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5 **Metabolomic Study of Stress Responses Leading to Plant Resistance**
6 **in Mandarin Fruit Mediated by Preventive Applications of *Bacillus***
7 ***subtilis* Cyclic Lipopeptides**
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26 **ABSTRACT**

27 Although green mold rot caused by *Penicillium digitatum* is a major postharvest disease in
28 mandarin fruit, the fruit's defense mechanism at the metabolomic level is largely unknown.
29 Here, the expressed metabolome network leading to plant resistance to stresses induced by
30 exposing of different agents was analyzed. Inoculation of mandarin fruits with eight
31 individual agents, including four *Bacillus* cyclic lipopeptides (CLPs) produced by *B. subtilis*
32 ABS-S14, three phytohormones and *P. digitatum*, resulted in different wound appearance on
33 flavedo (mandarin peel) tissues. Subsequent metabolomic analysis with dansylation isotope
34 labeling LC-MS method detected and quantified 4,717 metabolites, including 77 metabolites
35 positively identified belonging to 39 metabolic pathways. The preventive applications of
36 CLPs showed the greatest effect with many up-regulated metabolite changes in fruit tissues,
37 including two important secondary metabolites, serotonin and tyramine, which were reported
38 to stimulate plant defensive system during stress. Further analysis indicated that CLPs
39 triggered the metabolism of glycine, serine and threonine, a major pathway to induce
40 serotonin production, and activated tyrosine metabolism resulting in an increase of tyramine
41 production. These findings provide the new insights for fruit protection manipulation from
42 green mold pathogen invasion during postharvest storage.

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44 **Keywords:** *Bacillus subtilis*; Citrus; Cyclic lipopeptides; Metabolomics; *Penicillium*
45 *digitatum*; Plant hormone.

46

47 **1. Introduction**

48 Green mold rot caused by *Penicillium digitatum* is a major factor in economic loss of
49 citrus fruit (Eckert, 1978). The infection of *P. digitatum* generally occurs on the pores or
50 wound sites on mandarin peel during postharvest period (Palou et al., 2008). Synthetic
51 fungicides have been used in the prevention and treatment of *P. digitatum* infection, but
52 consumers are more concerning about their negative effects on environment and human
53 health (Talibi et al., 2014). Identification and use of safer biological control agents to reduce
54 rot during the postharvest period therefore has the potential to be a feasible strategy to
55 overcome the limitations inherent in the usage of synthetic fungicides (Palou et al., 2008;
56 Talibi et al., 2014).

57 Microbial compounds such as cyclic lipopeptides (CLPs) isolated from *Bacillus* species
58 have been exploited as biological control agents to restrict plant pathogen invasion (Ongena
59 and Jacques, 2008; Rosier et al., 2018). *B. subtilis* can not only elicit plant defense responses,
60 but also act as an alternative agent to replace pesticide usage for the control of green mold
61 disease in citrus fruit (Leelasuphakul et al., 2008; Waewthongrak et al., 2015). For instance,
62 an antagonistic strain *B. subtilis* ABS-S14 was recently reported to produce CLPs consisting
63 of three main families (i.e., fengycins, iturins and surfactins), which manifested strong
64 antifungal activities against *P. digitatum* and induced resistance in citrus fruit
65 (Waewthongrak et al., 2014). Moreover, CLPs of *Bacillus subtilis* were shown to be involved
66 in complex regulatory processes of host plant immunity such as induced systemic resistance
67 system (Ongena and Jacques, 2008).

68 In addition to CLPs, plant phytohormones have also been studied for their potential as
69 biological control agents to inhibit plant pathogen infection (Guo et al., 2014; Moscoso-
70 Ramírez and Palou, 2013). Potent exogenous plant hormones such as salicylic acid (SA),
71 methyl jasmonate (MeJA) and ethephon (Et) involve in SA, jasmonic acid and ethylene

72 signaling transduction pathways during plant defense responses, respectively (Guo et al.,
73 2014; John-Karuppiah and Burns, 2010; Moscoso-Ramírez and Palou, 2013; Vlot et al.,
74 2009; Zhou et al., 2018). Preventive applications of SA and MeJA could inhibit the green
75 mold disease in citrus fruit (Guo et al., 2014; Moscoso-Ramírez and Palou, 2013). It could
76 induce the expression of *1-amino-cyclopropane-1-carboxylate synthase-1* and *1-amino-*
77 *cyclopropane-1-carboxylate oxidase* genes in the ethylene signaling pathway, which is
78 related to the induced systemic resistance system in sweet orange (John-Karuppiah and Burns,
79 2010).

80 As potential biological control agents, *Bacillus* CLPs and plant hormones offer different
81 extent of protection to the citrus fruits against green mold pathogen (Guo et al., 2014;
82 Moscoso-Ramírez and Palou, 2013; Waewthongrak et al., 2015). During postharvest period,
83 metabolic changes induced by *Bacillus* CLPs and phytohormones in mandarin fruit occurred
84 in response to stresses such as wound stress, which could induce plant immunity through the
85 induced systemic resistance system (Tunsagool et al., 2019). Since *P. digitatum* infects citrus
86 fruit from the wound sites (Eckert, 1978), this mold would also induce plant immunity in a
87 similar way (Ballester et al., 2013). If preventive application of CLPs or exogenous plant
88 hormones can effectively stimulate wound recovery of citrus fruit's peel and plant immunity
89 against pathogen defense, this approach would effectively prevent green mold infection.
90 Therefore, in order to develop an effective method of using these biological control agents,
91 especially the bioactive *Bacillus* CLP compounds, to prevent green mold infection, it is
92 critical to investigate if these agents can stimulate wound healing of citrus fruit's peel and to
93 explore how these agents regulate metabolites against wound stress and related pathogen
94 defense responses in citrus fruit.

95 Metabolomics is a key technology to profile the metabolites in metabolic pathways (Guo
96 and Li, 2009; Huan et al., 2015; Shen et al., 2016). Metabolomics is also a powerful approach

97 to expand the knowledge of plant defense mechanism through signaling pathways by
98 determination of the metabolites which occurred in specific metabolic networks and stress
99 responses of plant (Hall et al., 2002). Metabolomics has been used to investigate primary and
100 secondary metabolites to monitor and assess gene function (Asai et al., 2017; Hall et al.,
101 2002; Sampaio et al., 2016). It has also been used to characterize post-genomic processes
102 from a broad perspective (Tugizimana et al., 2013), and to find biomarker candidates during
103 defense responses in citrus leaves (Asai et al., 2017). Based on gas chromatography mass
104 spectrometry (GC-MS) and statistical analysis, Asai et al. (2017) reported the involvement of
105 tryptophan and serine regulation in stress responses in citrus leaves when being wounded and
106 exposed to MeJA and SA. With the recent development of high-performance chemical
107 isotope labeling (CIL) liquid chromatography mass spectrometry (LC-MS), it is now possible
108 to quantify the metabolomic changes of biological systems with high metabolic coverage and
109 high quantification accuracy, thereby allowing the possibility of revealing up- and down-
110 regulated metabolites in many metabolic pathways.

111 In this study, we examined the wound recovery outcome of mandarin fruit responding to
112 various agents applied on the wound surface of flavedo, including *B. subtilis* CLPs and
113 exogenous phytohormones, and compared with green mold pathogen infection. In order to
114 better understand the CLPs' action on induction of host plant resistance, the metabolic
115 changes induced by different agents applied to citrus fruit were analyzed using metabolomics.
116 Dansylation LC-MS was applied to profile the metabolomic differences of the wound tissues
117 to examine metabolic responses, in particular, the defense response pathways.

118 **2. Materials and methods**

119 *2.1. Chemicals and reagents*

120 SA, MeJA and Et used as exogenous plant hormones were purchased from Sigma-Aldrich
121 and solutions were prepared in 80% ethanol. For dansylation labeling, ^{12}C -dansyl chloride
122 was purchased from Sigma-Aldrich and ^{13}C -dansyl chloride was synthesized according to a
123 previous study (Guo and Li, 2009). LC-MS grade water, methanol and acetonitrile were
124 purchased from Thermo Fisher Scientific (Alberta, Canada). All other chemicals and reagents
125 were purchased from Sigma-Aldrich Canada (Ontario, Canada).

126 2.2. Plant materials

127 Mandarin fruit (*Citrus reticulata* Blanco cv. Shogun) of uniform size and color without
128 physical injuries or infections were selected from the commercial orchards located in Loei
129 province of northeastern Thailand. The citrus plants were grown under organic control with
130 non-postharvest treatment applied. The fruit were surface-sterilized with 1% sodium
131 hypochlorite solution for 5 min and then allowed to air dry at 25°C (Waewthongrak et al.,
132 2015).

133 2.3. Microorganisms

134 *P. digitatum* was isolated from decayed mandarin fruit (Leelasuphakul et al., 2008). It was
135 identified microbiologically and collected in the Fungal Biodiversity Laboratory, National
136 Center for Genetic Engineering and Biotechnology, National Science and Technology
137 Development Agency, Thailand. A pure culture was placed on a potato dextrose agar (PDA)
138 plate and incubated at 24°C for 7 days. To maintain their pathogenicity, the fungal spores
139 were routinely inoculated back into mandarin fruit. The *P. digitatum* spore suspension was
140 prepared from a 7-day-old culture and a haemocytometer was used to adjust it to 10^7 spores
141 L^{-1} (Waewthongrak et al., 2014).

142 *B. subtilis* ABS-S14 was isolated from soil collected from citrus groves around the south
143 of Thailand. It was identified by Gram staining, cell shape and the presence of spores, and

144 bio-chemical analysis. The screening test of its antagonistic properties against a fungal
145 colony of *P. digitatum in vitro* was performed prior to use (Leelasuphakul et al., 2008).

146 2.4. Preparation and purification of CLPs

147 The *B. subtilis* ABS-S14 strain was grown in LB medium (Waewthongrak et al., 2014),
148 following the method reported (McKeen et al., 1986). A crude extract was prepared from
149 culture-free filtrates obtained after drying in a rotary vacuum evaporator at 65°C. It was
150 weighed and re-dissolved in 80% ethanol, then crude CLP extract was adjusted to a 50 g L⁻¹
151 stock solution (Leelasuphakul et al., 2006). Fengycin, iturin A and surfactin, belonging to the
152 CLP members, were prepared from *B. subtilis* ABS-S14 crude CLP extract using preparative
153 thin-layer chromatography (PTLC) following the reported method (Waewthongrak et al.,
154 2014). They were recovered by ethanol extraction from silica gel on a PTLC plate and further
155 separated using a C18 solid phase extraction (SPE) cartridge (Sep-Pak®, Vac 12 cc (2 g),
156 silica, 15-105 µm, 125 Å pore size, Waters, USA) to increase the purity. A sorbent matrix
157 was eluted by step gradients of acetonitrile in 0.1% trifluoroacetic acid: 40% to 55%, 25% to
158 35% and 60% to 80% for fengycin, iturin A, and surfactin, respectively. The fractions were
159 dried by rotary evaporator and re-dissolved in 80% ethanol.

