag counts were used to reduce the redundancy in the expression data, as in most cases two or more tags matched to the same transcript. Consistent up or down regulation in three different genotype pools represented a stringent criterion for reproducibility. The tags that did not match to any transcript present in the SGTI expressed sequences database (“no hits”) were filtered with the same criteria. The filtering resulted in 1536 potato transcripts (Supplementary Table S8) and 498 “no hit” tags (Supplementary Table S9), which were differentially expressed upon infection with compatible P. infestans isolates. These total 2034 expressed sequences were transcribed from approximately 1830 potato genes, of which 6% were not annotated and 12% were not identified in the potato genome sequence. Tags matching known PR genes such as PR1b, PR10, beta-1,3 glucanases (PR2), several chitinases (e.g., PR3), and thaumatin-like genes (PR5) (Van Loon et al., 2006) were strongly induced (Supplementary Table S8). Two hundred and seventy six “no hit” tags (55%) matched to the potato genome sequence (allowing for three mismatches), 80 of those to intergenic regions. Closer inspection of these intergenic regions in the potato genome revealed that 48 such tags either corresponded to non-annotated genes, as indicated by expressed sequences matching to the corresponding genomic region (genome browser at http://solanaceae.plantbiology.msu.edu/pgsc_download.shtml) or were located in the 3′-untranslated region of annotated genes. Interestingly, 32 tags remained that appeared truly located in intergenic regions. Most of these were up regulated upon infection with P. infestans (Table 4, Supplementary Table S9). Twenty five percent of all differentially regulated transcripts had
no meaningful functional annotation, and 4.6% had ambiguous annotations, meaning that the tags matched several transcripts with different annotations or transcripts with inconsistent annotation in the SGTI expressed sequences database and the matching locus in the potato genome (Table 4).

**Potato transcripts differential between genotype pool A2 vs. both pools A1 and B2**

Comparisons between genotype pools at the same time point were performed in order to identify candidate genes responsible for the different MCR levels of the genotype pools. The rationale for data filtering was as follows: Besides from random biological variation, differential transcript levels between genotype pools A1 and A2 could result from the different MCR levels of these pools as well as from the fact that A1 and A2 genotypes originated from different cross combinations. On the other hand, differential transcript levels between pools B2 and A2 can be the consequence of the contrasting genotype at the StAOS2 locus and/or the higher MCR in pool A2 as compared to B2 (see above). Transcripts expressed at higher or lower level in pool A2 as compared to both pools A1 and B2 are therefore, candidates for playing a role in MCR, irrespective of the genotype at the StAOS2 locus. To identify such transcripts, six comparisons between pool A1 vs. A2, and B2 vs. A2, at the infection time points T0, T1, and T2 (Figure 1) were filtered for differential transcripts. To reduce random differences further, hpm had to be significantly different in at least five of the six comparisons. The “no hit” tags were filtered with the same criteria. Under these conditions, 629 potato transcripts (Supplementary Table S10) and 177 “no hit” tags (Supplementary Table S11) were selected, which were transcribed from approximately 720 genes. Around 19% of these genes were either not annotated or not detected in the potato genome sequence. One half of the “no hit” tags matched to the potato genome sequence, 14 of those to intergenic regions. Twenty percent of all transcripts could not be functionally annotated and 7.3% had ambiguous annotations (Table 4).

In order to prioritize further the total 806 transcripts with respect to their potential relevance for MCR, additional selection criteria were applied. First, the transcripts were examined for an expression pattern corresponding to the observed MCR values, that is higher or lower expression in genotype pool A2 when compared to both pools A1 and B2. Prior to infection (T0), 75% of the transcripts showed this expression pattern, to which we refer subsequently as the “MCR conform expression pattern.” This number dropped to 51% and 54% at T1 and T2, respectively. MCR conform expression patterns at all three time points were observed for 112 transcripts (highlighted green in Supplementary Tables S10 and S11). Three hundred and fifty two and 256 transcripts were expressed at higher and lower level, respectively, in uninfected plants (T0) of pool A2 as compared to both pools B2 and A1. Two days after infection (T2), 245 transcripts were expressed at higher level in pool A2 as compared to both pools B2 and A1, and 196 transcripts were expressed at lower level. Expression patterns at T1 were more heterogeneous due to transient up or down regulation of transcripts. Second,

