

Unravelling the metabolic reconfiguration of the post-challenge primed state in *Sorghum bicolor* responding to *Colletotrichum sublineolum* infection

Fidele Tugizimana¹, Paul A. Steenkamp¹, Lizelle A. Piater¹, Nico Labuschagne² and Ian A. Dubery^{1,*}

¹ Research Centre for Plant Metabolomics, Department of Biochemistry, University of Johannesburg, Auckland Park, South Africa

² Department of Plant and Soil Science, University of Pretoria, Pretoria, South Africa

* Correspondence: idubery@uj.ac.za; Tel.: + 27-011-559-2401.

Figure S1. Evaluation of disease symptoms in *Colletotrichum sublineolum* infected sorghum plants.

Figure S2. Representative MS chromatograms of ESI(+) data (3 d.p.i.).

Figure S3. Unsupervised chemometric modelling of ESI(-) data.

Figure S4. OPLS-DA modelling and variable/feature selection.

Table S1. Annotated (MSI-level 2) metabolites reported in Table 1, with fragmentation information (refs. [1-11]).

References [1] – [14].

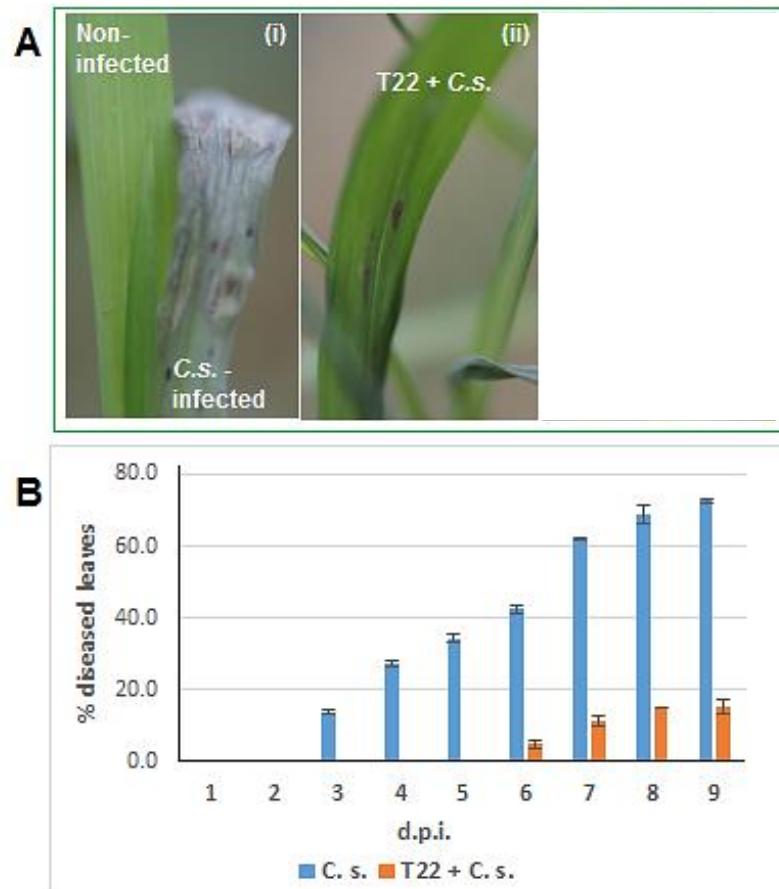


Figure S1. Evaluation of disease symptoms in *Colletotrichum sublineolum* infected sorghum plants. (A): Symptoms observed on sorghum leaves following the fungal challenge. (i) The sorghum plants that were not inoculated with a bacterial suspension show anthracnose symptoms that were severe at late stage of the disease development (9 d.p.i.). (ii) The sorghum plants inoculated with the *Paenibacillus alvei* (T22) bacterial strain, then challenged with *C. sublineolum*. The symptoms developed very late (from 6 d.p.i.) and were very few even at 9 d.p.i.. Chlorosis and wilting of the leaf could be observed. Purple spots indicate the accumulation of 3-deoxyanthocyanidins. (B): The rating of the symptom development over time (1 – 9 d.p.i.). The disease symptoms were scored by measuring the percentage of infected leaves relative to the total number of analysed leaves. The values are the means of the percentage of diseased leaves per plant ± SD.

