

New Phytologist Supporting Information

Article title: Elevated CO₂ alters soybean physiology and defense responses, and has disparate effects on susceptibility to diverse microbial pathogens

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Table S6 Significantly differentially expressed genes (DEGs, FDR < 0.01) responding to *P. syringae* pv. *glycinea* (Psg) infection (6 hours) in plants grown in elevated (eCO₂, 550 ppm) or ambient CO₂ (aCO₂, 419 ppm) conditions.

Table S7 Gene ontology (GO) biological process (BP) terms significantly overrepresented among differentially expressed gene clusters comparing *P. syringae* pv. *glycinea* (Psg) response (Psg v. Mock) in elevated (eCO₂, 550 ppm) and/or ambient (aCO₂, 419 ppm) CO₂ conditions (419 ppm).

Table S8 Significantly differentially expressed genes (DEGs, FDR < 0.01) responding to changes in CO₂ (elevated (eCO₂) 550 ppm or ambient (aCO₂) 419 ppm) in *P. syringae* pv. *glycinea* (Psg)-infected (6 hours) and/or mock-infected leaves.

Table S9 Gene ontology (GO) biological process (BP) terms significantly enriched in the STRING (v12) network developed from clusters of DEGs responding to elevated (550 ppm) versus ambient CO₂ conditions (419 ppm) in *P. syringae* pv. *glycinea* (Psg) and/or mock-infected leaves at 21 days.

Table S10 Annotation of significantly differentially expressed transcription factors (TFs, FDR < 0.01) responding to changes in CO₂ (elevated (eCO₂) 550 ppm or ambient (aCO₂) 419 ppm) in *P. syringae* pv. *glycinea* (Psg)-infected (6 hours) and/or mock-infected leaves.

Table S11 Identification of transcription factors (TF) with overrepresented binding sites (TFBS) among the promoters of differentially expressed genes (DEG) responding to changes in CO₂ (elevated (eCO₂) 550 ppm, ambient (aCO₂) 419 ppm) in *P. syringae* pv. *glycinea* (Psg)-infected (6 hours) and/or mock-infected leaves.

Table S12 Differentially expressed gene (DEG) targets of transcription factors with overrepresented binding sites (Supplemental Table S10) responding to changes in CO₂ (elevated (eCO₂) 550 ppm, ambient (aCO₂) 419 ppm) in *P. syringae* pv. *glycinea* (Psg)-infected (6 hours) and/or mock-infected leaves.

Table S13 Gene ontology (GO) biological process (BP) terms significantly overrepresented

among differentially expressed gene (DEG) targets of transcription factors (TFs) with overrepresented transcription factor binding sites (TFBS) in expression clusters from *P. syringae* pv. *glycinea* (Psg) and/or mock-infected leaves responding to CO₂ levels (elevated=550 ppm versus ambient=419 ppm).

Methods S1 QuantSeq 3'mRNA-Seq data analysis and jasmonic acid (JA) and salicylic acid (SA) quantification.

Fig. S1 Schematic representation of growth chamber conditions.