| Function category<sup>a</sup> | No. of potato differential transcripts after infection in genotype pools A1, A2, and B2<sup>b</sup> | No. of potato differential transcripts between genotype pools A1, A2, and B2<sup>c</sup> |
|-------------------------------|-------------------------------|-------------------------------|
| **Unknown function**           | 505 (332 ↑, 173 ↓)             | 166 (16)                       |
| **Biotic/abiotic/oxidative stress, pathogenesis, defense** | 289 (207 ↑, 82 ↓) | 91 (14) |
| **Chloroplast processes, photosynthesis, carbon fixation** | 190 (21↑, 169 ↓) | 107 (24) |
| **Membrane and transport proteins** | 137 (91 ↑, 46 ↓) | 36 (8) |
| **Transcription** | 90 (63 ↑, 27 ↓) | 27 (4) |
| **Protein biosynthesis** | 72 (37 ↑, 35 ↓) | 84 (9) |
| **Protein degradation** | 68 (54 ↑, 14 ↓) | 33 (5) |
| **Protein conformation, chaperone** | 42 (27 ↑, 15 ↓) | 19 (2) |
| **Carbohydrate metabolism** | 64 (37 ↑, 27 ↓) | 10 (1) |
| **Secondary metabolism** | 59 (27 ↑, 32 ↓) | 10 (0) |
| **Cell wall** | 57 (23 ↑, 34 ↓) | 24 (2) |
| **Development, regulation, signaling** | 58 (34 ↑, 24 ↓) | 15 (0) |
| **Lipid metabolism** | 54 (34 ↑, 20 ↓) | 14 (2) |
| **Central metabolism** | 48 (29 ↑, 19 ↓) | 29 (3) |
| **Hormone metabolism/signaling** | 44 (27 ↑, 17 ↓) | 14 (3) |
| **Amino acid metabolism** | 39 (22 ↑, 17 ↓) | 15 (3) |
| **Other** | 257 (171 ↑, 86 ↓) | 92 (15) |
| **Intergenic regions** | 32 (25 ↑, 7 ↓) | 14 (1) |
| **No match in potato genome** | 226 (141 ↑, 85 ↓) | 88 (6) |
| **Ambiguous annotation** | 93 (44 ↑, 49 ↓) | 67 (9) |

**Table 4 | Grouping of differential transcripts according to putative function category.**

<sup>a</sup>Words in bold letters can be used to text filter Supplementary Tables S8–S11 in columns “putative function,” “chromosome” (no match) and “locus” (intergenic).

<sup>b</sup>The number of up (↑) and down (↓) regulated transcripts are shown in parenthesis.

<sup>c</sup>The number of transcripts showing the MCR conform expression pattern at all time points are shown in parenthesis.

281 transcripts were selected, which showed consistent differential expression after infection with *P. infestans* as well as between genotype pools (Supplementary Table S12).

**FUNCTIONAL CATEGORIES OF DIFFERENTIAL TRANSCRIPTS**

Based on the annotations in the DFCI potato gene index, of the matching loci in the potato genome and in few cases by BLAST searches against the NCBI database (http://blast.ncbi.nlm.nih.gov/blast.cgi), transcripts that showed consistent differential expression over the infection time course or between
genotype pools A1, A2, and B2, were grouped in 17 functional categories, with overlaps between categories (Table 4, Supplementary Tables S8–S11). The largest group of differential tags/transcripts did not allow any function assignment. It included, besides tags derived from annotated and non-annotated genes of unknown function, 226 tags that did not match to the potato genome sequence. Many but not all of these tags contained A stretches originated from the poly(A) tail of mRNA, which prevented genomic matching. Non-matching tags without A stretches may derive from genes not covered by the draft potato genome sequence or from genes highly divergent between *S. tuberosum* Group *phureja* (genome sequence) and *S. tuberosum* Group *tuberosum*, from which the samples originated. The largest groups of transcripts with functional annotation were related to biotic as well as abiotic stress responses and to chloroplastic processes. The majority of transcripts were up regulated upon infection with *P. infestans*, except transcripts functional in the chloroplast, which were mostly down regulated. Details on the individual functional categories are provided as Supplementary material (Supplementary Results 3.1–3.17).

