Transcriptomic analysis reveals flavonoid biosynthesis of Syringa oblata Lindl. in response to different light intensity

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Abstract

Background: Hazy weather significantly increase air pollution and affect light intensity which may also affect medicinal plants growth. Syringa oblata Lindl. (S. oblata), an effective anti-biofilm medicinal plants, is also vulnerable to changes in plant photoperiods and other abiotic stress responses. Rutin, one of the flavonoids, is the main bioactive ingredient in S. oblata that inhibits Streptococcus suis biofilm formation. Thus, the present study aims to explore the biosynthesis and molecular basis of flavonoids in S. oblata in response to different light intensity.

Results: In this study, it was shown that compared with natural (Z0) and 25% ~ 35% (Z2) light intensities, the rutin content of S. oblata under 50% ~ 60% (Z1) light intensity increased significantly. In addition, an integrated analysis of metabolome and transcriptome was performed using light intensity stress conditions from two kinds of light intensities which S. oblata was subjected to: Z0 and Z1. The results revealed that differential metabolites and genes were mainly related to the flavonoid biosynthetic pathway. We found out that 13 putative structural genes and a transcription factor bHLH were significantly up-regulated in Z1. Among them, integration analysis showed that 3 putative structural genes including 4CL1, CYP73A and CYP75B1 significantly up-regulated the rutin biosynthesis, suggesting that these putative genes may be involved in regulating the flavonoid biosynthetic pathway, thereby making them key target genes in the whole metabolic process.

Conclusions: The present study provided helpful information to search for the novel putative genes that are potential targets for S. oblata in response to light intensity.

Keywords: Syringa oblata Lindl., Rutin, Flavonoid biosynthetic pathway, Light intensity, Metabolite, Transcriptome

Background

Hazy weather is a normal event in China [1] that can severely increase the occurrence of air pollution [2]. It is characterized by scattering of particles which constitute the main factor for visibility degradation [3, 4]. On a longer timescale, these pollutants from China may affect North America, the Pacific and the Arctic, making Chinese air pollution a global problem [5]. Its direct result is that plants receive less light than in better weather condition. However, the plants in their natural habitats need to modify their growth and development to suit various environmental conditions including light signals [6], cold hardiness, drought resistance and shade abilities [7]. High light intensity [8] and UV-B radiation [9] also play an important role in the regulation of the flavonoids biosynthesis which is one of the most intensively studied area [10, 11]. It is well-known that flavonoids are the main secondary metabolites in medicinal plants in nature and have various biological functions, such as flower pigmentation, pollen fertility, plant microbe interaction, protection from UV radiation [12], antioxidant functions [13], anti-fungal and anti-bacterial properties [14]. Rutin is one of the well-known flavonoids in plants and has been reported as the main bioactive ingredient.
in *S. oblata* that inhibits *Streptococcus suis* (*S. suis*) biofilm formation [15]. Thus, it is unknown whether the concentration of flavonoids in *S. oblata* is affected by different light intensity groups and the effect of secondary metabolites against the formation of biofilm by *Staphylococcus xylosus* (*S. xylosus*) remains unexplored.

Recently, metabolomics analysis and transcriptomics [16, 17] have been successfully applied as connection networks in identifying differential gene expression in several plants, including *Yukon thellungiella* [18], sunflower [19] and *Nicotiana tabacum* [20]. However, no studies on bio-information analysis and functional identification have been published exploring the molecular mechanism of the secondary metabolite of flavonoid biosynthesis in the absence of *S. oblata* genomic information. Transcriptome sequencing [21, 22] is a rapid technique for obtaining functional genomic information that is widely used to determine gene structures and expression profiles in medicinal plants. Nevertheless, de novo assembly of RNA-Seq data makes it possible to conduct gene analysis in the absence of reference genomes [23, 24].