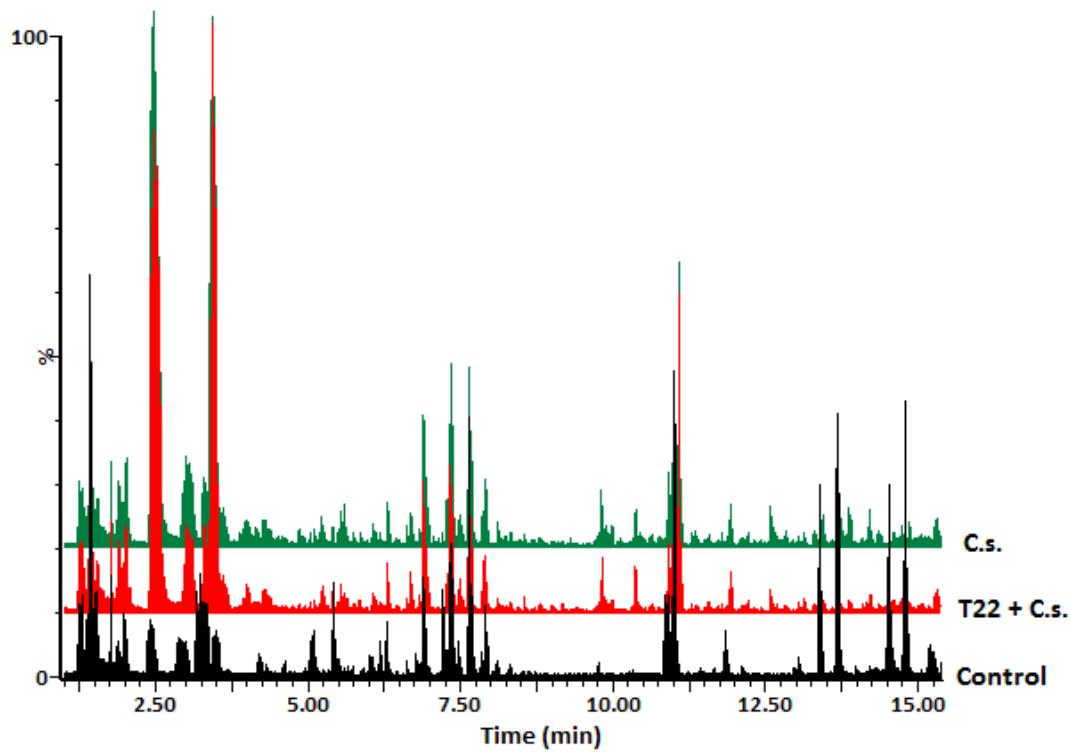


Figure S2. Representative MS chromatograms of ESI(+) data (3 d.p.i.). Base peak intensity (BPI) mass chromatograms displaying comparative chromatographic differences in different conditions: (i) samples from non-treated plants (Control, NT), (ii) samples from *Paenibacillus alvei* (T22)-primed and *C. sublineolum* (C.s.)-challenged plants and (iii) samples from C.s.-infected plants. Visual inspection of the chromatograms evidently shows differential peak populations, for instance in the 4–12 min chromatographic region.