**CO-LOCALIZATION OF DIFFERENTIALLY EXPRESSED GENES AND RESISTANCE QTL**

Overlapping genomic positions of candidate genes with resistance QTL can be used as additional criterion to rank a large number of functional candidates. QTL linkage mapping for resistance to late blight and other pathogens of potato has been performed in several genetic backgrounds using as markers restriction fragment length polymorphisms (RFLP’s) of potato and tomato (Gebhardt and Valkonen, 2001; Simko et al., 2007; Danan et al., 2011). Resistance QTL were anchored to the potato genome sequence by positioning linked RFLP markers on the physical maps of the 12 potato chromosomes (Xu et al., 2011) (Figure 5). Due to variable ratios between genetic and physical map distances (Xu et al., 2011), sizes and positions of resistance QTL on the physical maps deviated from previous consensus maps for resistance QTL based on genetic linkages (Gebhardt and Valkonen, 2001; Danan et al., 2011) [see also the Solanaceae (SOL) function map for pathogen resistance at http://www.gabipd.org/database/maps.shtml]. The genomic positions of transcripts that showed the MCR conform expression pattern at all time points and matched with high confidence to unique loci are shown on the physical map (Figure 5). Several of these transcripts matched to novel loci, not annotated in the potato genome, and one matched to an intergenic region on chromosome II. Co-localization with resistance QTL was observed for a number of loci. For example, the central QTL for late blight resistance on chromosome III (Pin3B and lhb3 on the SOL function map for pathogen resistance at http://www.gabipd.org/database/maps.shtml) extended over approximately 3 Mbp. This region contains the *StKI* locus, which consists of mixed clusters of protease inhibitors genes (Odeny et al., 2010). The inhibitor encoded at the locus PGSC0003DMG400010170 was expressed consistently at higher level in pool A2, whilst the inhibitor encoded at the tightly linked locus PGSC0003DMG400010144 was expresses at lower level in pool A2 compared with both pools A1 and B2 (Figure 5, Supplementary Table S10).

**SEMIQUANTITATIVE REAL TIME PCR (qRT-PCR) OF SELECTED CANDIDATE GENES**

qRT-PCR was performed with one *P. infestans* transcript and 15 potato transcripts, in order to assess the reproducibility of differential expression in independent infection experiments. (Figure 6). The genomic positions of the corresponding potato loci are included in Figure 5. The transcript for the ribosomal protein L23a of *P. infestans* (Figure 6A) and potato transcripts known to be up (osmotin, 1,3-beta-glucan glucanohydrolase, lipoxigenase) (Ros et al., 2004; Restrepo et al., 2005; Gyetvai et al., 2012) or down regulated (2-oxoglutarate-dependent dioxygenase) (Restrepo et al., 2005) in response to infection by a compatible *P. infestans* isolate served as controls (Figures 6F–I). The remaining transcripts were chosen based on novelty according to their annotation, strong differential expression upon infection (log-fold change ≥ 2) and differences between genotype pools. The strong increase in *P. infestans* L23a transcripts after 2 days confirmed successful infection in the three experiments analyzed (Figure 6A). Up or down regulation as quantified by normalized hit counts (hpm) in SuperSAGE was reproducible for all transcripts tested in cDNAs of experiment 1, the same as used for generating the SuperSAGE samples. High relative expression levels in qRT-PCR did however, not correspond to high hpm counts in SuperSAGE and vice versa, indicating that both methods for quantifying transcript levels were qualitative but not quantitative comparable. Also the relative ranking of the three genotype pools based on hpm at the three time points was in 40% of the cases not congruent with the ranking observed in qRT-PCR. A particularly intriguing example was transcript TC221793 (Supplementary Table 10), which had the highest hit counts in genotype pool A1 compared with pools A2 and B2, but was not detectable by qRT-PCR in pool A1, whereas the lower transcript level in pool A2 vs. B2 was observed in all three experiments (Figure 6L). In experiment 2 and 3, induction or repression was generally reproducible, except for transcripts TC195330 and TC195401, which showed inconsistent results in one of three experiments (Figures 6O,P). As observed previously (Gyetvai et al., 2012), relative expression levels and time courses were highly variable between experiments. The most reproducible expression pattern showed transcript TC197025 (Figure 6B) annotated as “kiwellin,” which was originally described as allergen from Kiwi fruits (Tamburrini et al., 2005). This transcript was strongly up regulated after infection in all genotype pools and raised to higher levels in pool A2 compared to both pools A1 and B2. The MCR conform expression pattern was also reproducibly detected at T0 for transcripts TC225046 (Figure 6C), TC208021 (Figure 6D), TC218213 (Figure 6E) and TC209317 (Figure 6J).