In this study, biofilm formation and flavonoids biosynthesis were examined under different light intensity groups in *S. oblata*. First, the changes in rutin content and the effect of *S. oblata* against the formation of biofilm by *S. xylosus* were tested in vitro in different months and different light intensity groups. Then, under the different light intensity groups, the presence of flavonoids in *S. oblata* were detected using histochemical method. Finally, comparative metabolomic and transcriptomic analysis were performed to identify the differentially expressed metabolites and genes. To the best of our knowledge, the study offers a new approach to the use of multi-omics technology for high-throughput sequencing in elucidating the molecular mechanism underlying the changes in the rutin content and flavonoid accumulation in *S. oblata* under different light intensity groups.

**Results**

**Biofilm formation ability of *S. xylosus***

The study involves the evaluation of the biofilm inhibitory properties of rutin and *S. oblata* extract against *S. xylosus* biofilm formation during the period of different months and light intensity groups (Fig. 1a and Fig. 2a). The results revealed that 0.8 mg/mL of rutin significantly inhibited biofilm formation compared with the control (*p* < 0.05) (Fig. 1b). And the results showed that the MICs of all the different months of *S. oblata* against *S. xylosus* was 62.5 mg/mL, and 1/2 MIC (31.25 mg/mL) of the different months was able to significantly inhibit biofilm formation compared with the positive control (*p* < 0.05) (Fig. 1c). Furthermore, the MICs of *Z*₀, *Z*₁ and *Z*₂ of *S. oblata* against *S. xylosus* were 62.5 mg/mL, 31.25 mg/mL and 31.25 mg/mL, respectively. Among them, the results revealed that in comparison with positive control, 1/2 MIC (15.625 mg/mL) of *Z*₁ and *Z*₂ significantly inhibited *S. xylosus* biofilm formation as against 1/2 MIC (31.25 mg/mL) of *Z*₀ (*p* < 0.05) (Fig. 1d). This indicated that *S. oblata* had the highest ability to inhibit the formation of *S. xylosus* biofilm only when subjected to light intensity.

**Variations in the rutin content**

In order to evaluate whether the content of rutin was influenced by the 6 months (May to October) and three light intensity groups (*Z*₀, *Z*₁ and *Z*₂), two experiments were conducted. The trends of the daily sunshine, rainfall, and atmospheric temperature all through the months in the first experiment are shown in Fig. 1e, f. It revealed that the rutin content in May was 1.3331 ± 0.5612 mg/g, and then it increased to the highest value (11.0787 ± 0.9570 mg/g) in August. Then, it decreased in September (5.0921 ± 1.8441 mg/g) and October (3.6752 ± 0.7840 mg/g) (Fig. 1g). At the same time, the trend of the daily sunshine and rain fall all through the months were negatively and positively correlated with the content of rutin, respectively (Fig. 1g). The air pollution indices in August and September were more excellent than those obtained in other months, the values were obtained from the Website (https://www.aqistudy.cn/historydata/) (Additional file 1: Table S1). At the same time, lower particles such as ambient particulate matter (PM 2.5 and PM 10) could enhance solar radiation, thus enhancing ambient temperature and biological plant growth rates [25]. This finding is in agreement with previously findings reported on the content of rutin in Fructus Sophorae which peaked in mid-August, then declined gradually [26]. Furthermore, in the second experiment, the photosynthetically active radiation (PAR) and light reflection values of *Z*₁ and *Z*₂ measured by spectrometer were significantly lower than *Z*₀ (Fig. 2b, c). The study revealed that the amount of rutin in *S. oblata* in September significantly increased in *Z*₁ (4.4729 ± 0.7738 mg/g) when compared with *Z*₀ (2.7518 ± 0.2854 mg/g) and *Z*₂ (2.5921 ± 0.5419 mg/g) (Fig. 1h). Due to insufficient exposure of the leaves to light, the flavonoid content decreased after excessive shading which is also the reason why the rutin content in *Z*₂ was lower than *Z*₁. Therefore, *Z*₀ and *Z*₁ of *S. oblata* were selected and used for the next phase of studies [27].