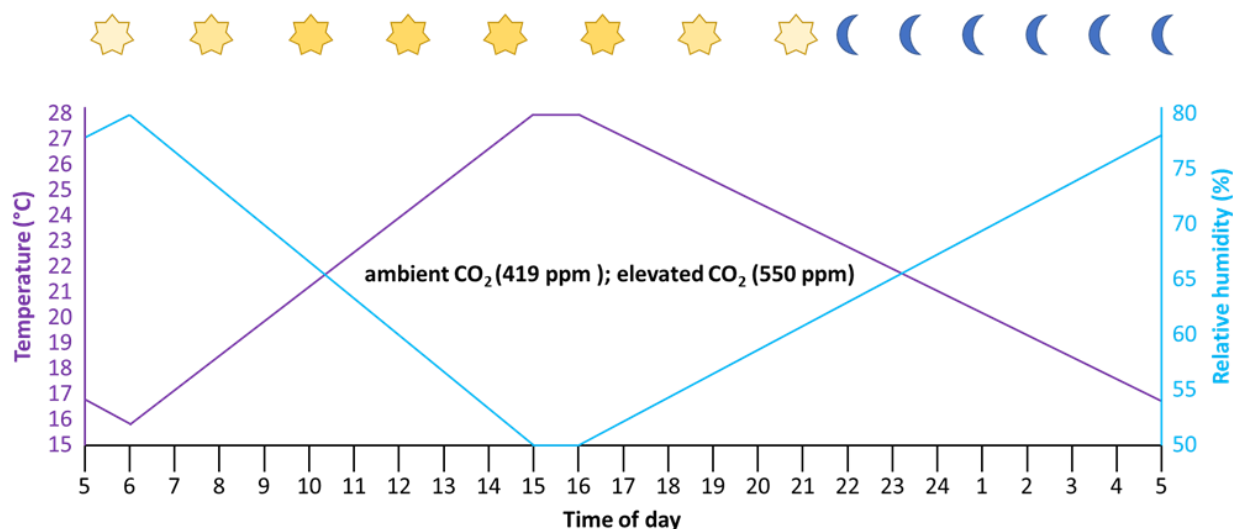


Fig. S1. Schematic representation of growth chamber conditions. Schematic representation of Envirotron growth chamber conditions. All chambers were set to the same conditions, except that CO₂ was 419 ppm in chambers set to ambient (*a*CO₂) conditions and CO₂ was 550 ppm in elevated (*e*CO₂) conditions. Lights were on at 5:38 am (05:38) and off at 8:52 pm (20:52) (sunrise and sunset times represent June 15 in central Iowa), minimum temperature was 15.6 °C at 6 am (06:00) with ramping to 27.6 °C maximum temperature from 3:00 – 4:00 pm (15:00 – 16:00). The relative humidity was ramped between 80% at 6:00 am (06:00) to the minimum of 50% at 3:00 – 4:00 pm (15:00-16:00).

Fig. S2 Stomatal index on the abaxial leaf surface of soybean.

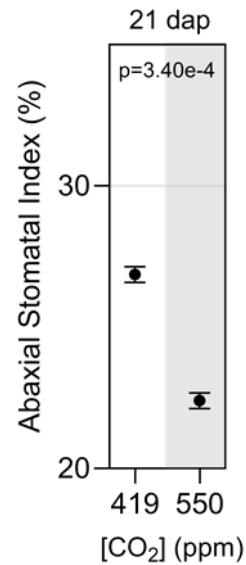


Fig. S2. Stomatal index on the abaxial leaf surface of soybean. Stomatal index measurements were performed as described in materials and methods. Three replicates were carried out simultaneously in independent CO₂ control chambers, using 5 plants per treatment for each CO₂ concentration. Stomata were counted in three randomly selected fields of view per leaf. Data points represent the mean value across the three replicates, and error bars represent the standard error. P-value was computed based on an F-test for the main effect of CO₂ in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10, and dap indicates days after planting.

Fig. S3 Effect of elevated CO₂ on growth rate of *Pseudomonas syringae* pv. *tomato* DC3000 (*PstDC3000*).

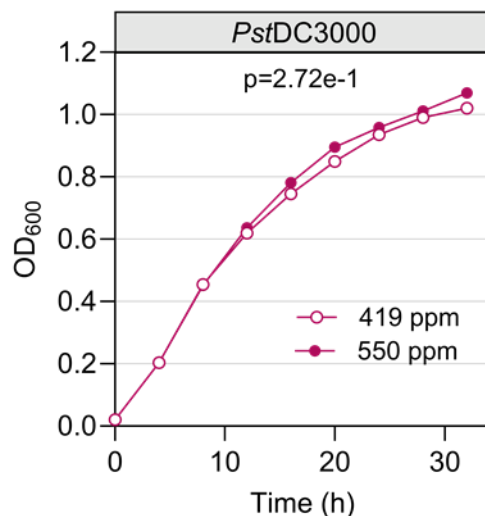


Fig. S3. Effect of elevated CO₂ on growth rate of *Pseudomonas syringae* pv. *tomato* DC3000 (*PstDC3000*). Bacterial cultures were grown in LB broth without antibiotics for 32 h and OD₆₀₀ was measured every four hours using a portable spectrophotometer. Two experimental replicates were conducted simultaneously, for each CO₂ concentration, in independent CO₂ chambers. Data points represent the average OD₆₀₀ (optical density at 600 nm) between the two chambers. P-value was computed based on an F-test for the interaction of CO₂ and time in linear mixed effect model analysis using data after 4 hours due to unequal variance. The letter e denotes an exponent to the power of 10.

Fig. S4 Accumulation of salicylic acid (SA) and jasmonic acid (JA) at 6 h post-inoculation with *Pseudomonas syringae* pv. *glycinea* (*Psg*) in $a\text{CO}_2$ and $e\text{CO}_2$.