**DISCUSSION**

**MAS FOR MATURITY CORRECTED RESISTANCE TO LATE BLIGHT**

To the best of our knowledge, the MAS experiment described here is the first where, based on diagnostic SNP markers, plants with improved quantitative resistance to late blight not compromised by late maturity were selected. The results show that the predictive power of the two *StAOS2* SNP markers depends on the genetic background, in which MAS is performed. Genotypic selection
in two half-sib families based on two \textit{StAOS2} haplotypes, one expected to have positive and the other negative effects on MCR, resulted in genotype groups A1, A2, and B2. Groups A1 and A2 were homozygous for the positive \textit{StAOS2} haplotype, whereas group B2 carried one copy of the positive and three copies of the negative haplotype. As expected, group A2 was more resistant than group B2. Unexpected, group A1 was as susceptible as group B2 (Table 1). The reason for the observed difference in MCR
FIGURE 6 | Continued
level between groups A1 and A2 may be the missing genotypic information at additional, unknown loci that also influence MCR. The two SNPs at the StAOS2 locus explained approximately one third of the genetic variation of MCR (Pajerowska-Mukhtar et al., 2009), which means that two thirds of the genetic variation were due to other loci. The alleles present at these additional loci might have been different in parents Phy14 and Phy16, thereby modifying or even suppressing the effects of StAOS2 alleles on MCR depending on the genomic context. Moreover the haplotypes StAOS2_A, G haplotype present only in B2 plants and StAOS2_G were not uniform with respect to additional SNPs at the StAOS2 locus. This was indicated by finding two allelic tags corresponding to haplotype StAOS2_A (Results Section Differential expression of alleles at the StAOS2 locus). Multiple, segregating StAOS2 alleles may modulate the phenotypic effects of the specific SNPs of the same plant tissues as used to construct the SuperSAGE libraries. Relative expression in genotype pools A1, A2, and B2 is shown as green, blue, and pink bars, respectively. The normalized hit counts of each transcript are indicated on top of each bar (numbers rounded). The middle and right panels show the results obtained in experiment two and three. Error bars were derived from two technical replicates. They represent the maximum deviation from the mean.

StAOS2 transcript levels are a limiting factor for the production of jasmonates (Kombrink, 2012). Higher StAOS2 transcript levels in uninfected plants may lead to higher constitutive amounts of one or more products of the jasmonate pathway (Göbel et al., 2002), consequently to increased jasmonate signaling and to higher constitutive expression of jasmonate responsive genes (Reymond and Farmer, 1998), which could influence quantitative pathogen resistance. Indeed, higher transcript levels prior to infection in pool A2 as compared to pools B2 and A1 were observed for known jasmonate responsive genes such as lipoygenases (TC196405, TC204989) and chitinases (TC194776, TC198089, TC220862, TC222828, Supplementary Table S10). This model is consistent with functional analyses of potato and rice AOS genes in transgenic plants. AOS silencing led to increased susceptibility and AOS overexpression to increased resistance (Mei et al., 2006; Pajerowska-Mukhtar et al., 2008). Some of the differential transcripts specific for the comparisons between genotype pools A2 and B2 (Supplementary Table S7) might be components of a regulatory network in which StAOS2 participates, particularly those that show, like StAOS2, differential expression prior to infection. This set of transcripts corresponded mostly to genes of unknown function or with very unspecific annotation, most of them expressed at low levels. Annotated genes had a putative function in hormone, lipid, and calcium signaling, proteasome mediated protein degradation, transport and in chloroplastic processes, for example fibrillin 8 (TC207935, Supplementary Table S7). Fibrillin may have a role in that part of jasmonate biosynthesis which takes place in the chloroplast (Singh and McNellis, 2011). The same transcript was down regulated in compatible interactions of potato with P. infestans based on transcript profiling by DeepSAGE (Gyetvai et al., 2012). The expression patterns of the tags matching to fibrillin 8 were similar in both transcript profiling experiments.