**Histochemical analysis of flavonoids with different light intensity groups**

Flavonoid accumulation in plants can be visualized using DPBA, a reagent, which indicates the presence of many flavonoids in histochemical analysis [28]. Leaves were stained with DPBA. Cross-sections of green fluorescence were observed in *S. oblata* (Fig. 1i, j) using LCSM. This
Fig. 1 (See legend on next page.)
finding is in agreement with earlier findings previously reported [28] that the presence of flavonoids was confirmed in different light intensity groups, especially observable in the epidermal cell layers and vascular bundles [29, 30]. To visualize the accumulation of flavonoids in leaf tissue more quantitatively and precisely, DPBA imagery was used. The results showed that the green fluorescence area of Z₁ (2707423) was also larger than Z₀ (903,872.5) after handling them with ImageJ (https://imagej.nih.gov/ij/index.html) [31].

Metabolites analysis with different light intensity groups
To identify key metabolic alteration after different light intensity groups, the metabolite levels between Z₀ and Z₁ were compared using LC-MS analysis. All data on retention time, exact mass, and peak intensity were recorded for multiple statistical analysis, including principal component analysis (PCA) (Fig. 3a, b) and partial least squares-discriminate analysis (PLS-DA) (Fig. 3c, d). These analytical methods revealed a trajectory of the different light intensity groups by the combination of the two main components. Volcano plot can visually screen for differentially expressed metabolites of S. oblata between the different light intensity groups. All metabolites from secondary mass spectrometry identification statistics were significantly different by ratio ≥ 2 or ratio ≤ 1/2, q-values≤0.05 and VIP ≥ 1 in different ion modes. The result showed that there were 7402 and 9481 total metabolites in the negative and positive ions modes contained as (1212, 1553) up-regulated and (1439, 1749) down-regulated metabolites, respectively (Fig. 3e, f). Furthermore, the secondary metabolite matching to specific biosynthesis pathways in terms of the number of matches in the positive and negative ions modes were enriched and analyzed by the KEGG pathway. The results showed that the secondary metabolite biosynthesis pathways were mainly matched to isoquinoline alkaloid biosynthesis, glucosinolate biosynthesis, and flavonoid biosynthesis (Table 1).

Transcriptome sequencing and data analysis with different light intensity groups
Transcriptome sequencing results and sequence assembly.
In order to understand the potential molecular synthesis mechanisms, four RNA-seq libraries were constructed by Z₀ and Z₁. These RNA-seq libraries were subjected to pair-end reading with the Illumina HiSeq 4000 platform. In addition, since there is no reference genome sequence in *S. oblata*, all clean reads were de novo assembled into contigs using the Trinity software, and reads were mapped back to contigs, redundancy was removed and the longest transcripts were defined as unigene. The final assembly of *S. oblata* had 73,159 unigenes with an N₅₀ length of 1116 nucleotides (nt) (Table 2).

Unigenes sequence functional annotation.

For functional annotation of the *S. oblata* transcriptome, all assembled unigenes of the 73,159 unigenes were BLASTed against 6 public databases, including NCBI_nr, eggNOG, Swiss-Prot, Pfam, KEGG and GO databases using the DIAMOND [32] program with E-value threshold of 1E-5. From the results in Table 3, a total of 34,764 (42.19%) sequences showed high homology to public

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Fig. 3 PCA, PLS-DA and Volcano of natural (Z₀) and 50% ~ 60% (Z₁) light intensity treatment of *Syringa oblata* Lindl. in the negative (a, c, e) and positive (b, d, f) ion modes.
Among them, 36,713 (50.18%), 19,592 (26.78%), 32,939 (45.02%), 29,896 (40.86%), 42,229 (57.72%), 42,374 (57.92%) unigenes were annotated in the GO, KEGG, Pfam, Swiss-Prot, eggNOG and Nr databases, respectively.