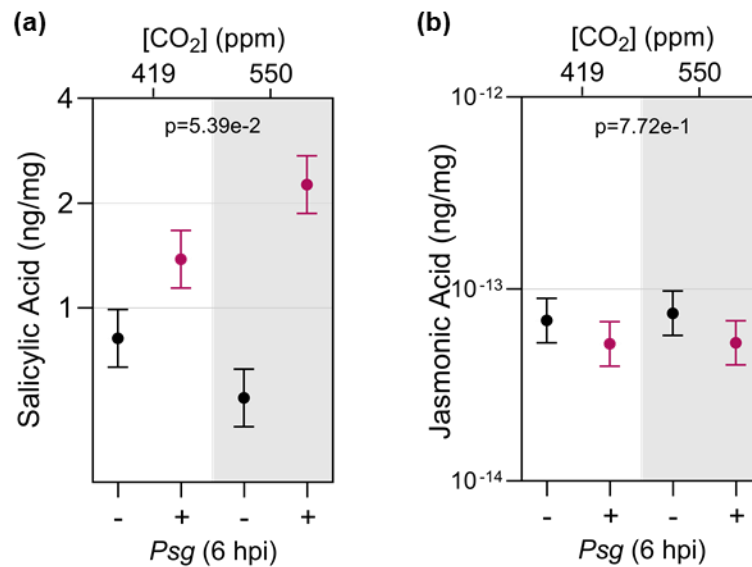


Fig. S4. Accumulation of salicylic acid (SA) and jasmonic acid (JA) at 6 h post-inoculation (hpi) with *Pseudomonas syringae* pv. *glycinea* (*Psg*) in $a\text{CO}_2$ and $e\text{CO}_2$. (a) SA and (b) JA levels were quantified using non-targeted and targeted metabolite analysis, respectively (Methods S1). Experiments were conducted independently, three times, using 8 plants per treatment for each CO_2 concentration. Data points represent the mean value across the three replicates and error bars represent the standard error. P-value was computed based on an F-test for the interaction of CO_2 and *Psg* treatment in linear mixed effect model analysis on log-transformed data. The letter e denotes an exponent to the power of 10.

Fig. S5 Hierarchical clustering of 21,822 DEGs (FDR < 0.01) responding to *Pseudomonas syringae* pv. *glycinea* (Psg) at 6 h post-inoculation in plants grown in eCO₂ (550 ppm) or aCO₂ (419 ppm) conditions.

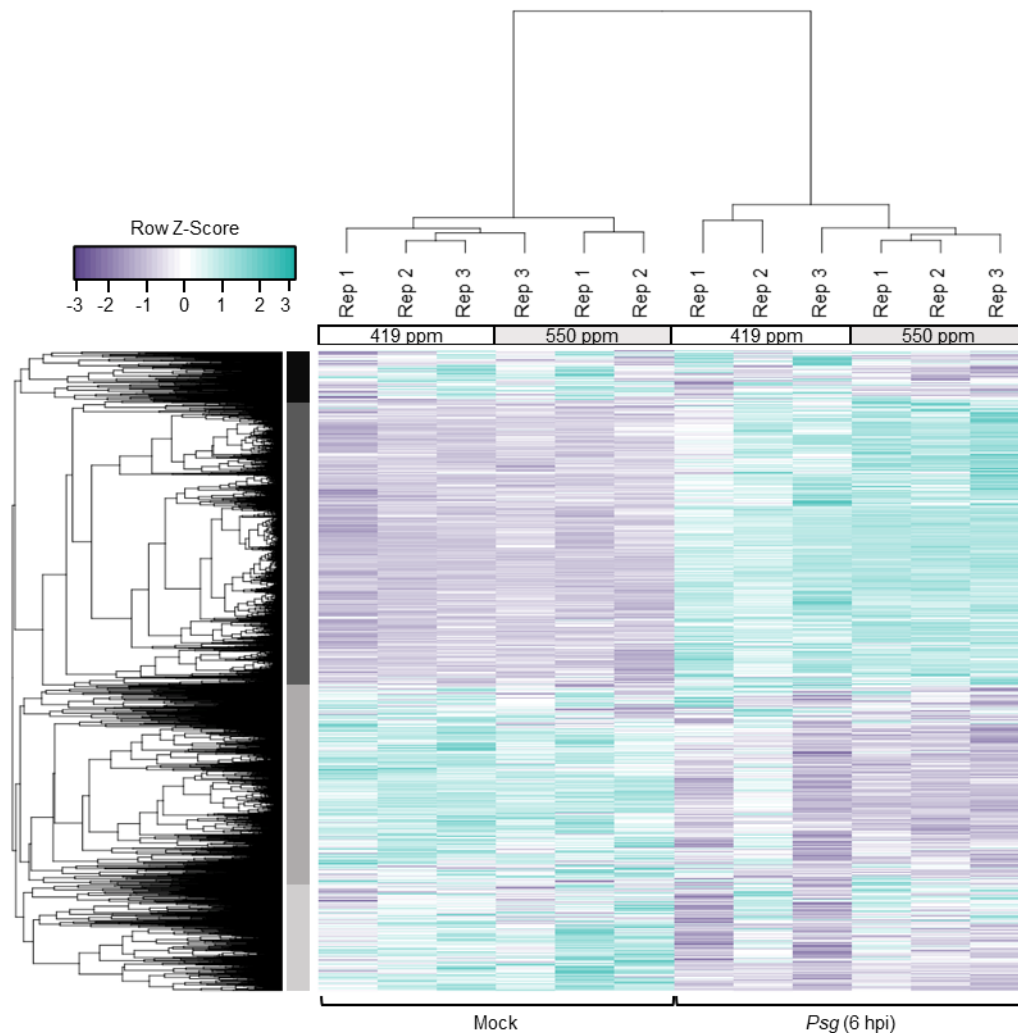


Fig. S5. Hierarchical clustering of 21,822 DEGs (FDR < 0.01) responding to *Pseudomonas syringae* pv. *glycinea* (Psg) at 6 h post-inoculation (hpi) in plants grown in eCO₂ (550 ppm) or aCO₂ (419 ppm) conditions. The full list of DEGs is provided in Table S7. Samples for 3' mRNA-Seq analysis were taken from the unifoliate leaves of 14-day old plants at 6-hpi with mock treatment or Psg. The three independent replicates were conducted using eight plants per CO₂ treatment. Rep indicates independent biological replicates. Row Z-scores were used for hierarchical clustering of DEGs, based on expression across samples and replicates. Purple indicates expression values below the row mean and teal indicates expression values above the row mean. Four different expression clusters were identified. These are indicated by differently shaded gray boxes to the left of the heat map.