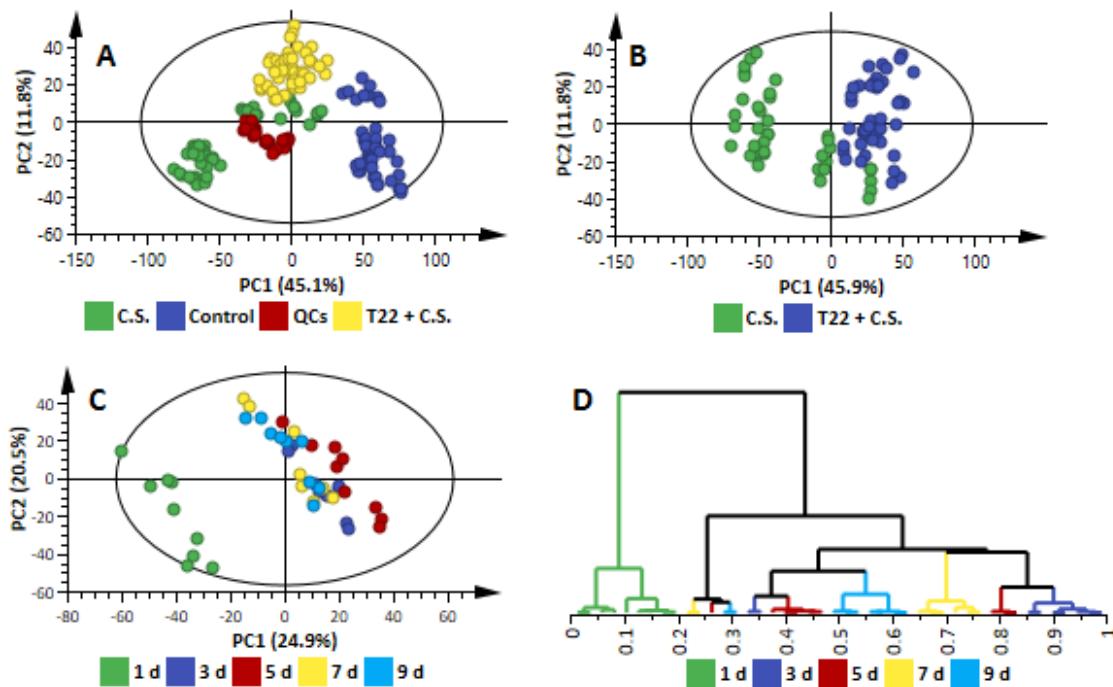


Figure S3. Unsupervised chemometric modelling of ESI(-) data:

(A): A PCA scores scatter plot (first two components) of a 12-component model, explaining 85.1% of the variation in Pareto-scaled data X (of all the samples, including the QC samples), and predicting 74.4% variation, according to cross-validation; and coloured according to the treatment: non-treated (control - blue), fungal treated plants (C.s. - green), *Paenibacillus alvei* (T22)-primed and challenged with *C. sublineolum* plants (T22 + C.s. - yellow). The scores plot shows treatment-related grouping, and the QC samples (brown) clustered together, indicating the reliability and good quality of the acquired data.

(B): Explorative analyses of the ESI(-) data of samples from fungal treated plants (C.s. - green) and T22-primed and challenged with *C. sublineolum* plants (blue). The PCA scores plot of an 8-component model ($R^2 = 0.803$ and $Q^2 = 0.677$) show treatment-related sample grouping, pointing to differential metabolic changes in the two treatment groups.

(C): A PCA scores plot of a 6-component model, explaining 70.8% of the variation in Pareto-scaled data X (of T22 + C.s. – primed and infected samples), with a predictive ability of 69.7%, according to cross-validation. The scores plot indicates time-related grouping, pointing to time-dependent metabolic changes in the response of T22-primed sorghum plants responding to the fungal infection.

(D): HCA dendrogram corresponding to (C), allowing the identification of natural clustering in multivariate metabolite space: treatment-related grouping.