### Analysis of the Differentially Expressed Genes (DEGs)

The DEGs of the four transcriptome libraries were identified with significant differences expression under the thresholds of log 2 (Fold-change) over 1 and FDR less than 0.001 with an adjusted p-value < 0.05. In this study, the expression of genes was calculated by TPM. According to the expression differences, a total of 73,159 genes were detected using KEGG pathway analysis, and only 8015 genes displayed significant changes in expression levels between Z0 and Z1. The numbers of up-regulated and down-regulated unigenes were 4568 and 3447, respectively in volcano (Additional file 1: Figures S1, S2).

GO analysis was performed again based on DEGs in Additional file 1: Figure S3. Of the 36,713 unigenes, 26,072 DEGs were divided into 3 GO terms such as biological process (9588), cellular component (7847) and molecular function (8637). Among them, the most frequently annotated genes involved in the biological process were biological process (627), regulation of transcription, DNA-templated (337) and protein phosphorylation (303). The most frequently annotated genes involved in the cellular component were nucleus (1206), plasma membrane (922) and integral component of membrane (708). And also, the most frequently annotated genes involved in the molecular function were molecular function (607), protein serine/threonine kinase activity (390) and ATP binding (348).

The KEGG function annotation were performed and ggplot 2 was used to analyze the KEGG enrichment results as a scatter plot (Fig. 4). The results showed that the top 10 pathways for KEGG enrichment were other types of O-glycan biosynthesis, sesquiterpenoid and triterpenoid biosynthesis, glucosinolate biosynthesis, caffeine metabolism, stilbenoid, diarylheptanoid and gingerol biosynthesis, anthocyanin biosynthesis, flavonoid biosynthesis, vitamin B6 metabolism, diterpenoid biosynthesis and limonene and pinene degradation.

### Integration of metabolites and expressed Unigenes with different light intensity groups

#### Differential metabolites of flavonoid biosynthesis pathway

All metabolites related to flavonoid biosynthesis pathway were significantly differentiated by the ratio > 2 or ratio ≤ 1/2, q-value ≤ 0.05 and VIP ≥ 1. It included rutin, kaempferol, naringin, tras-2-hydroxyacinamic acid, 3,4-dihydroxyhydrocinamic acid, astragal, daidzin, glycitin, and other components.

#### Table 1: KEGG pathway analysis on biosynthesis of other secondary metabolites

| Pathway                                | Number of Compound (Detected) | Pathway ID |
|----------------------------------------|------------------------------|------------|
|                                        | All | Negative | Positive |
| Isoquinoline alkaloid biosynthesis     | 93  | 37       | 64       | map00950 |
| Glucosinolate biosynthesis             | 75  | 19       | 28       | map00966 |
| Flavonoid biosynthesis                 | 68  | 52       | 43       | map00941 |
| Tropane, piperidine and pyridine alkaloid biosynthesis | 68  | 8        | 30       | map00960 |
| Phenylpropanoid biosynthesis           | 66  | 33       | 47       | map00940 |
| Anthocyanin biosynthesis               | 66  | 20       | 44       | map00942 |
| Isoflavonoid biosynthesis              | 63  | 52       | 44       | map00943 |
| Flavone and flavonol biosynthesis      | 49  | 38       | 35       | map00944 |
| Indole alkaloid biosynthesis           | 47  | 25       | 28       | map00901 |
| Monobactam biosynthesis                | 39  | 21       | 22       | map00261 |
| Carbapenem biosynthesis                | 32  | 11       | 11       | map00332 |
| Stilbenoid, diarylheptanoid and gingerol biosynthesis | 25  | 12       | 16       | map00945 |
| Betalin biosynthesis                   | 24  | 10       | 15       | map00965 |
| Caffeine metabolism                    | 21  | 11       | 7        | map00232 |
| Benzoxazinoid biosynthesis             | 9   | 3        | 4        | map00402 |
| Acridone alkaloid biosynthesis         | 7   | 4        | 5        | map001058 |