Fig. S6 Effect of elevated CO₂ on the shoot fresh weight and shoot dry weight of soybean plants infected with bean pod mottle virus (BPMV) and soybean mosaic virus (SMV).

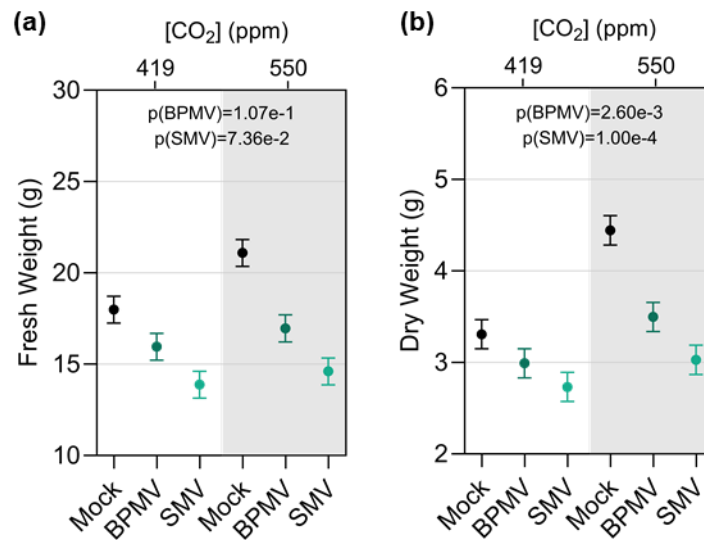


Fig. S6. Effect of elevated CO₂ on the shoot fresh weight and shoot dry weight of soybean plants infected with bean pod mottle virus (BPMV) and soybean mosaic virus (SMV). **(a)** Fresh weight and **(b)** dry weight of soybean plants at 21 days post-inoculation with BPMV, SMV, or mock treatment. Three experiments were conducted simultaneously in independent CO₂ control chambers, using 8 plants per treatment for each CO₂ concentration according to a replicated complete design. Data points represent the mean value across the three replicates and error bars represent the standard error. P-values were computed based on t-tests for comparing the difference between virus and mock under eCO₂ with that under aCO₂ in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10.

Fig. S7 Bean pod mottle virus (BPMV) and soybean mosaic virus (SMV) accumulation at 21 days post-inoculation in soybean plants growing in αCO_2 and $e\text{CO}_2$.

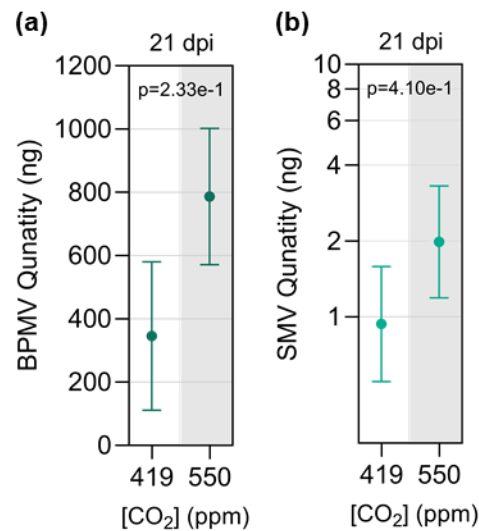


Fig. S7. Bean pod mottle virus (BPMV) and soybean mosaic virus (SMV) accumulation at 21 days post-inoculation (dpi) in soybean plants growing in αCO_2 and $e\text{CO}_2$. (a) BPMV and (b) SMV quantity was determined by RT-qPCR using multiplexed probes for BPMV or SMV and *Skp1* as an internal reference control. Samples were taken from the newest fullest expanded leaf. Three experiments were conducted simultaneously in independent CO₂ chambers, using 8 plants per infection treatment for each CO₂ concentration according to a replicated complete block design. Data points and bars represent mean group values and standard error across the three replicates. P-value for BPMV was computed based on an F-test for the main effect of CO₂ in linear mixed effect model analysis. P-value for SMV was computed based on an F-test for the main effect of CO₂ in linear mixed effect model analysis on log-transformed data. The letter e denotes an exponent to the power of 10.

Fig. S8 Effects of $e\text{CO}_2$ on root and shoot biomass of soybean plants infected with *Fusarium virguliforme*.