160 2.5. Identification of CLPs

161 Identification of the CLPs was performed by HPLC (Agilent 1200, Agilent Technologies
162 Inc., USA) and matrix-assisted laser desorption ionization time-of-flight (MALDI-TOF) MS
163 (Bruker Daltonic Ultraflex III TOF/TOF, Bruker Daltonics Ltd., Germany) following the
164 previous study (Waewthongrak et al., 2014). The matrix used was α -cyano-4-
165 hydroxycinnamic acid. The bacterial CLPs were detected in the range of molecular ion peaks
166 from m/z 600 to 1,800.

167 2.6. Inoculation of mandarin fruit

168 The mandarin oranges were divided into ten treatment groups (five fruit per group with
169 three replicates per group). Artificial wounds were made at two sites on opposite sides of the
170 fruit. Five wounds of a circle of 0.5 cm in diameter and 3 mm in depth were made by
171 puncturing the fruit rind on the equator of the fruit with a sterile needle (Waewthongrak et al.,
172 2014). An aliquot of 20 μL of crude CLP extract (10 g L^{-1}) (Waewthongrak et al., 2015),
173 fengycin (1 g L^{-1}) (Waewthongrak et al., 2014), iturin A (1 g L^{-1}) (Waewthongrak et al.,
174 2014), surfactin (1 g L^{-1}) (Waewthongrak et al., 2014), SA ($3.4 \times 10^{-2} \text{ g L}^{-1}$) (Moscoso-
175 Ramírez and Palou, 2013), MeJA ($22.2 \mu\text{L L}^{-1}$) (Guo et al., 2014), Et ($450 \mu\text{L L}^{-1}$) (John-
176 Karupiah and Burns, 2010) in 80% ethanol and *P. digitatum* conidial suspension (10^7 spores
177 L^{-1}) in sterile distilled water was dropped into both wound sites on each fruit, respectively.
178 Consistently, sterile distilled wahter was used as the negative control of *P. digitatum* and
179 ethanol was used as the negative control of exogenous plant hormones (SA, MeJA, and Et)
180 and the extracted products of *B. subtilis* ABS-S14 (CLP extract, fengycin, iturin A, and
181 surfactin). Each treatment was placed in a plastic box containing a cup of water to maintain a
182 high humidity at 25°C and incubated for 24, 48 and 72 h. The appearance of wounds on the
183 treated flavedo tissues (outer colored part of mandarin peel) was observed and photographed
184 daily. The disease incidence was measured from the visible lesion on flavedo tissue by
185 summing the number of fruit with an average lesion $\times 100$ divided by the total number of
186 inoculated wounded fruit (Leelasuphakul et al., 2008; Waewthongrak et al., 2015).

187 2.7. Extraction of metabolites

188 The treated flavedo tissue was collected approximately 1 cm away from the wound site.
189 The flavedo tissues in each replicate of the same treatment were pooled and grounded to a
190 fine powder in liquid nitrogen. The metabolites from the fine powder of 15 mg were extracted,

191 based on a method previously reported (Asai et al., 2017), by dissolution in 1 mL of a
192 mixture of methanol/water/chloroform (2.5:1:1 v/v/v), incubation at 40°C for 5 min, and then
193 centrifugation at 14,000 g for 10 min. The supernatant (950 µL) was transferred to a new tube
194 and 400 µL of water was added. The mixture was centrifuged at 14,000 g for 10 min to
195 obtain a polar phase solution. Each sample was transferred to a new tube and stored at -80°C.
196 The workflow of this study for metabolomic analysis is shown in Supplementary Fig. 1.

197 2.8. Dansylation LC-MS metabolomic profiling

198 To prepare the samples for dansyl labeling, an individual sample was prepared from 25 µL
199 of the polar phase solution in each replicate. A pooled sample was obtained by mixing small
200 aliquots of all the samples. Methanol was evaporated by a SpeedVac concentrator and 25 µL
201 of water was added just prior to labeling. An individual sample was labeled with ¹²C-dansyl
202 chloride and the pooled sample was labeled with ¹³C-dansyl chloride following a reported
203 method (Han et al., 2017). The labeled sample was injected into LC-UV for measuring the
204 total concentration of labeled metabolites for sample amount normalization. For LC-UV, a
205 Waters ACQUITY UPLC system with a photodiode array (PDA) detector operating with a
206 Phenomenex Kinetex C18 column (50 mm × 2.1 mm, 1.7 µm particle size, 100 Å pore size)
207 was used. The total peak areas were detected under 338 nm for the quantification of each
208 sample (Hooton et al., 2016). An equal mole amount of ¹²C-labeled sample and ¹³C-labeled
209 pool was mixed. The mixture was then injected into an HPLC system interfaced to an
210 electrospray ionization (ESI) source in a Bruker 9.4 T Apex-Qe Fourier-transform ion
211 cyclotron resonance (FT-ICR) mass spectrometer (Bruker, Billerica, MA). The ESI mass
212 spectra were acquired in the positive ion mode. The setup of FTICR-MS was following a
213 previous report: nitrogen nebulizer gas, 2.3 L min⁻¹; dry gas flow, 7.0 L min⁻¹; dry
214 temperature, 195 °C; capillary voltage, 4200 V; spray shield, 3700 V; acquisition size, 256 k;

215 mass scan range, m/z 200 - 1000; ion accumulation time, 1 s; TOF (AQS), 0.007 s; DC
216 extract bias, 0.7 V (Peng et al., 2014). The HPLC flow rate used was $180 \mu\text{L min}^{-1}$ and the
217 sample injection volume was 13 μL . The samples were randomly injected in no particular
218 order. Quality control (QC) sample (^{12}C -/ ^{13}C -labeled pool) was injected every ten sample-
219 runs.

220 2.9. Data processing

221 To analyze the ^{12}C -/ ^{13}C -peak pairs from each LC-MS run, the IsoMS, IsoMS-Align, Zero-
222 fill and IsoMS-Quant programs were used. Metabolites was identified using the DnsID
223 standards library for positive metabolite identification (Huan et al., 2015). A search for
224 putative metabolite identification was performed by MyCompoundID MS software (Huan et
225 al., 2015), against the Human Metabolome Database (HMDB) library and the Evidence-based
226 Metabolome Library (EML) database following the methods described in previous reports
227 (Han et al., 2017; Huan et al., 2015; Shen et al., 2016).

228 2.10. Statistical analysis

229 MetaboAnalyst 4.0 was used for principal components analysis (PCA) and pathway
230 analysis of the LC-MS data (Chong et al., 2018). Microsoft Excel was used to calculate the
231 fold change and p-value between groups. Venn diagram was built using in-house R program.
232 Volcano plots (binary comparison) were performed using OriginPro 8.0 (OriginLab). The q-
233 value, multiple-testing-corrected p-value, was calculated using R and BioConductor
234 (www.bioconductor.org). A statistical test of variance of differences (ANOVA) was
235 performed for the data sets of disease incidence and the significant differences in mean values
236 were determined with the Tukey's range test ($p \leq 0.05$).

237 3. Results

238 CLPs of *B. subtilis* ABS-S14 are composed of three main compounds, fengycin, iturin A
239 and surfactin (Waewthongrak et al., 2014). To test their effects on wound appearance
240 development, these compounds were firstly separated with C18-SPE and verified with HPLC
241 and MALDI-TOF MS. Fengycin, iturin A and surfactin were recovered from the bacterial
242 CLP extract after eluted from the C18 SPE cartridge at the intervals of 19-32, 5-18 and 41-57
243 min of retention time (Supplementary Fig. 2). The MALDI-TOF MS spectra of the extracts at
244 three intervals of m/z values in the ranges of 1,449.74-1,547.82, 1,058.61-1,110.60 and
245 1,044.61-1,074.62 were found to match with those of the commercial standards (Sigma-
246 Aldrich, USA) (Supplementary Fig. 3).

247 3.1. Effect of plant tissue treatments on wound appearances

248 The progression of various treatments on wound appearance development on flavedo
249 tissues showed distinguished patterns of damage change (Fig. 1). In the *P. digitatum* infected
250 flavedo treatment, occurrence of watery spots was only observed on the wound sites of the
251 pathogen infected flavedo tissues when compared with the negative control (sterile distilled
252 water) at 72 h post-treatment. Treatments of exogenous plant hormones showed a similar
253 pattern of wound appearance of brown-damaged spots on flavedo tissues when compared
254 with the negative control (ethanol) from the first day to the end of treatment. On the contrary,
255 the treatments with the extracted *B. subtilis* ABS-S14 products (CLP extracts, fengycin, iturin
256 A and surfactin) showed much less brown-damaged spot and revealed statistically significant
257 reduction of disease incidence on flavedo tissues than plant hormone and ethanol treatments
258 (Supplementary Fig. 4).

259 3.2. Submetabolome profiling and metabolite identification

260 Metabolites in the treated flavedo tissues of the mandarin fruit were profiled using ^{12}C -
261 ^{13}C -dansylation LC-MS analysis. A total of $4,914 \pm 111$ peak pairs or metabolites were

262 detected in the duplicate analysis of 90 flavedo samples. Note that IsoMS filtered out all
263 redundant peak pairs of the same metabolite such as adducts or dimers and then retained only
264 one peak pair for one labeled metabolite. Thus, the number of peak pairs reflects the number
265 of metabolites detected. Among these detected metabolites, 77 metabolites were positively
266 identified using a dansyl standard library search based on accurate mass and retention time
267 matches (Supplementary Table 1) (Huan et al., 2015). In addition, 666 and 2,769 peak pairs
268 were mass-matched to metabolites in the HMDB library and the EML library, respectively
269 (Supplementary Tables 2 and 3) (Huan et al., 2015). These results indicated that dansylation
270 LC-MS could be used to detect a large number of amine- and phenol-containing metabolites.
271 More importantly, the metabolome composition of the flavedo samples appeared to be very
272 complex and the detected 77 positively identified metabolites plus many more putatively
273 matched metabolite covers a wide range of metabolic pathways, which will be discussed in
274 the section of pathway analysis.

275 *3.3. Plant metabolome profiles and comparison*

276 From the data set of 180 LC-MS runs of 90 samples, we used statistical tools to analyze
277 the metabolome differences among different groups of samples. PCA showed distinct
278 separations from tested samples using different treatments of various substances and from
279 samples of different time points as well as the 19 QC runs (Fig. 2A). The metabolome data
280 could be separated into four major groups plus the QC group. Very tight clustering of the QC
281 runs indicated the high reproducibility of the analysis, which was not surprising because the
282 ^{13}C -labeled pool was used as the internal control of all the ^{12}C -labeled samples in our analysis.
283 Using a heavy-isotope labeled internal control greatly reduced the ion suppression and matrix
284 effects as well as compensated for instrument sensitivity drift, if any, during the sample runs.