### Table 2: Summary of assembly results of Syringa oblata Lindl

| Index | All | GC% | Min Length | Median Length | Max Length | Total Assembled Bases | N50 |
|-------|-----|-----|------------|---------------|------------|-----------------------|-----|
| Transcript | 162,354 | 41.20 | 201 | 575.00 | 10,795 | 134,100,053 | 1225 |
| Gene   | 73,159 | 41.63 | 201 | 430  | 10,795 | 52,090,718    | 1116 |
L-phenylalanine, N-acetyl-, luteolin, quercetin 3′-methyl ether and so on (Table 4).

Differentially expressed Unigenes of flavonoid biosynthesis pathway.

Based on metabolite results, the content of rutin in *S. oblata* was significantly increased at Z1 compared with Z0. At the same time, combined with the results of transcriptome sequencing, it was observed that differential metabolite and DEGs were both related to the flavonoid biosynthetic process by GO and KEGG annotation analysis. Thus, based on TPM, 13 putative structural genes and 1 putative regulatory gene involved in the flavonoid biosynthesis pathway were screened as shown in Fig. 5. The results showed that a total of 13 putative structural genes such as phenylalanine ammonia-lyase (*PAL*, TRINITY_DN36967_c0_g1), 4-coumarate-CoA ligase (*4CL1*, TRINITY_DN35155_c0_g1), trans-cinnamate 4-monooxygenase (*CYP73A*, TRINITY_DN29851_c0_g2), two shikimate O-hydroxycinnamoyltransferase (*HST*, TRINITY_DN31867_c0_g1; *HST*, TRINITY_DN31867_c0_g2), three chalcone synthase (*CHS*, TRINITY_DN35859_c0_g3; *CHS*, TRINITY_DN35859_c0_g2; *CHS*, TRINITY_DN32109_c0_g3), naringenin 3-dioxygenase (*FHT*, TRINITY_DN36563_c0_g1), three flavonol synthase (*DLO2*, TRINITY_DN36038_c0_g2; *DMR6*, TRINITY_DN28124_c3_g2; *SGR1*, TRINITY_DN29800_c0_g2), flavonoid 3′-monooxygenase (*CYP75B1*, TRINITY_DN31579_c2_g5), and transcription factor *bHLH* (TRINITY_DN28965_c1_g8) significantly up-regulated the metabolite of rutin in *S. oblata* with Z1.

Correlation networks analysis between differential metabolites and differentially expressed Unigenes.

Pearson partial correlation analysis was used to conduct the analysis on the significant correlation networks (r > 0.9 or r < 0.9, p < 0.05). The genes and metabolic network properties constructed by 13 putative structural genes and 11 differential metabolites between Z0 and Z1 could be seen in Fig. 6. The results showed that metabolites of rutin were positively correlated with *4CL1* (TRINITY_DN35155_c0_g1), *CYP73A* (TRINITY_DN29851_c0_g2), *SGR1* (TRINITY_DN29800_c0_g2) and *CYP75B1* (TRINITY_DN31579_c2_g5) under the negative ion mode. These 4 putative genes were negatively correlated and positively correlated with 4 metabolites (rutin, kaempferol, naringin and tras-2-hydroxycinnamic acid) and 7 metabolites

**Table 3** Summary of functional annotation of *Syringa oblata* Lindl

| DB       | Num   | Ratio (%) |
|----------|-------|-----------|
| All      | 73,159| 100.00    |
| GO       | 36,713| 50.18     |
| KEGG     | 19,592| 26.78     |
| Pfam     | 32,939| 45.02     |
| swissprot| 29,896| 40.86     |
| eggNOG   | 42,229| 57.72     |
| NR       | 42,374| 57.92     |