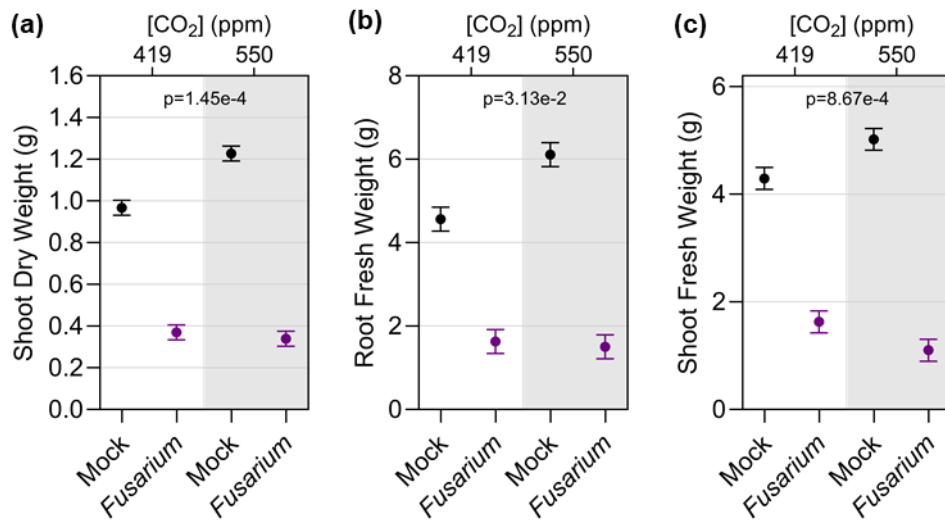


Fig. S8. Effects of $e\text{CO}_2$ on root and shoot biomass of soybean plants infected with *Fusarium virguliforme*. (a) Shoot dry weight, (b) root fresh weight and (c) shoot fresh weight at 35 days after germination in soil infested with *F. virguliforme* or mock infested. The three replicate experiments were conducted simultaneously in independent CO_2 control chambers using 6 plants per treatment for each CO_2 condition according to a replicated complete block design. Data points represent mean values with standard error across the three replicates. P-values were computed based on F-tests for the interaction effect between CO_2 and *Fusarium* treatment in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10.

Fig. S9 Expression of the JA marker gene, *KT11*, in roots of plants infected with *Fusarium virguliforme* in $a\text{CO}_2$ and $e\text{CO}_2$.

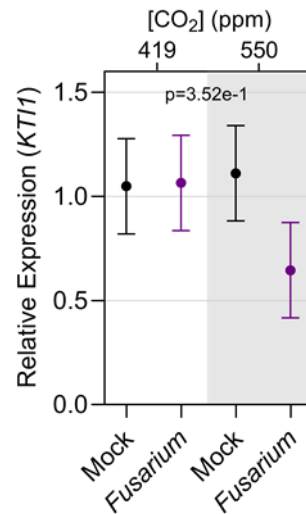


Fig. S9. Expression of the JA marker gene, *KT11*, in roots of plants infected with *Fusarium virguliforme* in $a\text{CO}_2$ and $e\text{CO}_2$. RT-qPCR analysis was conducted on RNA extracted from root tissue using gene-specific primers for the JA marker gene *KT11* with *Skp1* as an internal reference control. Experiments were conducted using 6 plants per treatment for each CO_2 concentration. Triplicate experiments were conducted simultaneously in independent CO_2 chambers. Data points and bars represent mean values and standard error across the three replicates. P-value was computed based on an F-test for the interaction effect between CO_2 and *Fusarium* treatment in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10.

Fig. S10 CO₂ concentration does not impact *in vitro* growth of *Fusarium virguliforme*.

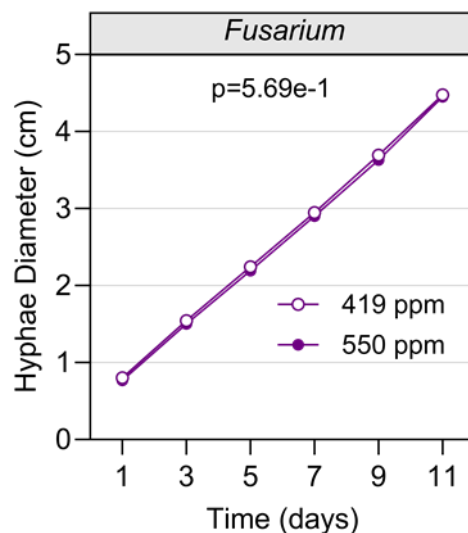


Fig. S10. CO₂ concentration does not impact *in vitro* growth of *Fusarium virguliforme*. Hyphae diameter was measured every 2 days over 11 days. Three experiments were conducted simultaneously using 10 plates per experiment, for each CO₂ treatment. Data points represent mean values across three replicates conducted simultaneously in independent CO₂ control chambers. P-value was computed based on an F-test for the interaction effect between CO₂ and time in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10.

Fig. S11 Soybean root and shoot biomass is impacted by $e\text{CO}_2$ in *Pythium sylvaticum*-infected plants.