285 In Fig. 2A, samples from different agents applied at different time points are labeled with
286 different colors. Samples from the same agent at the same time point such as the samples in
287 iturin A treatment at 24 h post-treatment are categorized into the same group. Fig. 2B shows
288 the four-color PCA plot of the samples with various treatments at all time points. The
289 samples with sterile distilled water and ethanol were clustered together (green) and defined as
290 the control group for wound stress. The sample groups colored in light blue, blue, and red
291 represented the treatment groups with *P. digitatum*, exogenous plant hormones, and extracts
292 of *B. subtilis* ABS-S14 (i.e., CLP extracts containing fengycin, iturin A, and surfactin and the
293 separated fengycin, iturin A and surfactin compounds), respectively.

294 The labeled samples in the groups of control (green color) and exogenous plant hormones
295 (blue color) at all time points had sub-groups in their treatments as shown in the red circle in
296 Fig. 2B. The sub-groups in the red circle of Fig. 2B were identified in different color-labeled
297 samples as shown in red circle of Fig. 2C. Based on the time point analysis of each treatment
298 (Fig. 2C), clustering of the samples was detected in tissues treated with *P. digitatum* (black
299 color) and CLPs (red color) at every time point. The similar pattern of clustering together at
300 24 and 48 h post-treatment from the treatment group of exogenous plant hormones and the
301 control (yellow and green colored groups, respectively) was showed in the PCA plot of the
302 metabolome data (Fig. 2C); however, the samples detected at 72 h post-treatment clustered
303 differently. Fig. 2C shows that the treatment of ethephon at 24 h post-treatment deviated from
304 the samples at 48 and 72 h post-treatment.

305 Supplementary Fig. 5 shows the Venn diagram of the peak pair or metabolite numbers
306 detected from the four major treatment groups. There were 3,957 metabolites commonly
307 detected in the four major groups. Only a very small fraction of the metabolites were
308 uniquely detected in each group (i.e., 57, 24, 72, and 66 unique metabolites were found in the
309 treatments of CLPs, *P. digitatum*, exogenous plant hormone, and control, respectively). To

310 determine the metabolic changes, we further analyzed the data using binary comparisons of
311 different groups. The resulting volcano plots indicated the metabolic changes of the fruit peel
312 samples after being exposed to fungal pathogen, exogenous plant hormones, and *B. subtilis*
313 ABS-S14 CLPs, respectively, as compared to their corresponding control group (Fig. 3). In
314 the preparation of the *P. digitatum* spore suspension, sterile distilled water was used, whereas
315 ethanol was used for the solutions of exogenous plant hormone and CLPs. Therefore, the
316 effect of water or ethanol was discounted using binary comparison analysis of the
317 submetabolome in the treated group vs. its corresponding control (i.e., *P. digitatum* treatment
318 vs. water, exogenous plant hormones vs. ethanol, and CLPs vs. ethanol). In addition, any
319 change in metabolite concentration of more than 1.5-fold and a q-value of below 0.05 with
320 associated p-value was considered as being significant.

321 The volcano plot analysis of *P. digitatum* treatment vs. control (Fig. 3A) indicates that out
322 of 4,577 metabolites, 1,061 metabolites were down-regulated (i.e., lower metabolite
323 concentration in the treated samples than that of the control) and 407 metabolites were up-
324 regulated (i.e., higher metabolite concentration in the treated samples). In the group of
325 exogenous plant hormones vs. control (Fig. 3B), out of 4,545 metabolites, only three
326 metabolites were found to be down-regulated and 17 metabolites were up-regulated.
327 Interestingly, in the group of CLPs vs. control (Fig. 3C), out of 4,594 metabolites, 748
328 metabolites were down-regulated and 600 metabolites were up-regulated. Regarding the
329 significantly changed metabolites in the three binary comparisons, the group of CLPs showed
330 the greatest number of up-regulated metabolites. This group also showed the greatest
331 numbers of positively identified metabolites among the tested groups, based on a search in
332 the dansyl standard library (Huan et al., 2015). Noticeably, the treated group of exogenous
333 plant hormones had the minor changes in metabolite regulation whereas the group of fungal
334 infection showed the greatest numbers of down-regulated metabolites.

335 3.4. Metabolite regulation in plant defense to stress responses

336 *B. subtilis* CLPs elicited the largest number of metabolite changes in the plant stress
337 responses (Fig. 3), indicating that CLPs are better than exogenous plant hormones in
338 triggering the synthesis of plant defense metabolites. The 77 identified positive metabolites
339 were exported for pathway analysis using the MetaboAnalyst 4.0 software based on the
340 *Arabidopsis thaliana* database since no citrus plant database was available (Chong et al.,
341 2018). Totally 39 pathways matched to the 77 positively identified metabolites. As shown in
342 Fig. 4, each pathway contained an impact and a p-value according to numbers of ‘hits’
343 recorded and the significance factors of the detected metabolites. The highest pathway impact
344 and statistical significance was located on the top right corner (Fig. 4). The metabolism of
345 glycine, serine and threonine, which plays an important role in postharvest stress response,
346 had the highest pathway impact and statistical significance. The activation of some other
347 metabolites was also detected in pathways that play a role in postharvest stress responses,
348 including alanine, aspartate and glutamate metabolism, beta alanine metabolism, isoquinoline
349 alkaloid biosynthesis, tyrosine metabolism, pantothenate and CoA biosynthesis, carbon
350 fixation in photosynthetic organisms, aminoacyl-tRNA biosynthesis, phenylpropanoid
351 biosynthesis, pyrimidine metabolism, glutathione metabolism and lysine biosynthesis (Fig.
352 4).

353 Since the glycine, serine and threonine pathway and the tyrosine pathway contain vital
354 amino acids (i.e. tryptophan and tyrosine) which are important for synthesis of secondary
355 metabolites (i.e. serotonin and tyramine) leading to plant defense, these two pathways were
356 further analyzed. The metabolites yet to be identified were shown in light blue boxes, the
357 positively identified metabolites were highlighted in green, and the putatively identified
358 metabolites were highlighted in orange (Figs. 5 and 6). The box plot of each positively or
359 putatively identified metabolite in the pathway was included to show the significant level

360 changes compared to the control group (Figs. 5 and 6). As shown in Fig. 5, six of the 77
361 identified metabolites involve in the metabolism of glycine, serine and threonine, and four
362 metabolites belong to the tryptophan metabolism. In terms of the tyrosine pathway, two
363 positively identified metabolites, i.e., tyrosine and tyramine (in green) and four putatively
364 identified metabolites (in orange) were found (Fig. 6).

365 The crude CLP extract and each CLP obtained from *B. subtilis* ABS-S14 were found to be
366 able to trigger the synthesis of secondary metabolites like serotonin (Fig. 7A) and 5-hydroxy-
367 *N*-methyltryptamine (Fig. 7B) in flavedo tissues. Accordingly, the box plot of tyrosine
368 concentration in the CLPs treated group showed a lower level than those found in the other
369 agent treated groups (Fig. 7C). Tyrosine had been immediately changed to tyramine as
370 validated by the box plot of Tyramine/Tyrosine (Fig. 7D). Thus, some agents in the CLPs
371 group such as crude CLP extract and fengycin treatments were able to highly activate
372 tyramine production, compared to the other groups (Fig. 7E). The comparative metabolic
373 changes of fungal infection, including the treatments of exogenous plant hormones and CLPs
374 on the production of primary and secondary metabolites in tryptophan and tyrosine pathways
375 (Fig. 7), provided evidence that *P. digitatum* had only a minor effect on those metabolites in
376 mandarin fruit. In the same manner, the effects of exogenous plant hormones were similar to
377 fungal infection on those metabolites with the exception that the treatments of SA, MeJA and
378 Et and their controls were able to elicit greater accumulations of tyrosine than the treatment
379 of CLPs and the fungal pathogen in mandarin fruit.

380 **4. Discussion**

381 Metabolomics studies of plant responses to various kinds of stresses have been carried out
382 in order to provide a better understanding of metabolic changes associated with the
383 enhancement of plant resistance, especially postharvest treatments of citrus fruit (Asai et al.,

2017; Servillo et al., 2013). For instance, differential accumulation of significant proteins and metabolites in Satsuma mandarin after heat treatment revealed reactive oxygen species and lignin that played roles in induced fruit resistance to pathogen (Yun et al, 2013). The application of exogenous phytohormones such as SA and MeJA also affected the metabolic changes in citrus leaves during defense response (Asai et al., 2017). The expression of plant defensive genes inducing by *B. subtilis* CLPs was demonstrated in citrus fruit at transcriptional levels (Waewthongrak et al., 2014). However, the action of *B. subtilis* CLPs on stimulating metabolic changes in defense mechanism in postharvest mandarin fruit was still unclear.

In this study, metabolomics was used to differentiate effects of agents including *B. subtilis* CLPs, exogenous plant hormones, and *P. digitatum* on the metabolic changes of postharvest mandarin fruit under wound stress. Mandarin fruit treated with *B. subtilis* CLPs showed significant better healing of the wound sites than *P. digitatum* and exogenous plant hormones (Fig. 1), indicating *B. subtilis* CLPs treatment may trigger a more effective defense mechanism under stress to protect the fruit. We subsequently applied dansylation LC-MS to examine the metabolic differences of the amine/phenol submetabolome in the wound tissues to study the affected metabolic pathways in the host plant, with the focus on defense responses.

Consistent with the healing observation (Fig. 1), these agents had different levels of capability to trigger metabolite accumulations in the treated flavedo tissues. A close relationship between the exogenous plant hormones and wounding effects on metabolite induction was observed in their treatments at the early time point (24 h post-treatment) and the late time points (48 and 72 h post-treatment) (Fig. 2C). The exogenous plant hormones and wounding effects might share a common pathway in their metabolite regulation function in mandarin peel (Fig. 2B and 2C). Our results were consistent with a previous study in citrus

409 leaves, which demonstrated that both tryptophan and serine were highly sensitive to stress
410 treatments (Asai et al., 2017). Moreover, the wounding and MeJA treatment showed that
411 amino acid abundances such as tryptophan were up-regulated with serine being down-
412 regulated, indicating that the *de novo* synthesis of tryptophan occurred due to the conversion
413 of serine and indole (Asai et al., 2017).