**Fig. 4** The significantly enriched KEGG pathway of Unigenes in *Syringa oblata* Lindl.
Table 4  Types of differences in secondary metabolites in the phenylpropanoid pathway

| Correlation | ID         | Metabolite                                | Ratio  | q-value | VIP   | Regulated |
|-------------|------------|-------------------------------------------|--------|---------|-------|-----------|
| neg         | M609T415   | Rutin                                     | 4.49   | 0.006   | 1.46  | up        |
| neg         | M285T182   | Kaempferol                                | 3.41   | 0.006   | 1.39  | up        |
| neg         | M223T280   | Tras-2-Hydroxycinnamic acid               | 0.45   | 0.006   | 1.11  | down      |
| neg         | M79T280    | Naringin                                  | 0.47   | 0.006   | 1.06  | down      |
| pos         | M317T170   | Quercetin 3′-methyl ether                 | 3.20   | 0.007   | 1.25  | up        |
| pos         | M449T177   | Astragalin                                | 3.09   | 0.021   | 1.29  | up        |
| pos         | M287T187   | Luteolin; Kaempferol                      | 3.11   | 0.007   | 1.39  | up        |
| pos         | M381T271   | Daidzin                                   | 0.45   | 0.006   | 1.09  | down      |
| pos         | M268T316   | L-Phenylalanine, N-acetyl-                | 0.32   | 0.007   | 1.40  | down      |
| pos         | M224T38    | 3,4-Dihydroxyhydrocinnamic acid           | 0.40   | 0.007   | 1.21  | down      |
| pos         | M447T297   | Glycitin                                  | 0.34   | 0.021   | 1.31  | down      |

Verification of differentially RNA-Seq expressed Unigenes by quantitative real-time PCR.

To confirm the accuracy of the RNA-Seq sequencing analysis, 13 putative genes were conducted to analyze their relative expression level between Z₀ and Z₁ (Fig. 7) by quantitative real-time PCR analysis. The 13 putative genes (PAL, 4CL1, CYP73A, two HST, two CHS, CHS2, FHT, DLO2, DMR6, SGR1, and CYP75B1) expressed from the biosynthesis of rutin were up-regulated and a similar phenomenon was observed with RNA sequencing data, meaning that the transcriptome analysis was reliable. Among them, the putative gene expression of CYP75B1 (TRINITY_DN31559_c2_g5), CYP73A (TRINITY_DN29851_c0_g2), HST (TRINITY_DN31867_c0_g2), HST (TRINITY_DN31867_c0_g1) and 4CL1 (TRINITY_DN35155_c0_g1) were highest in Z₁ and up-regulated to 19.31, 19.16, 11.70, 11.23 and 8.84, respectively. This confirms that RNA-Seq sequencing analysis helps to understand the regulatory mechanisms of flavonoid biosynthesis.

![Flavonoid Biosynthesis](https://example.com/flavonoid_biosynthesis.png)

Fig. 5  The putative genes involved in flavonoid biosynthesis in *Syringa oblata* Lindl.


Discussion

*S. xylosus*, one of the most common pathogens in the skin of mammals [33], is frequently isolated from milk, meat, and other food products such as cheeses and sausages [34], it is the leading cause of cow mastitis infection [35]. *S. xylosus* is also one of the coagulase-negative Staphylococci [36], which has strong ability to form biofilm [34, 37]. Therefore, in the quest of eradicating cow mastitis from dairy herds, it is necessary to look for a medicine which can resist the formation of biofilms in bacteria.