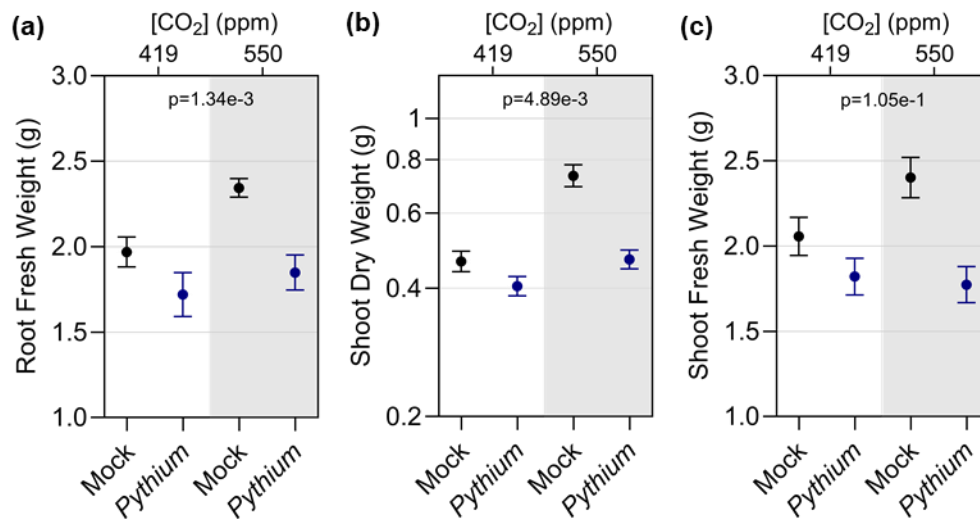


Fig. S11. Soybean root and shoot biomass is impacted by $e\text{CO}_2$ in *Pythium sylvaticum*-infected plants (a) Root fresh weight, (b) shoot dry weight and (c) shoot fresh weight in mock or *P. sylvaticum*-infected plants 21 days after inoculation. Experiments were conducted in triplicate using 6 plants per treatment, for each CO_2 condition. Triplicate experiments were conducted simultaneously in independent CO_2 controlled chambers. Data points represent mean values across the three replicates with standard error bars. P-values were computed based on F-tests for the interaction effect between CO_2 and *Pythium* treatment in linear mixed effect model analysis. Fourth power and log transformation were applied for root fresh weight and shoot dry weight respectively due to unequal variance on the original scale. The letter e denotes an exponent to the power of 10.

Fig. S12 *In vitro* growth of *Pythium sylvaticum* is accelerated under $e\text{CO}_2$.

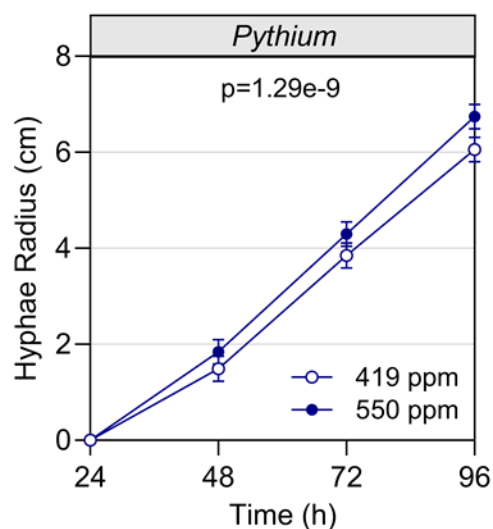


Fig. S12. *In vitro* growth of *Pythium sylvaticum* is accelerated under $e\text{CO}_2$. Experiments were conducted using 10 plates per treatment for each CO_2 condition in triplicate. The three experiments were done simultaneously in independent CO_2 controlled chambers. Hyphae radius was measured every day over 4 days. Data points and bars represent mean values and standard error across the three replicates. P-value was computed based on an F-test for the interaction effect between CO_2 and time in linear mixed effect model analysis. The letter e denotes an exponent to the power of 10.

Fig. S13 Expression of the SA marker gene, *PR1*, in roots of plants infected with *Pythium sylvaticum* in *aCO*₂ and *eCO*₂.

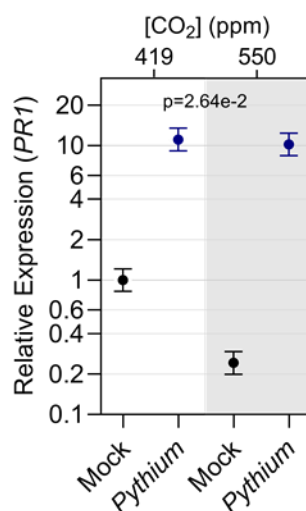


Fig. S13. Expression of the SA marker gene, *PR1*, in roots of plants infected with *Pythium sylvaticum* in *aCO*₂ and *eCO*₂. RT-qPCR analysis was conducted using RNA extracted from soybean roots 14 days after inoculation using gene specific primers for the SA marker gene *PR1* and *Skp1* was an internal reference control. Experiments were conducted simultaneously in three independent CO₂ controlled chambers using 6 plants per treatment for each CO₂ concentration according to a replicated complete block design. Data points represent mean values across the three replicates with standard error. P-value was computed based on an F-test for the interaction effect between CO₂ and *Pythium* treatment in linear mixed effect model analysis on log-transformed data. The letter e denotes an exponent to the power of 10.

Table S1 List of primers and probes used in this study.