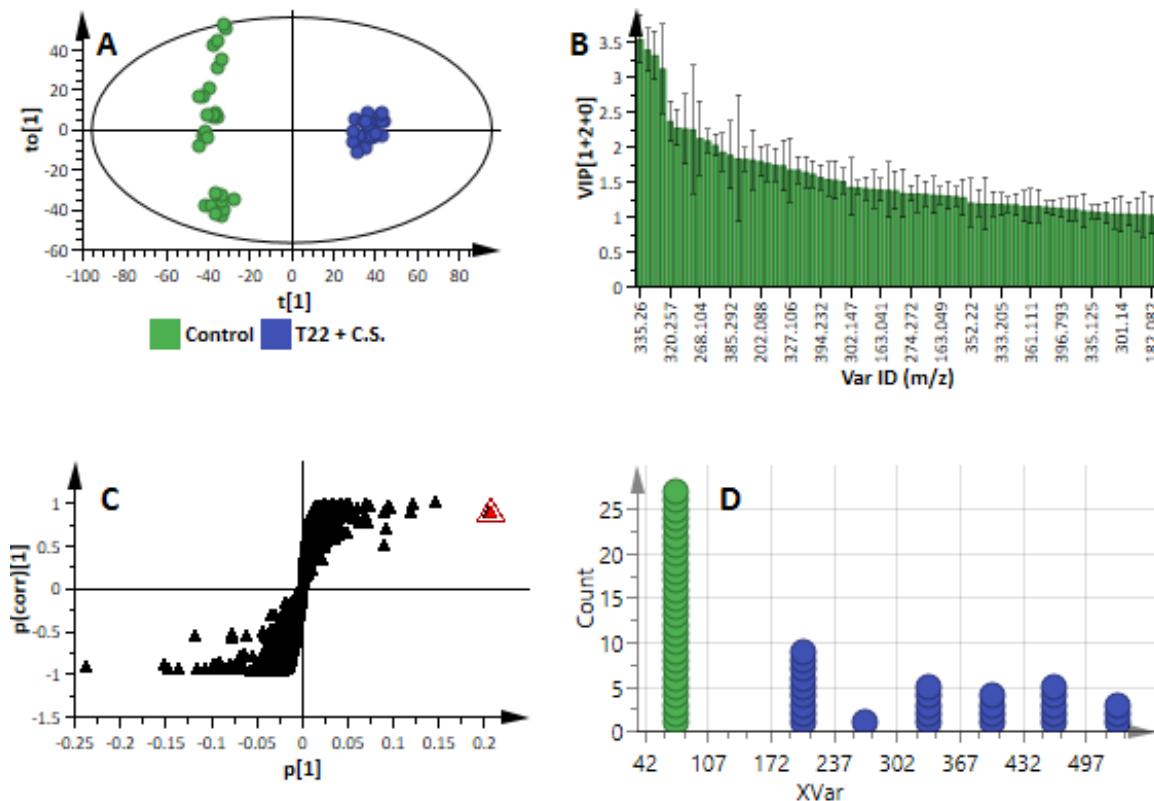


Figure S4. OPLS-DA modelling and variable/feature selection.

(A): A typical OPLS-DA scores scatter plot of the OPLS-DA model of ESI(-) data, separating 'control plants *vs.* challenged primed-plants' at 5-9 d.p.i. (1 + 2 + 0 components, $R^2X = 0.658$, $Q^2 = 0.978$, CV-ANOVA p -value = 0.00) (Figure 3). In the scores space, the two groups are clearly separated.

(B): A typical VIP plot (from the same OPLS-DA model as in Figure 3). Such a plot allows the identification / selection of variables (features) with high importance in driving the separation of the two groups in the binary model.

(C): A typical loadings S-plot used to select discriminating variables (features). The variables with high covariance and high correlation (those found in both extreme ends of the S-plot) are of interest – *e.g.* the red feature.

(D): A typical dot plot (of the red feature in S-plot) which allows the assessment of the discriminability of the selected variable. As displayed by the dot plot, the selected variable is a perfect discriminating feature, as no overlap can be seen in the separated groups.

Table S1. Annotated (MSI-level 2) metabolites reported in **Table 1**, with fragmentation information (refs. [1-14]).