414 The metabolomic analysis showed large metabolic changes in the wound tissues treated
415 with the *B. subtilis* CLP extract and its major bioactive compounds fengycin, iturin A and
416 surfactin (Fig. 3C). These agents appeared to be able to powerfully trigger metabolic changes
417 in the flavedo tissues in responses to wound stress during the postharvest period. This
418 metabolomic finding was consistent with a previous study which demonstrated that *B. subtilis*
419 ABS-S14 CLPs elicited the expression of defense-related genes and the accumulation of
420 enzymes in the induced systemic resistance system (Waewthongrak et al., 2014). Our work
421 also indicated that the metabolites specifically induced by CLP extracts and an individual
422 agent in *B. subtilis* CLPs were involved in twelve major pathways: the glycine, serine and
423 threonine metabolism and the tyrosine metabolism (Fig. 4).

424 Glycine, serine and threonine metabolism may play significant physiological roles in some
425 aspects of stress response in mandarin flavedo tissues induced by bacterial CLPs. On the
426 other hand, phenylalanine, tryptophan, and tyrosine were reported to accumulate after
427 wounding and MeJA and SA treatments (Asai et al., 2017), which were consistent with our
428 finding that the treatments of exogenous plant hormones after wounding increased the
429 abundance of tryptophan and tyrosine. Tryptophan and tyrosine metabolisms might be
430 involved in key regulation in response to stresses. The overall functional categories of each
431 amino acid and their derivatives of glycine, serine and threonine metabolism found to be
432 significant in this metabolomic study (Fig. 5). Specifically, aspartic acid not only acts as the
433 precursor for the synthesis of asparagine which mediates nitrogen transport and storage in

434 plants, but also serves as the precursor for generating aspartate-derived amino acids which are
435 located in leaves, seeds and roots (Azevedo et al., 2006). Moreover, homoserine can activate
436 plant growth (Palmer et al., 2014) and threonine, a substrate of threonine deaminase in α -
437 ketobutyrate and ammonia biosynthesis, is involved in plant defense (Gonzales-Vigil et al.,
438 2011). In addition, serine was one of the most important amino acids involving in the
439 photorespiratory glycolate pathway for plant metabolism and development (Ros et al., 2014),
440 glycine is involved in plant development via the root system (Domínguez-May et al., 2013),
441 and tryptophan is a precursor for generating secondary metabolites in plant immunity (Dixon
442 and Paiva, 1995; Servillo et al., 2013).

443 Previous studies also showed that tryptophan has important role in plant defense (Ishihara
444 et al., 2008; Servillo et al., 2013) and acts as the precursor of serotonin (Ishihara et al., 2008).
445 Surprisingly, serotonin and its derivative 5-hydroxy-*N*-methyltryptamine were significantly
446 up-regulated in the treatment group of *B. subtilis* CLPs (Fig. 7A and B). The bacterial CLPs
447 treatment could activate the serotonin and 5-hydroxy-*N*-methyltryptamine synthesis at a
448 higher level than the other treatments at all time-points (Fig. 7A and B). Serotonin is involved
449 in the plant defense mechanism (Servillo et al., 2015). Specifically, it serves as a substrate of
450 peroxidase to create polymers like lignin that functions as a physical barrier to inhibit the
451 spread of the pathogen infection (Ishihara et al., 2008). Even though the direct function of 5-
452 hydroxy-*N*-methyltryptamine in plant defense is still unclear, this metabolite may act as a
453 precursor for glucosylated serotonin derivatives which can produce toxic aglycones to attack
454 the pathogen via glycosidase activity (Servillo et al., 2015).

455 The present study also found that tyrosine abundance was down-regulated in *B. subtilis*
456 CLP treatment group since it was acting as a precursor of tyramine (Fig. 7D). Therefore, the
457 role of tyramine was confirmed as the precursor of tyramine derivatives that are involved in
458 the pathways of the plant defense mechanism (Servillo et al., 2017). Moreover, tyramine

459 derivatives were coproduced with specialized defensive secondary metabolites, such as toxic
460 substances produced by the plant, to attack a pathogen in response to an infection (Servillo et
461 al., 2014). The large abundance of tyrosine induced by exogenous plant hormones might
462 serve as a precursor to produce downstream substances in the pathway (Fig. 6), but it did not
463 influence tyramine accumulation (Fig. 7E). In addition, unlike *B. subtilis* CLPs, *P. digitatum*
464 and exogenous plant hormones did not stimulate the synthesis of serotonin, 5-hydroxy-*N*-
465 methyltryptamine, tyrosine or tyramine in the metabolism of tryptophan and tyrosine in
466 mandarin fruits under wound stress.

467 Extracts of *B. subtilis* ABS-S14, exogenous phytohormones, and *P. digitatum* infection
468 affected the accumulation of primary and secondary metabolites to stress responses in
469 mandarin fruit during storage period. Each amino acid can be the precursor to produce a
470 number of metabolites. The function of each metabolite could be affected in many pathways
471 (Dinkeloo et al., 2018). The altered levels of some amino acids could impact the systems of
472 host plant such as in defensive pathway (Zeier, 2013), e.g., serotonin could be induced by the
473 application of *B. subtilis* ABS-S14 CLPs, but not by *P. digitatum*. To increase or decrease
474 the production of amino acids, stress is one of the factors which can trigger the mechanisms
475 of host plant (Tunsagool et al., 2019). Thus, amino acids are the most changed pathway of
476 different treatments. Moreover, the changes of metabolites on flavedo tissues by these agents
477 resulted in healing ability of wound sites. The greater effect of wound healing was observed
478 in the group of *B. subtilis* ABS-S14 CLP treatments than those of other agents. *Bacillus* CLPs
479 employed the metabolisms of glycine, serine and threonine, tryptophan, and tyrosine to
480 increase the production of secondary metabolites such as serotonin and tyramine during
481 stresses, leading to the induction of plant immunity. According to the factors of green mold
482 infection in mandarin fruit, wound sites on mandarin peel were the main factor for *P.*
483 *digitatum* growth. Healing ability by the effect of preventive applications of *B. subtilis* ABS-

484 S14 CLPs could reduce the chance of green mold infection in mandarin fruit. Taking together,
485 the healing outcome observed, in combination with the understanding of metabolic changes
486 in flavedo tissues by the treatments of *B. subtilis* ABS-S14 CLPs, suggested this extract as an
487 alternative agent for the protection of mandarin fruit against green mold pathogen invasion
488 during postharvest storage.

489 **5. Conclusions**

490 In summary, application of extracted CLP compounds from *B. subtilis* ABS-S14 showed
491 the greatest effect on healing ability of wound sites in mandarin fruit than exogenous plant
492 hormones, indicating that they are promising biological reagents in prevention of green mold
493 rot on mandarin fruit during postharvest period. The metabolome differences of wound
494 tissues treated with different agents were profiled with extensive metabolomic analysis with
495 dansylation LC-MS and the results showed that the applied agents played a role in regulation
496 of primary and secondary metabolites during storage period. The metabolisms of glycine,
497 serine and threonine, tryptophan, and tyrosine were found to be the important signal
498 transductions of the action of *B. subtilis* CLPs in stress responses. An increase of secondary
499 metabolite accumulation such as serotonin and tyramine in CLP treatments involved the
500 elicitation of defensive system of plant resistance.

501

502 **Conflict of interest**

503 The authors declare no competing financial interest.

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515

516 **Appendix A. Supplementary data**

517 Supplementary materials are included.

518

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- 646
- 647

648 **Figure captions**

649 **Fig. 1.** Wound appearance in each treatment at 24, 48, and 72 h post-treatment.

650 **Fig. 2.** PCA plots of the metabolomic data obtained from different groups of samples
651 including QC samples. (A) Color-coded groups according to the treatments at all
652 time points and the QC group. (B) Color-coded groups according to the treatment
653 times of 24, 48 and 72 h from four major groups of treatments. (C) Color-coded sub-
654 groups within the control and exogenous plant hormone groups. W, sterile distilled
655 water; E, ethanol; Pd, *Penicillium digitatum*; SA, salicylic acid; MeJA, methyl
656 jasmonate; Et, ethephon; CE, crude CLP extract; F, fengycin; I, iturin A; S, surfactin;
657 CLPs, cyclic lipopeptide group; PH, exogenous plant hormones; 24, 24 h post-
658 treatment; 48, 48 h post-treatment; 72, 72 h post-treatment.

659 **Fig. 3.** Volcano plots of binary comparisons of metabolites in treatment vs. control. (A)
660 *Penicillium digitatum* treatment group vs. water group. (B) Exogenous plant
661 hormone treatment group vs. ethanol group. (C) CLP treatment group vs. ethanol
662 group. The significant metabolites are shown in red or blue with fold change > 1.5
663 and q-value <0.05 with corresponding P <0.134 in (A), P <0.014 in (B) and P
664 <0.216 in (C).

665 **Fig. 4.** Overview of metabolic pathway analysis relating to stress responses.

666 **Fig. 5.** Metabolic pathway of glycine, serine and threonine metabolism including tryptophan
667 metabolism. The metabolite in green box represents the positive ID, orange box
668 represents the putative ID, and blue box represents no ID, and the box plots of
669 identified metabolites are displayed beside the corresponding metabolites. E, ethanol;
670 CLPs, cyclic lipopeptide group.

671 **Fig. 6.** Metabolic pathway of tyrosine metabolism. The metabolite in green box represents
672 the positive ID, orange box represents the putative ID, and blue box represents no ID,

673 and the box plots of identified metabolites are displayed beside the corresponding
674 metabolites. E, ethanol; CLPs, cyclic lipopeptide group.

675 **Fig. 7.** Box plots of positively identified metabolites related to plant defense pathway: (A)
676 serotonin, (B) 5-hydroxy-*N*-methyltryptamine, (C) tyrosine, (D) tyramine/tyrosine,
677 and (E) tyramine. W, sterile distilled water; E, ethanol; Pd, *Penicillium digitatum*;
678 SA, salicylic acid; MeJA, methyl jasmonate; Et, ethephon; CE, crude CLP extract; F,
679 fengycin; I, iturin A; S, surfactin. Vertical bars represent standard errors of the mean
680 value of six trials.