DIFFERENTIAL EXPRESSION OF StAOS2 ALLELES AND PUTATIVE JASMONATE RESPONSIVE TRANSCRIPTS

The sensitivity of digital transcript profiling by SuperSAGE was demonstrated by the differential tags corresponding to the StAOS2 alleles that were used to select A1, A2, and B2 plants. The negative StAOS2_G haplotype present only in B2 plants was expressed at much lower level than the positive haplotype StAOS2_A. For both haplotypes no differential regulation in response to infection was observed. Moreover, StAOS2-G showed the MCR conform expression pattern in uninfected plants (Figure 4). This expression pattern suggests that
sampling strategy was demonstrated by observing the expected expression patterns of well known pathogenesis related genes, which served as internal controls besides \textit{P. infestans} transcripts. The compatible interaction activated not only responses to biotic but also to abiotic stress, indicating that the distinction between biotic and abiotic stress is rather artificial and results from the way scientists study plant-environment interactions rather than from naturally distinct response pathways. The differential expression of components of most known hormone signaling pathways also supports the view that plants react to a compatible \textit{P. infestans} infection with a global stress response. A clear symptom of senescence and beginning cell death was the down regulation of numerous transcripts for chloroplast processes, particularly for photosynthesis and chloroplastic protein biosynthesis, which was observed already 1 day after infection. Down regulation of photosynthesis has been observed in many transcriptome studies and is an unspecific plant reaction to pathogen attack (Bilgin et al., 2010).

**Differential Expression Between Genotype Pools with Different MCR Levels**

The 806 transcripts from approximately 720 genes, which showed most consistent differential expression between the genotype pools A1, A2, and B2 are considered as primary candidates for playing a functional role in MCR (Supplementary Tables S10, S11). This type of differential expression has not been analyzed before. Top candidates among those might be the genes that showed the MCR conform expression pattern before as well as after infection with \textit{P. infestans}, and the genes that were induced or repressed upon infection in addition to showing differences between genotype pools (Supplementary Table 12). However, the 806 candidate genes include an unknown portion of false positives. Transcript levels differing between genotype pools were less reproducible in independent infection experiments than transcript induction or repression upon infection, as assessed by qRT-PCR for a small number of candidates. Best reproducibility was achieved for differences present in uninfected plants, likely due to the variability of \textit{P. infestans} infection kinetics. On the other hand, sensitivity and resolution of qRT-PCR might be insufficient for validation of quantitative differences between low expressed, sometimes allele specific transcripts detected by SuperSAGE.

Differential expression originates either directly or indirectly from DNA variation in genes causal for MCR. The former possibility has to be distinguished from the latter when aiming at the identification of diagnostic markers for breeding applications. Diagnostic DNA markers have to reside within the genes causal for MCR or be in linkage disequilibrium with the causal genes. The identification of such markers requires association mapping, as exemplified for \textit{Sta002} (Pajerowska-Mukhtar et al., 2009). The genes showing differential expression between genotype pools A1, A2, and B2 set the stage for the discovery of additional diagnostic markers for MCR and of genes underlying quantitative resistance to late blight. The confirmation of a function in pathogen resistance of individual candidate genes requires further studies such as gene silencing and complementation analysis of corresponding mutants in the model plant Arabidopsis (Pajerowska-Mukhtar et al., 2008). Such studies can be practically performed only with a small number of genes. The physical map for resistance QTL (Figure 5) provides an auxiliary tool for choosing the most interesting candidates for further analysis, that is the co-localization of candidate gene and resistance QTL. The criterion of co-localization is not exclusive, as QTL delimitations are not precise and not all potato QTL are known. Whole genome association mapping (Zhu et al., 2008) using genome wide SNP’s (Hamilton et al., 2011) will be another option for choosing candidates based on co-localization with peaks of marker-trait associations.