Metabolites	Rt (min)	m/z	Fragmentation	Adduct	mode	MF	Class
1 L-Tyrosine	1.25	182.0819	136	H	pos	C ₉ H ₁₁ NO ₃	Amino acid
2 5-Hydroxytryptophan	2.65	236.1036	188	NH ₃	neg	C ₁₁ H ₁₂ N ₂ O ₃	Amino acid
3 L-Tryptophan	3.02	205.0978	188	H	pos	C ₁₁ H ₁₂ N ₂ O ₂	Amino acid
4 Dhurrin	4.02	329.1335	185, 307	NH ₃	pos	C ₁₄ H ₁₇ NO ₇	Cyanogenic glucoside
5 Naringin chalcone	2.52	627.1912	339, 315, 273	HCOOH	neg	C ₂₇ H ₃₄ O ₁₄	Flavonoid
6 Naringin	3.46	625.1761	271, 151	HCOOH	neg	C ₂₇ H ₃₂ O ₁₄	Flavonoid
7 Peptahydroxychalcone 4'-O-glucoside	4.73	449.1067	207, 287	H	neg	C ₂₁ H ₂₂ O ₁₁	Flavonoid
8 Hesperidin	5	609.1809	343, 301, 179, 151	H	neg	C ₂₈ H ₃₄ O ₁₅	Flavonoid
9 Apigenin 7-O-[beta-D-apiosyl-(1->2)-beta-D-glucoside]	5.05	563.139	269, 225, 197	H	neg	C ₂₆ H ₂₈ O ₁₄	Flavonoid
10 Kaempferol 3-O-rhamnoside-7-O-glucoside	5.68	593.1501	447, 285	H	neg	C ₂₇ H ₃₀ O ₁₅	Flavonoid
11 Cyanidin 3-O-rhamnosylglucoside	5.75	595.1657	449, 300, 287	H	pos	C ₂₇ H ₃₀ O ₁₅	Flavonoid
12 Kaempferol-3-glucoside	5.88	447.0921	285	H	neg	C ₂₁ H ₂₀ O ₁₁	Flavonoid
13 Quecertin	5.92	301.0367	273, 193, 179, 151	H	neg	C ₁₅ H ₁₀ O ₇	Flavonoid
14 Apigenin	6.02	271.1544	243, 227, 203, 153, 109	H	pos	C ₁₅ H ₁₀ O ₅	Flavonoid
15 Apigeninidin	6.1	255.1533	227, 181, 171, 157, 115	H	pos	C ₁₅ H ₁₁ O ₄	Flavonoid
16 Luteolin 7-O-beta-D-glucoside	6.19	447.0921	357, 287, 153, 135, 117	H	neg	C ₂₁ H ₂₀ O ₁₁	Flavonoid
17 Apigenin 7-O-neohesperidoside	6.27	579.1709	227, 153	H	pos	C ₂₇ H ₃₀ O ₁₄	Flavonoid
18 Luteolin	6.3	287.0536	259, 243, 201, 177	H	pos	C ₁₅ H ₁₀ O ₆	Flavonoid
19 1,2-Bis-O-sinapoyl-beta-D-glucoside	6.35	591.1705	367, 206	H	neg	C ₂₈ H ₃₂ O ₁₄	Flavonoid
20 7-O-Methylvitexin 2"-O-beta-L-rhamnoside	6.39	615.168	225, 187, 115	Na	pos	C ₂₈ H ₃₂ O ₁₄	Flavonoid
21 Isovitexin 2"-O-beta-D-glucoside	6.68	593.1501	473, 447	H	neg	C ₂₇ H ₃₀ O ₁₅	Flavonoid
22 Luteolinidin	6.87	271.0616	225, 197, 187, 115	H	pos	C ₁₅ H ₁₁ O ₅	Flavonoid
23 12,13-Epoxy-9-hydroxy-10-octadecenoate	9.26	395.204	197, 171	HCOONa	neg	C ₁₈ H ₃₂ O ₅	Lipid
24 Phytosphingosine	10.52	318.3009	---	H	pos	C ₁₈ H ₃₉ NO ₃	Lipid