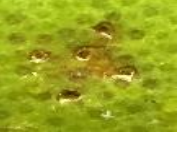



















Treatments	24 h	48 h	72 h
Sterile distilled water			
<i>Penicillium digitatum</i>			
Ethanol			
Salicylic acid			
Methyl jasmonate			
Ethephon			
CLP extract			
Fengycin			
Iturin A			
Surfactin			

Figure 1

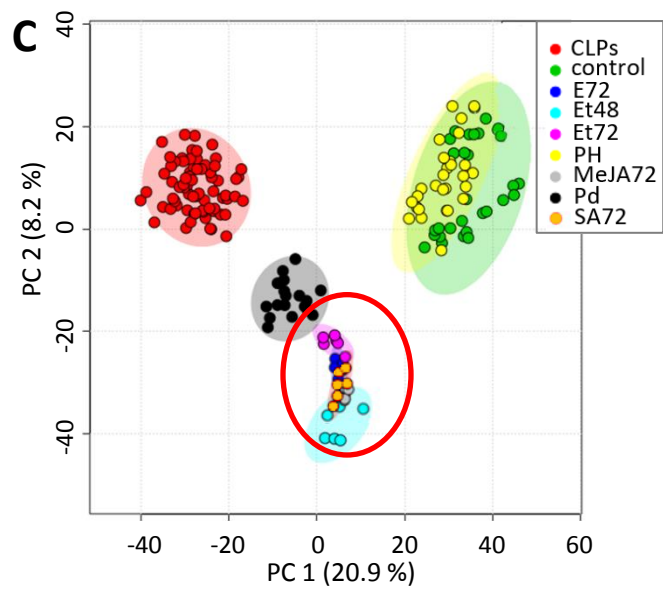
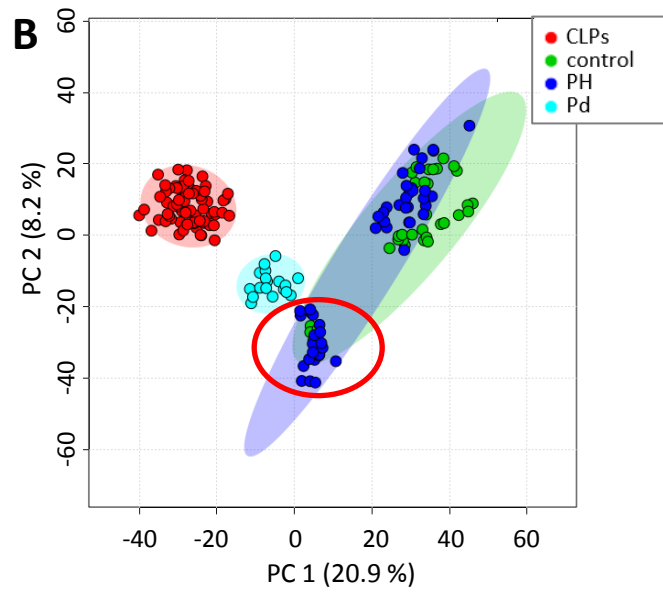
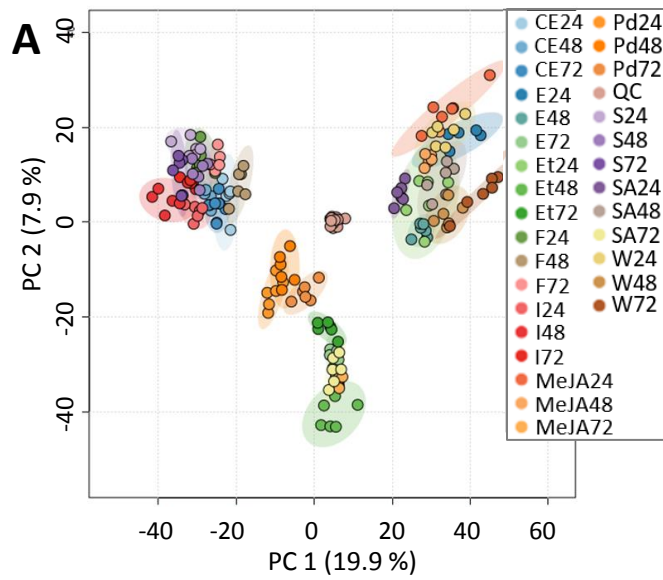


Figure 2

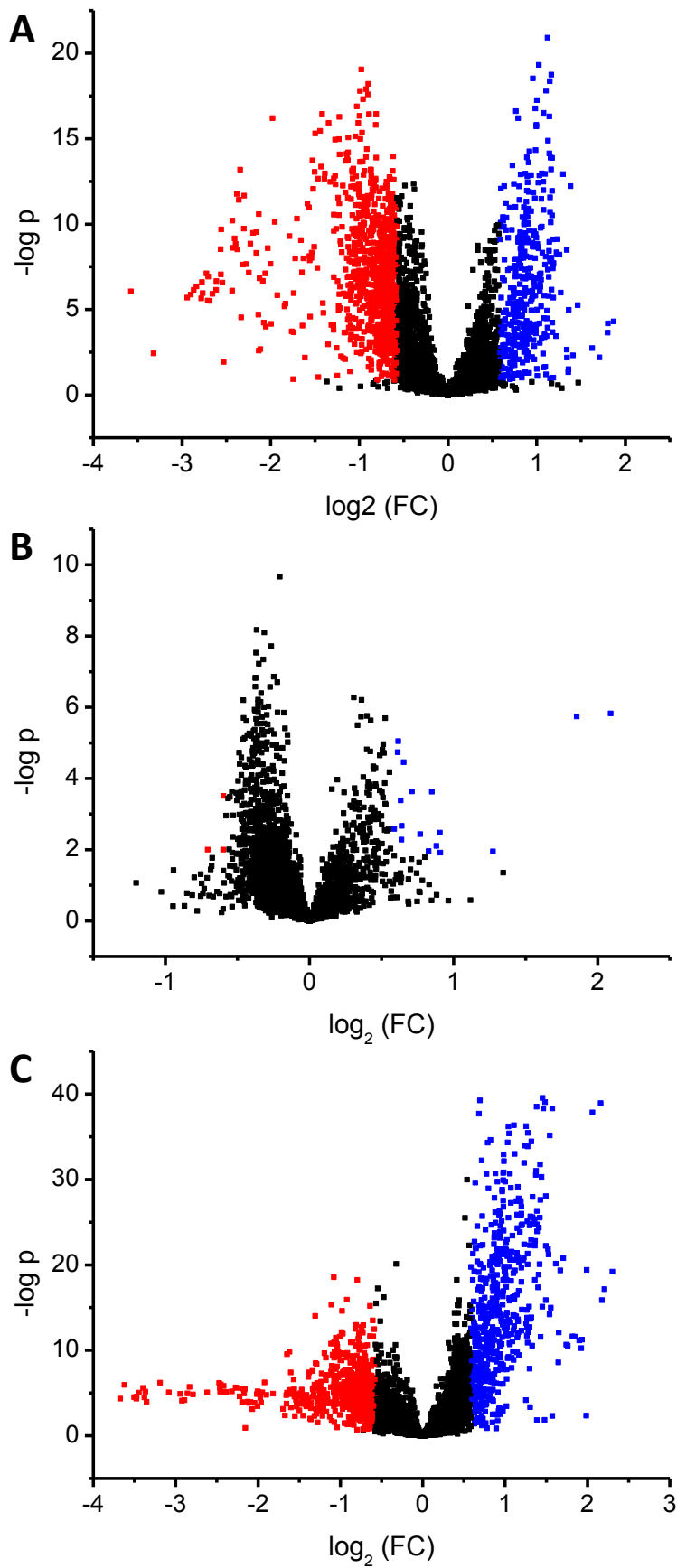
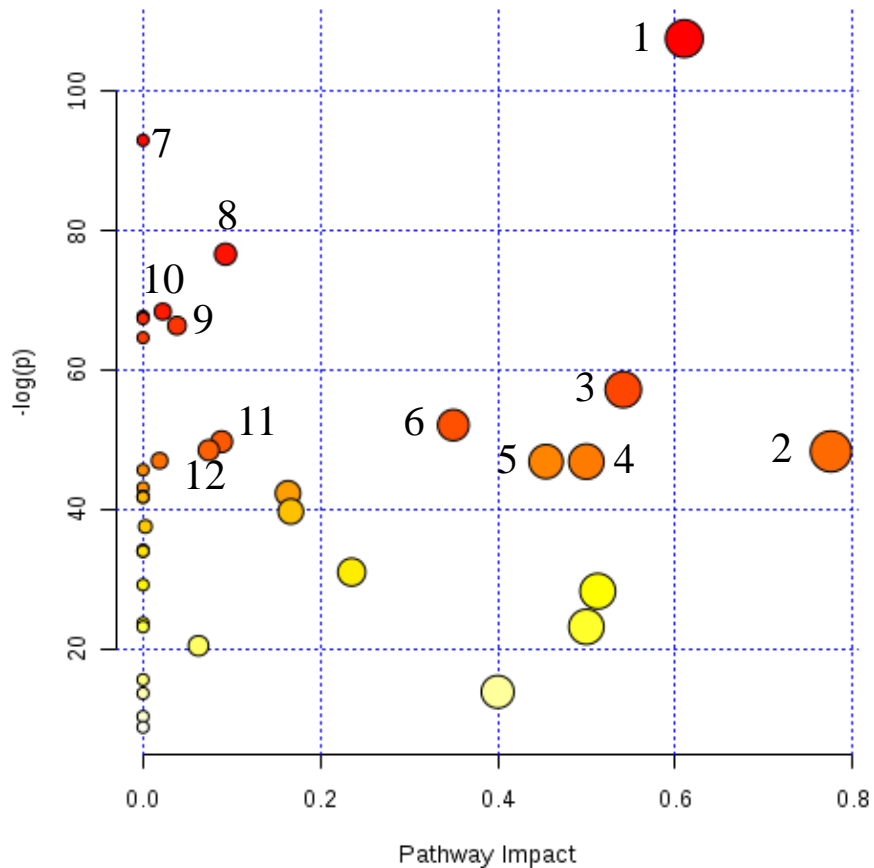


Figure 3



1	Glycine, serine and threonine metabolism
2	Alanine, aspartate and glutamate metabolism
3	Beta alanine metabolism
4	Isoquinoline alkaloid biosynthesis
5	Tyrosine metabolism
6	Pantothenate and CoA biosynthesis
7	Carbon fixation in photosynthetic organisms
8	Aminoacyl-tRNA biosynthesis
9	Phenylpropanoid biosynthesis
10	Pyrimidine metabolism
11	Glutathione metabolism
12	Lysine biosynthesis