Table S1. List of primers and probes used in this study.		
Name	Oligonucleotide primer and Taqman probe (5'-3') sequences	Citation
BPMV	F: GAGAGTTGCTGCTTGAAGA	This study
	R: GTACCAATGAGGCCCAATCA	
	P: FAM/TGTGATGTC/ZEN/TGGGAATGCAGCAGT/IABkFQ ^a	
SMV	F: GAGAAGTTTACCCGATGGATGG	This study
	R: CATTGAGTAGTAAGGGCGTCAG	
	P: FAM/ACTGTCATG/ZEN/CAGATGGCTCACAGT/IABkFQ ^a	
<i>GmPR1</i>	F: GCTATGCTAATCAACGCAAAGG	This study
	R: ACTGCATCTGTGCCACTTAG	
	P: HEX/ACTGCCAAC/ZEN/TGATCCACTCTGGTG/IABkFQ ^b	
<i>GmKTI</i>	F: CCCATATCGAGTTCCTGTTATCC	This study
	R: GTCCTTCTGGTTGGCCATTA	
	P: TexRd-XN/CCCTTCGTTCTGAGTGGACCGT/IAbRQSp ^c	
<i>GmSkp1</i>	F: GAGCTCTCAGCACTAGGAATG	This study
	R: CAGGTCTGCCTCTAGAATCATC	
	P: Cy5/TCAGCAACAGTAACCACTGTAGCAATAGC/IAbRQSp ^d	
<i>Pythium sylvaticum</i>	F: GAAGGTGAAGTCGTAACAAGG	(Cooke et al., 2000)
	R: AGCGTTCTTCATCGATGTGC	
<i>Fusarium virguliforme</i>	F: GTAAGTGAGATTTAGTCTAGGGTAGGTGAC	(Wang et al., 2015)
	R: GGGACCACCTACCCTACACCTACT	
<i>GmDCL2</i>	F: GTTGCCCTCTACAGCCTATTT	This study
	R: CTTTAGTGACACGCTACCTCTG	
<i>GmAGO1</i>	F: CAAGATCCAGTCAGAGGAACAG	This study
	R: GCAGCAAGGTGAGCATAGTA	

^aFAM = 6-Carboxyfluorescein, IABkFQ = ZEN-Iowa Black FQ

^bHEX = Hexachlorofluorescein, IABkFQ = ZEN-Iowa Black FQ

^cTexRd-XN = Texas Red-X NHS ester, IAbRQSp = Iowa Black RQ-Sp

^dCy5 = Cyanine-5, IAbRQSp = Iowa Black RQ-Sp

Table S2 SDS rating scale of foliar symptoms ranging from 0 to 7.

Table S2. SDS rating scale of foliar symptoms ranging from 0 to 7 (Roth et al., 2019).	
Score	Foliar symptoms
0	No visible symptom
1	1 to 10% of leaf surface chlorotic and/or 1 to 5% necrotic
2	10 to 20% chlorotic and/or 6 to 10% necrotic
3	20 to 40% chlorotic and/or 11 to 20% necrotic
4	40 to 60% chlorotic and/or 21 to 40% necrotic
5	> 60% chlorotic and/or > 40% necrotic
6	premature leaf drop up to one-third defoliation
7	premature leaf drop up to two-thirds defoliation

Table S3 Disease rating scale of *P. sylvaticum* root symptoms.

Table S3. Disease rating scale of <i>P. sylvaticum</i> root symptoms (Zhang & Yang, 2000)	
Score	Root symptoms
0	roots were long and full with no signs of discoloration or infection by the pathogen
1	roots were slightly stunted and moderately discolored
2	roots were severely stunted and may be severely discolored
3	roots were very short and rotted
4	seed was rotted and had not germinated

Methods S1 QuantSeq 3'mRNA-Seq data analysis and jasmonic acid (JA) and salicylic acid (SA) quantification.

Bioinformatic and Statistical Analysis of 3' mRNA-Seq data

For each of the 3' mRNA-Seq datasets (48 and 96 libraries), libraries were each sequenced on two lanes, generating two sequence files per library. Files were inspected with FASTQC (de Sena Brandine & Smith, 2019) to confirm sequence quality and quantity. Trimmomatic was used to remove adaptor sequences, low quality bases and reads less than 50 bp in length (Bolger *et al.*, 2014). STAR version 2.7.10 (Dobin *et al.*, 2013) was used to align reads to the Williams 82 reference genome sequence (Wm82.a4, (Schmutz *et al.*, 2010)). Samtools version 1.17 (Li *et al.*, 2009) was used to merge read data from separate lanes and identify uniquely mapping reads within the combined mapping files (bam files). Mapping files were imported into RStudio version 4.2.3 (Dobin *et al.*, 2013) using Rsamtools (Bandyopadhyay & Thilmony, 2021). The gene feature file (GFF) corresponding to the reference genome was imported using rtracklayer (Lawrence *et al.*, 2009). The number of reads per sample aligning to each gene was counted using summarizeOverlaps and a count table for all expressed genes within a tissue was generated. Over 329 million and 659 million uniquely mapping reads were identified across QuantSeq datasets 1 and 2, respectively. Data were normalized using the Trimmed Mean of M values (Gupta *et al.*, 2021) in the Bioconductor package edgeR (Robinson & Smyth, 2007; Robinson *et al.*, 2010; McCarthy *et al.*, 2012; Zhou *et al.*, 2014). Only genes with log2 counts per million (cpm) > 1 in at least three replicates were used in the analysis. ggplot2 (Wickham, 2009) was used to generate principal components, biological coefficients of variance, and MA plots to visually compare sample replicates and ensure reproducibility (Dziuda, 2010). Two and eight libraries with less than 1 million uniquely mapping reads were removed from subsequent analyses from QuantSeq datasets 1 and 2, respectively, leaving a minimum of six plant replicates per condition. Following renormalization, edgeR was used to analyze differentially expressed genes (DEGs) among the different datasets. For QuantSeq dataset 1, we identified DEGs whose expression changed in response to CO₂ levels (eCO₂ levels/aCO₂) at 21 days while the false discovery rate (FDR) was controlled at 0.01. For QuantSeq dataset 2, we identified DEGs responding to *Psg* treatment