25	16-Hydroxypalmitate	10.58	290.27	229, 173, 159, 145	NH3	pos	C ₁₆ H ₃₂ O ₃	Lipid
26	(9Z)-(13S)-12,13-Epoxyoctadeca-9,11-dienoic acid	11.44	363.2137	---	HCOONa	pos	C ₁₈ H ₃₀ O ₃	Lipid
27	13(S)-hydroxyperoxyoctadecatrienoic acid	11.79	309.2071	314, 135	H	neg	C ₁₈ H ₃₀ O ₄	Lipid
28	25-Hydroxy-24-epi-brassinolide	13.34	519.3267	---	Na	pos	C ₂₈ H ₄₈ O ₇	Lipid
29	Oleanolate 3-beta-D-glucuronoside-28-glucoside	15.36	795.4497	---	H	pos	C ₄₂ H ₆₆ O ₁₄	Lipid
30	Oleanoic acid 3-O-glucuronide	15.4	655.382	---	Na	pos	C ₃₆ H ₅₆ O ₉	Lipid
31	Caffeoylquinate	3.83	377.0846	193, 181, 175, 121	Na	pos	C ₁₆ H ₁₈ O ₉	Phenylpropanoid
32	p-Coumaroyl quinic acid	1.03	427.0621	337, 191, 173	NaHCOONa	neg	C ₁₆ H ₁₈ O ₈	Phenylpropanoid
33	Feruloyltyramine	2.01	331.165	314, 192, 180, 137	NH ₃	pos	C ₁₈ H ₁₉ NO ₄	Phenylpropanoid
34	4-Coumaroylshikimate	3.16	319.1062	301, 283, 163, 119	H	neg	C ₁₆ H ₁₆ O ₇	Phenylpropanoid
35	2-Coumarate	3.25	165.0554	119	H	pos	C ₉ H ₈ O ₃	Phenylpropanoid
36	1-O-Sinapoyl-beta-D-glucose	3.56	387.1279	225, 207, 181	H	pos	C ₁₇ H ₂₂ O ₁₀	Phenylpropanoid
37	4-O-beta-D-Glucosyl-4-hydroxycinnamate	4.09	395.0947	217, 193, 175	HCOONa	pos	C ₁₅ H ₁₈ O ₈	Phenylpropanoid
38	Ferulate	4.58	209.0448	194, 178, 161, 134	H	neg	C ₁₀ H ₁₀ O ₅	Phenylpropanoid
39	O-Feruloylquinate	4.88	367.1017	193, 173, 134	H	neg	C ₁₇ H ₂₀ O ₉	Phenylpropanoid
40	Coniferyl acetate	1.09	291.0844	182	HCOONa	pos	C ₁₂ H ₁₄ O ₄	Phenylpropanoid
41	Zeatin	2.38	220.1192	202, 136, 119	H	pos	C ₁₀ H ₁₃ N ₅ O	Phytohormone
42	Salicylate-glucoside	1.79	299.0758	136, 92	H	neg	C ₁₃ H ₁₆ O ₈	Phytohormone
43	6-Hydroxy-indole-3-acetyl-phenylalanine	2.76	405.1077	120, 103, 93	HCOONa	neg	C ₁₉ H ₁₈ N ₂ O ₄	Phytohormone
44	6-Hydroxy-indole-3-acetyl-valine	2.82	335.0962	173, 129	Na_Na	pos	C ₁₅ H ₁₈ N ₂ O ₄	Phytohormone
45	(-)Jasmonoyl-L-isoleucine	4.33	406.1626	338, 336, 301	HCOOK	neg	C ₁₈ H ₂₉ NO ₄	Phytohormone
46	12-Hydroxyjasmonic acid 12-O-beta-D-glucoside	5.59	429.1514	370, 267	Na_Na	neg	C ₁₉ H ₃₀ O ₈	Phytohormone
47	<i>trans</i> -Zeatin-7-beta-D-glucoside	8.14	399.199	202, 181, 136, 119	NH ₃	pos	C ₁₆ H ₂₃ N ₅ O ₆	Phytohormone
48	Riboflavin	5.8	419.0969	398, 381, 355, 224, 143	Na_Na	neg	C ₁₇ H ₂₀ N ₄ O ₆	Riboflavin
49	Feruloylserotonin	11.66	351.1333	177, 149, 145	H	neg	C ₂₀ H ₂₀ N ₂ O ₄	Trp pathway

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