Figure 4

Glycine, serine and threonine metabolism

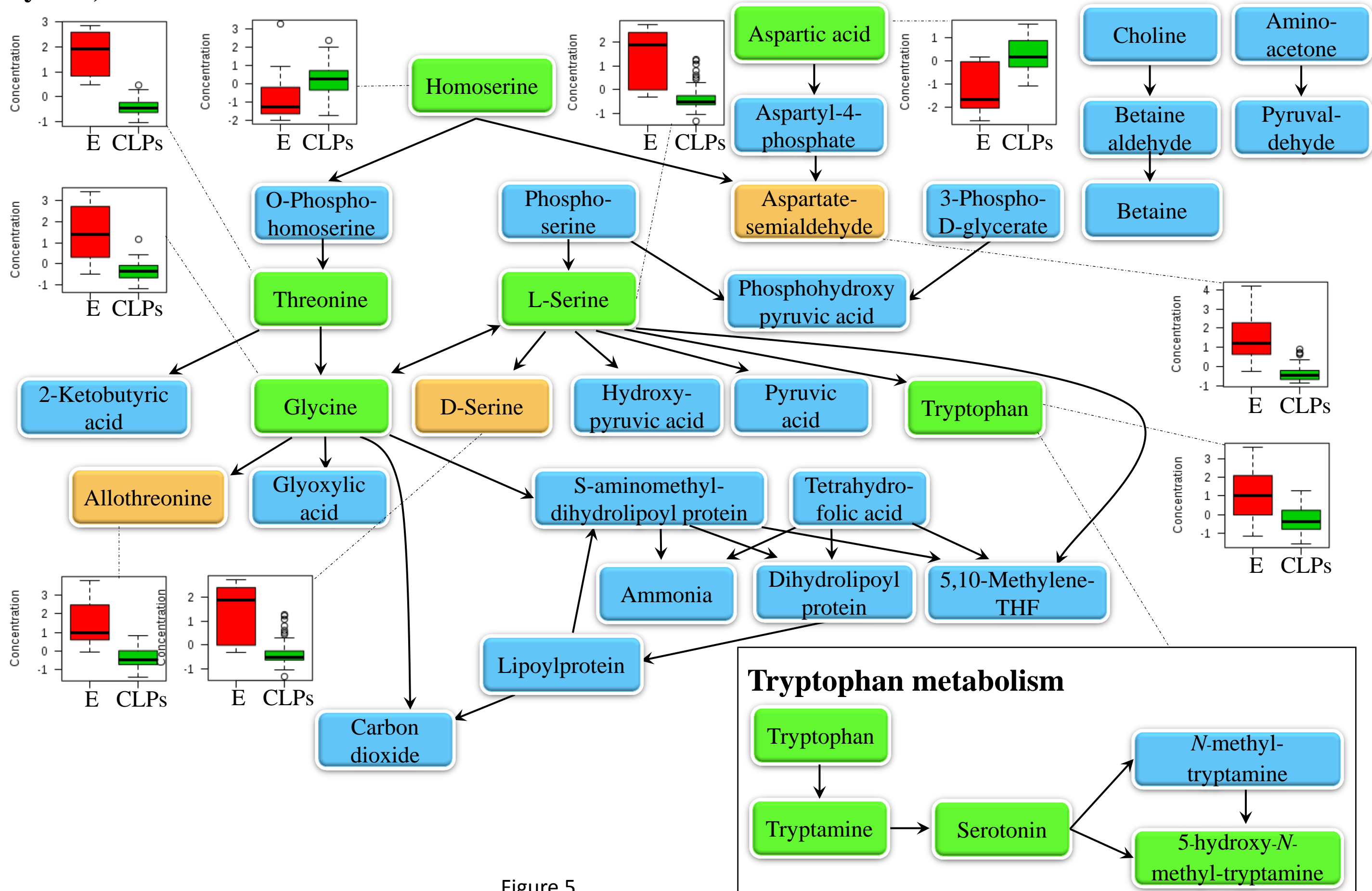


Figure 5

Tyrosine metabolism

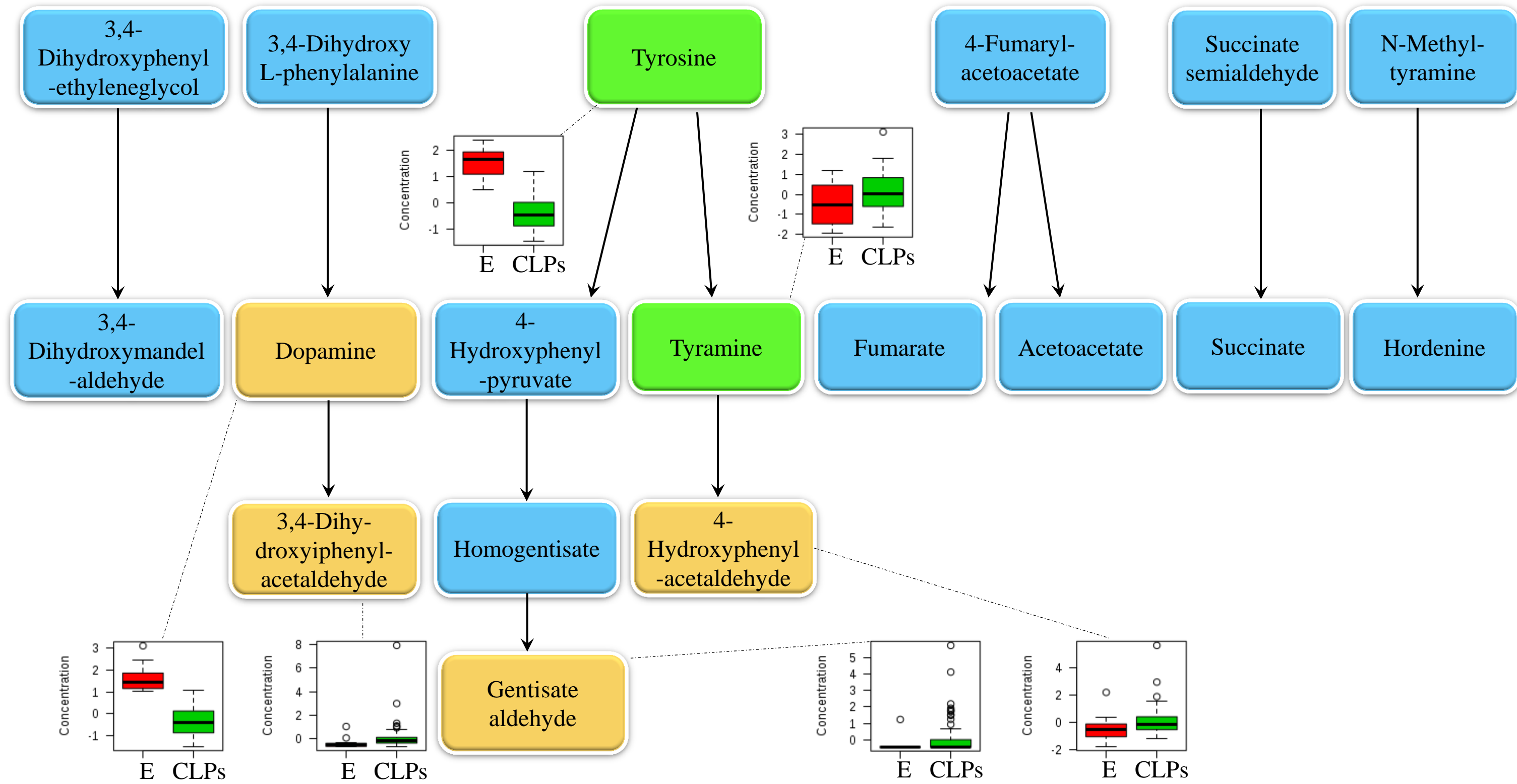


Figure 6

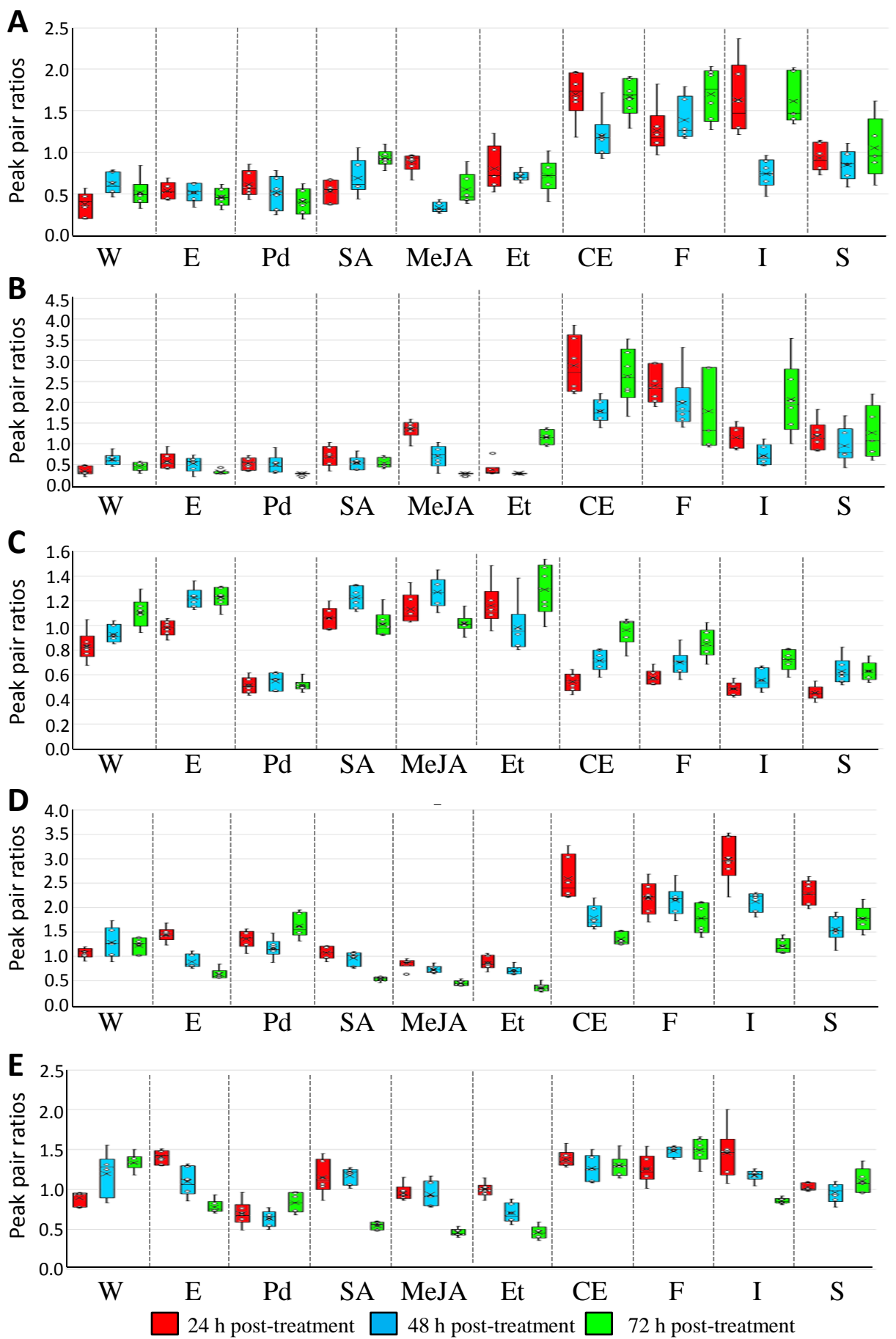