The transcripts differentially expressed between genotype pools A1, A2, and B2 were only partially identical with the transcripts differentially expressed upon infection (Table 4, Supplementary Table S12). This shows that MCR is not only controlled by genes induced or repressed upon infection but also by constitutively expressed genes. The majority of the 806 transcripts showed the MCR conform expression pattern prior to infection, that is higher or lower expression in genotype pool A2 compared to both pools A1 and B2. This also indicates that part of the quantitative resistance to late blight is constitutive. Genes with genotype dependent, constitutive differential expression provide excellent targets for developing novel diagnostic tools for breeding cultivars with improved quantitative resistance to late blight and possibly other biotic and abiotic stresses. Approximately 200 of 266,361 unitags matched to 90 \textit{R} genes or \textit{R} gene homologs, of which only two were differentially expressed between genotype pools, based on our stringent filtering criteria. Considering that the potato genome contains 438 putative \textit{R} genes (Jupe et al., 2012), most of these genes were not expressed in leaves before and after infection, or their expression was below detection level. This class of genes does not seem to play a dominant role in MCR, at least not at the transcript level. Genes with roles in downstream defense signaling and defense responses were present more frequently among the 806 candidates. In addition to these, our comparative transcript profiling experiment identified novel candidate genes for MCR in general cellular processes such as photosynthesis, protein biosynthesis, protein degradation via the 26S proteasome, transport of proteins, lipids, ions and other small molecules, in structural cellular components such as the cell wall and the cytoskeleton. Resistance to late blight not compromised by late maturity might be associated with a generally more "healthy" physiological state of the plant, which involves most aspects of metabolism. The MCR conform expression pattern was also observed for numerous genes of unknown or ill-defined function. One such gene was a potato homolog of the allergen “kiwellin” in Kiwi fruits (Tamburrini et al., 2005; Ciardiello et al., 2008; Tuppo et al., 2008). The expression of this gene was more strongly up regulated upon infection in the A2 pool than in both the A1 and B2 pools in three infection experiments (Figure 6B). This expression pattern makes it a potential transcriptional marker for MCR. The MCR conform expression pattern was also observed for some tags matching to intergenic regions. The sequences in these regions require further analysis.
They might encode microRNA precursor transcripts (Cuperus et al., 2011; Li et al., 2012) or novel genes, which are neither annotated in the potato genome nor represented in expressed sequence databases.

CONCLUSIONS
1. Based on diagnostic SNPs in the StAOS2 gene, it was possible to select potato plants with increased resistance to late blight not compromised by late maturity. Additional diagnostic markers are required to improve predictability of MCR independent from genetic background.
2. Comparing by SuperSAGE the expression profiles of genotype pools before and after infection with P. infestans yielded a large number of novel candidate genes for quantitative resistance to late blight. They constitute a rich resource for marker development in breeding applications as well as for functional studies, which may unravel molecular mechanisms governing quantitative plant pathogen resistance.
3. Generally, induction or repression of selected transcripts upon infection with P. infestans was reproducible by qRT-PCR in three independent infection experiments. Differential expression between genotype pools was in part reproducible by qRT-PCR, particularly differences prior to infection.
4. Comparative expression profiling by SuperSAGE revealed quantitative differences between StAOS2 alleles prior to infection, which suggest a mechanism for different resistance levels caused by natural variation at the StAOS2 locus.
5. Maturity-corrected resistance to late blight is, in part, constitutive. This was shown by detecting numerous differential transcripts in uninfected plants showing different levels of MCR.
6. Maturity-corrected resistance to late blight involves, besides biotic and abiotic stress responses mediated by hormone signaling, cellular housekeeping processes like photosynthesis, protein biosynthesis and degradation, transport of proteins, lipids, ions and other small molecules.

ACKNOWLEDGMENTS
This work was funded by the German Ministry for Research and Education under the GABI (Genome analysis in the biological system plant) program (grant no. 0315065, PAPATOMICS) and by the Max Planck Society. Part of the work was carried out in The Department for Plant Breeding and Genetics, headed by Maarten Koornneef. The authors thank Maarten Koornneef and Imre Somssich for critically reading the manuscript.

SUPPLEMENTARY MATERIAL
The Supplementary Material for this article can be found online at: http://www.frontiersin.org/journal10.3389/fpls.2013.00423/abstract

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Conflict of Interest Statement: The 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.

Received: 17 June 2013; accepted: 05 October 2013; published online: 14 November 2013.

Citation: Draffehn AM, Li L, Krezdorn N, Ding J, Lübeck J, Strahwald J, Muktar MS, Walkemeier B, Rotter B and Gebhardt C (2013) Comparative transcript profiling by SuperSAGE identifies novel candidate genes for controlling potato quantitative resistance to late blight not compromised by late maturity. Front. Plant Sci. 4:423. doi: 10.3389/fpls.2013.00423

This article was submitted to Plant Genetics and Genomics, a section of the journal Frontiers in Plant Science.

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