(*Psg*/mock) in control and $e\text{CO}_2$ conditions and DEGs whose expression changed in response to CO_2 in *Psg*-treated (*Psg* $e\text{CO}_2$ /*Psg* control CO_2) or mock-treated samples (Mock $e\text{CO}_2$ / $a\text{CO}_2$) while controlling FDR at 0.01. Hierarchical clustering using hclust (Murtagh, 1985) was used to group similarly expressed genes from DEG lists of interest. Heatmaps were generated using Heatmap.2 in ggplot2 (Wickham, 2009).

DEG Annotation and Analysis of Overrepresented Transcription Factors and Promoters

DEGs were annotated using the SoyBase Gene Annotation Lookup Tool for Wm82.a4 (<https://www.soybase.org/genomeannotation/>). For given DEGs of interest, the best BLASTP Arabidopsis hits (TAIR version 10) from the annotation were used to query the STRING database (Szklarczyk *et al.*, 2023). To identify transcription factors (TFs) within DEG lists of interest, we used custom perl scripts to query *G. max* transcription factor information downloaded from the Plant Transcriptional Regulatory Map/Plant Transcription Factor Database (PlantRegMap/PlantTFDB v5.0, (Tian *et al.*, 2020), <https://plantregmap.gao-lab.org>). To identify TF binding sites significantly overrepresented within promoters of DEGs of interest we used the PlantRegMap TF enrichment tool (https://plantregmap.gao-lab.org/tf_enrichment.php) for *G. max*, using all available methods and threshold p-value < 0.05.

Single Phase Extraction for Metabolite Analyses of JA and SA

Approximately 100 mg of each sample was ground and spiked with an internal standard containing 10 μg nonadecanoic acid (1 mg/mL in ethanol) and JA-d5 (Cayman Chemical, Ann Arbor, MI) (JA-d5) (2.5 μg /mL in ethyl acetate). The extraction was initiated with the addition of 90% ice cold methanol using a modified version of the methanolic extraction protocol established previously (A *et al.*, 2005).

GC-MS Acquisition and Analysis for SA

Six hundred microliters of the combined extracts were aliquoted and dried using a speed-vac concentrator for 10 hours and then derivatized according to established protocols (Koek *et al.*, 2006). Gas Chromatography Mass Spectrometry (GC-MS) analysis was performed with an Agilent

6890 gas chromatograph coupled to a model 5973 Mass Selective Detector (Agilent Technologies, Santa Clara, CA) using a HP-5MSI 5% phenyl methyl silox with 30 m × 250 µm × 0.25 µm film thickness (Agilent Technologies). Identification and quantification were conducted using AMDIS (Automated Mass spectral Deconvolution and Identification System, National Institute of Standards and Technology, Gaithersburg, MD) with a manually curated, retention indexed GC-MS library with additional identification performed using the NIST20 and Wiley 11 GC-MS spectral library (Agilent Technologies, Santa Clara, CA). Final quantification was calculated by integrating the corresponding peak areas relative to the area of the nonadecanoic acid internal standard. Raw data were normalized to the amount of tissue used.

LC-MS Acquisition for Targeted JA Analysis

An Oasis Hydrophilic-Lipophilic-Balanced (HLB) column (30 mg sorbent) (Waters Corporation, Milford, MA) was used for the purification of JA as established previously (Kojima *et al.*, 2009; Kojima & Sakakibara, 2012). Liquid chromatography (LC) separations were performed with an Agilent Technologies 1290 Infinity II UHPLC instrument equipped with an Agilent Technologies ZORBAX Eclipse C18 (1.8 µm 2.1 mm × 100 mm) analytical column that was coupled to a 6470 triple quadrupole mass spectrometer with an electrospray ionization source (Agilent Technologies, Santa Clara, CA). A volume of 20 µL of each sample was injected into the LC system with the mass spectrometer in negative mode. Multiple reaction monitoring (MRM) was used for detection with transitions of m/z 209→59 (JA) and 214→62 (JA-d5), with fragmentation conducted at 25 eV for both JA and JA-d5.

References

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