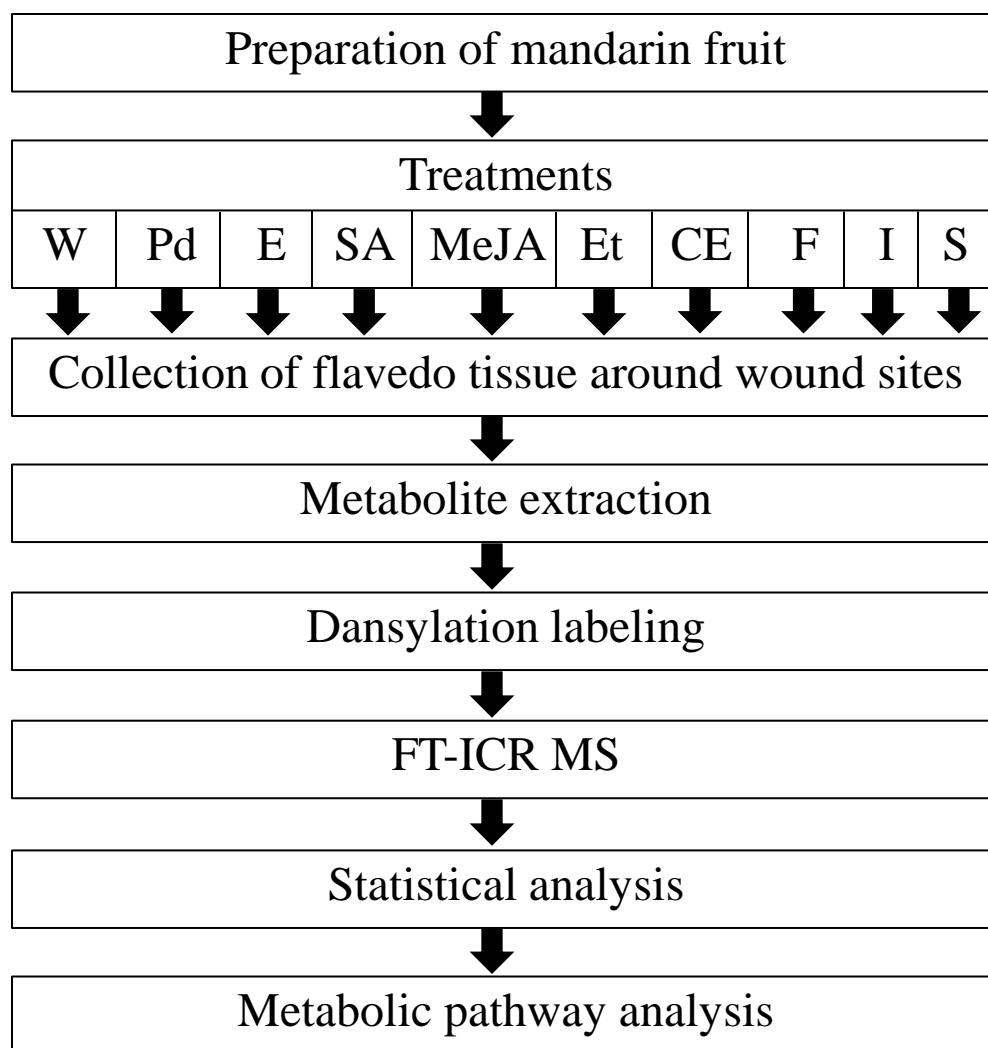
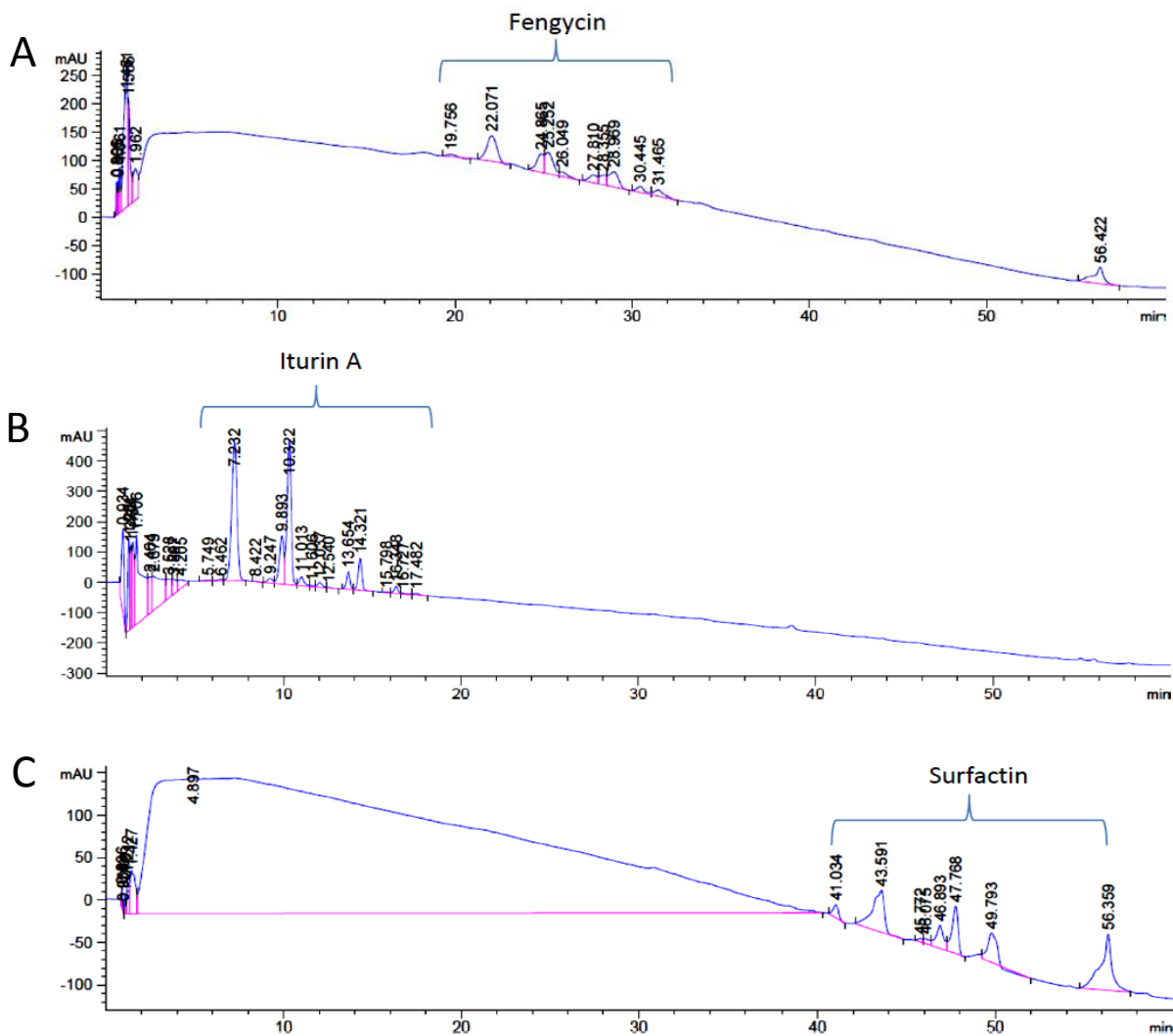


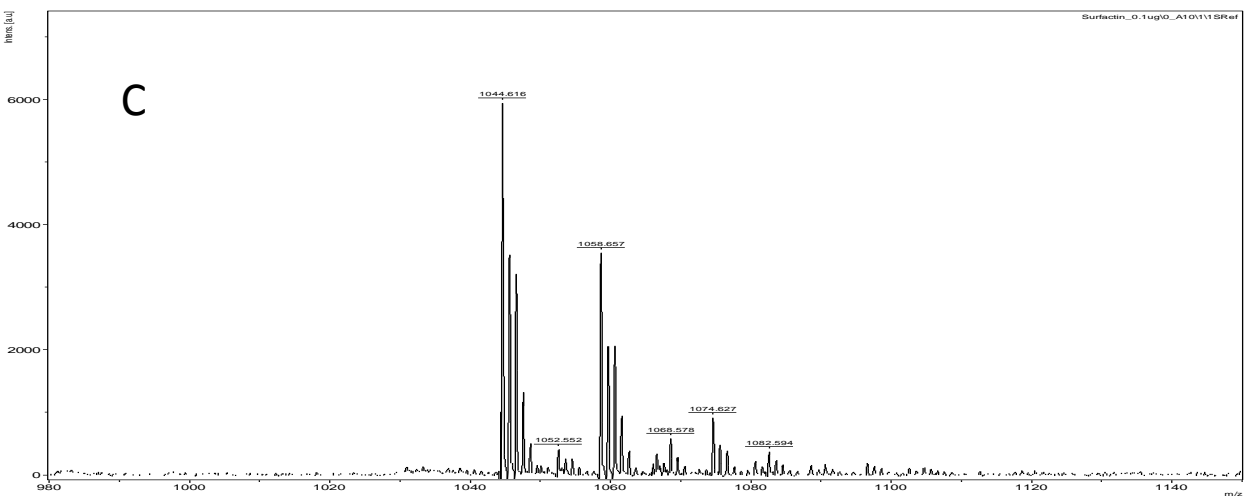
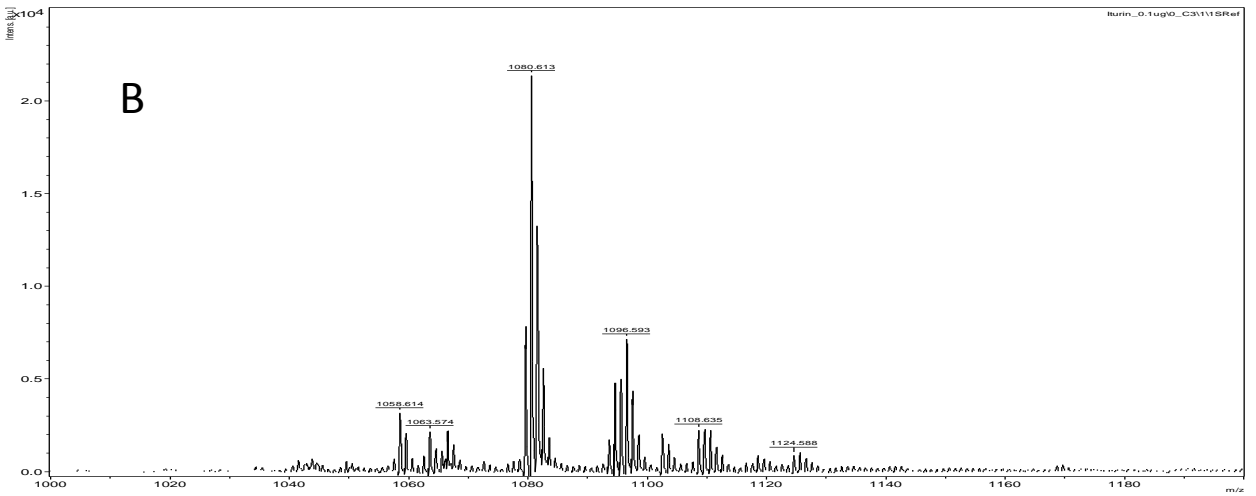
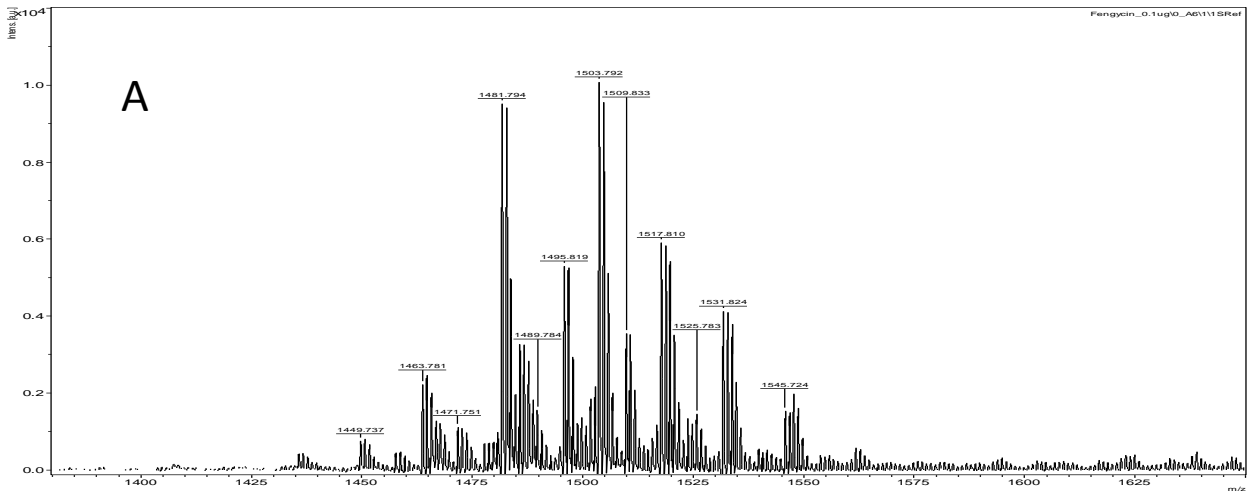
Figure 7



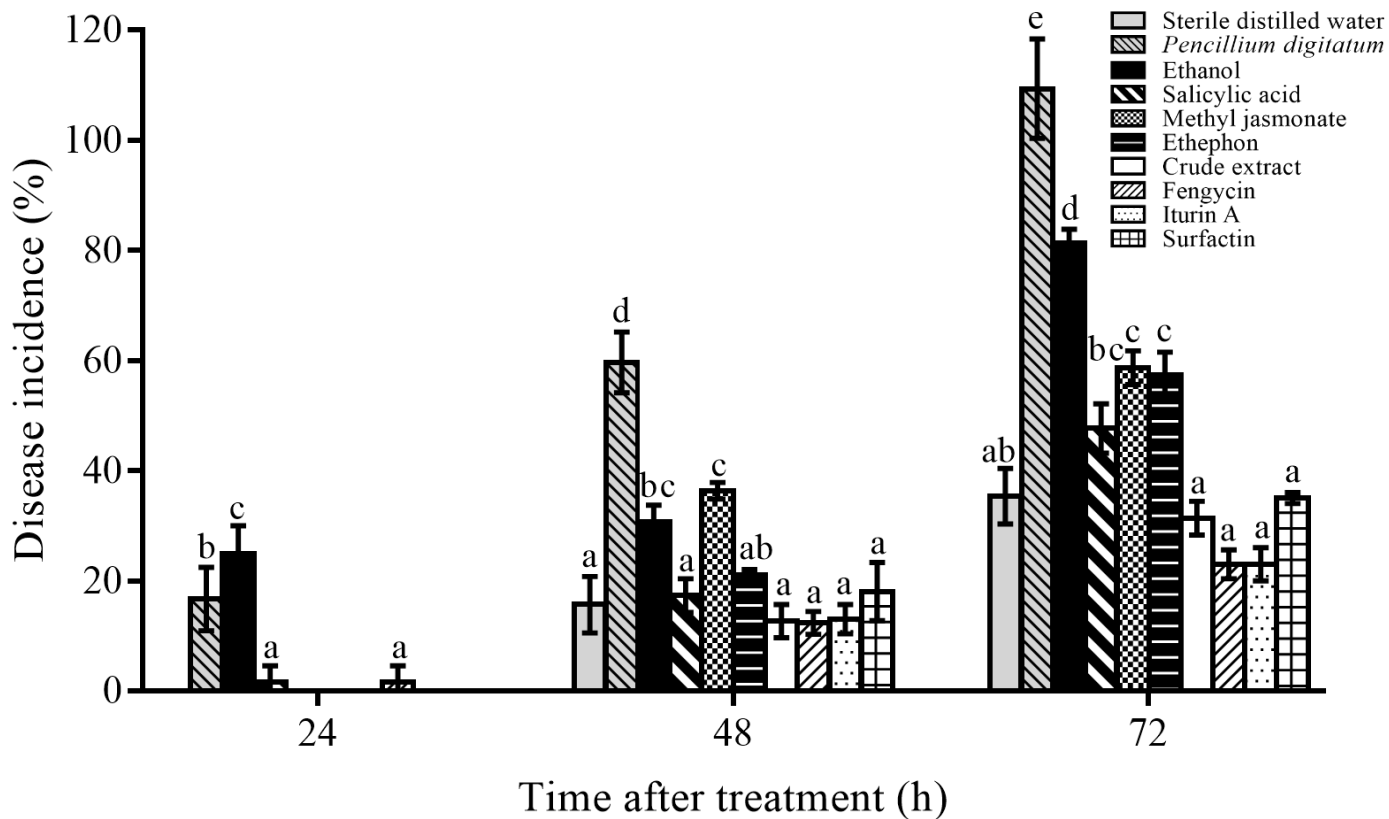
Supplementary Fig. 1. Workflow of the study in the treatments of sterile distilled water (W), *Penicillium digitatum* (Pd), ethanol (E), salicylic acid (SA), methyl jasmonate (MeJA), ethephon (Et), CLP extract (CE), fengycin (F), iturin A (I), and surfactin (S).



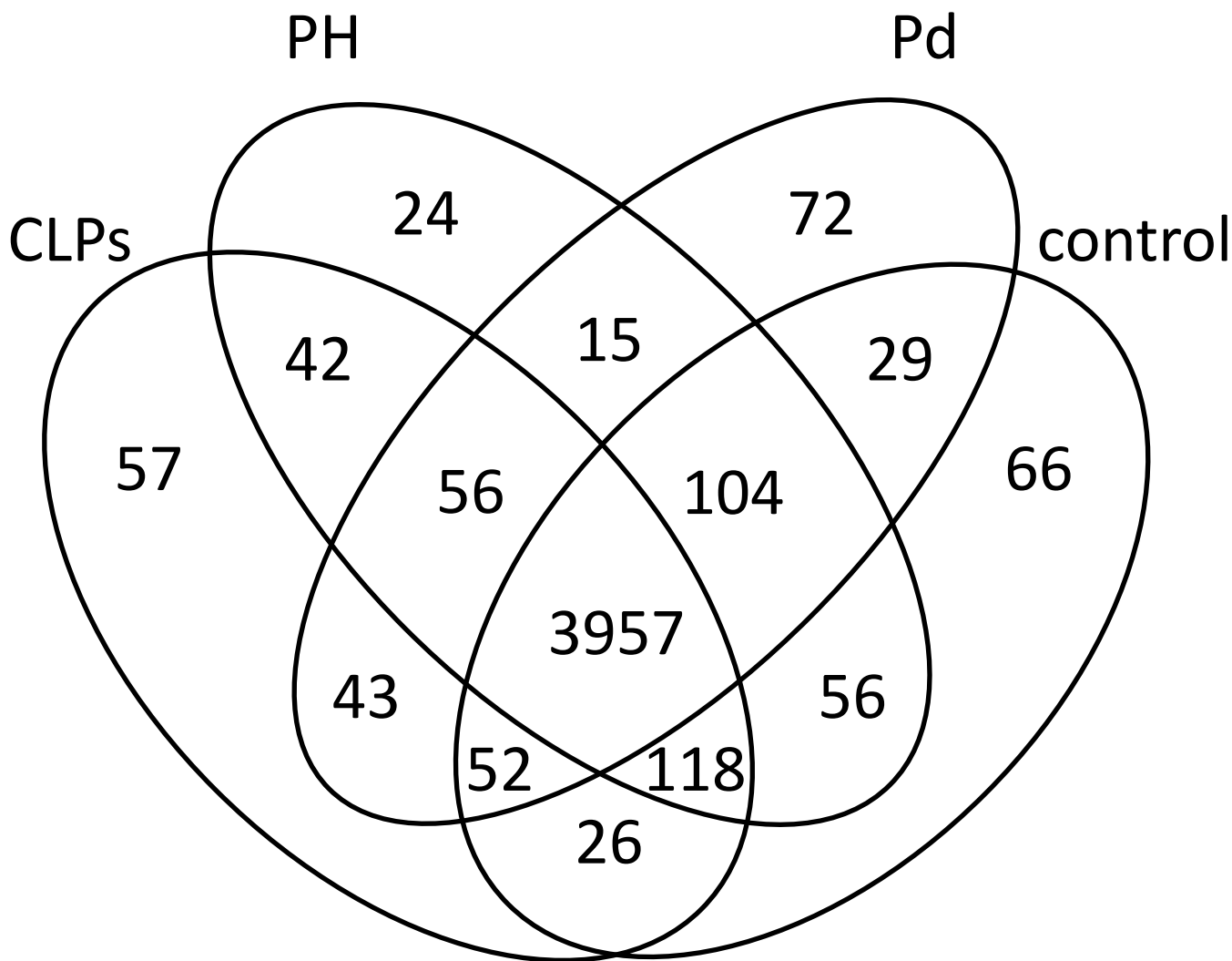
Supplementary Fig. 2. HPLC-UV analysis of bacterial CLPs.



Supplementary Fig. 3. MALDI-TOF MS spectra of bacterial CLPs. Details such as peak masses can be visualized by enlarging the figure.



Supplementary Fig. 4. Disease incidence in mandarin fruit of various treatments. Error bars represent standard errors of the mean value of three trials. Bars with the same letter above them show no significant difference to each other using a criterion of $p \leq 0.05$ as significant difference according to Tukey's range test. For example, all the bars marked with "a" in the group of 48 hours after various treatments do not show significant differences.



Supplementary Fig. 5. Venn diagram showing the metabolite number distribution in the four major groups of treatments. C, control stress; CLPs, cyclic lipopeptide group; PH, exogenous plant hormones; Pd, *Penicillium digitatum*.