Currently, many medicinal plants can inhibit the formation of biofilms, especially among the well-known experiments. Mediterranean herbal extracts have been reported to inhibit the formation of biofilm by *Streptococcus mutans* [38] and Glycyrrhiza physic liquor, Glabrous crazyweed liquor [39] and Glabrous Crazyweed [40] have also been noted for their activities against the formation of biofilm by *Staphylococcus epidermidis*. This study revealed that 1/2 MIC of *S. oblata* extract in Z1 had a significant ability to inhibit *S. xylosus* biofilm formation (*p* < 0.05). At the same time, previous study [41] has confirmed rutin as the main bioactive ingredient in *S. oblata* which inhibits biofilm formation in *S. suis* [15]. This is consistent with the report of this study that rutin is the main bioactive ingredient of *S. oblata* extract which inhibits *S. xylosus* biofilm formation. Therefore, it is significant to improve the content of rutin in *S. oblata* extract against *S. xylosus* biofilm formation.

Current data on PM 2.5 collected from 20 European sites may be pertinent in epidemiologic studies [42]. Visibility degradation is related to the PM components [43]. Light intensity, which is one of the environmental factors in the lower atmosphere, maybe also have effect on the growth of plants, especially in the synthesis of active ingredient and the concentration of their contents [44]. This study revealed that the rutin content of *S. oblata* had a significant increase in August compared to other months (May to October). This finding is contrary to the trend of air pollution index (Additional file 1: Table S1) and earlier findings previously reported on the content of rutin in Fructus Sophorae which was observed to peak in mid-August, and then decline gradually [26]. However, considering the natural growth of *S. oblata* and the role of beautifying the city, it is not recommended to select the month of August as the best harvesting period for *S. oblata*. At the same time, compared with *S. oblata* extract in August, there was no significant difference in the biofilm formation of *S. xylosus*. Therefore, it is recommended that the harvesting period of *S. oblata* should be selected before the emergence of the dry leaves in mid-September, which is consistent with previous studies [26]. In addition, previous study has found that anthocyanin and flavonoids accumulation is strongly associated with different flowering developmental stages in *S. oblata* [45, 46]. With regard to photoperiod, it was observed that flavonol compounds and the expression of flavonoid pathway genes were related to the increased light exposure imposed on sweet potato leaves [47]. In summary, this is consistent with this study that the amount of rutin content was related to the plant photoperiod or developmental stage. Shading treatments significantly affected flavonoid accumulation in plants, such as in tea plants [48]. However, the flavonoid levels in an excellent albino tea germplasm

![Pearson partial correlation between differential metabolites and expressed Unigenes involved in the biosynthesis of flavonoid in Syringa oblata Lindl: left: negative correlation; right: positive correlation](image-url)
Fig. 7  (a) PAL (TRINITY_DN36967_c0_g1), (b) 4CL1 (TRINITY_DN35155_c0_g1), (c) CYP73A (TRINITY_DN31867_c0_g1), (d) HST (TRINITY_DN31867_c0_g1), (e) HST (TRINITY_DN31867_c0_g2), (f) CHS2 (TRINITY_DN35859_c0_g3), (g) CHS (TRINITY_DN35859_c0_g2), (h) CHS (TRINITY_DN32109_c0_g3), (i) FHT (TRINITY_DN36563_c0_g1), (j) DLO2 (TRINITY_DN36038_c0_g2), (k) DMR6 (TRINITY_DN28124_c3_g2), (l) SGR1 (TRINITY_DN28900_c0_g2), (m) CYP75B1 (TRINITY_DN31579_c2_g5), and (n) bHLH (TRINITY_DN28965_c1_g8)
increased after moderate shading treatment [49] and all fruits do not require strong light exposure to accumulate high amounts of flavonoids [50]. And the shaded trees were significantly lower than trees without shades (exposed to sun) to enhance light capture and use efficiency in low-light environments [51], but would not greatly alter spectral quality [52]. This is consistent with this in low-light environments [51], but would not greatly posed to sun) to enhance light capture and use efficiency were significantly lower than trees without shades (ex-

High amounts of flavonoids [50]. And the shaded trees increased after moderate shading treatment [49] and all the results of the PAR of Z 1 and Z 2 were significantly lower compared with Z 0.

This indicated that the selection of S. oblata in Z 1 is of great significance for subsequent molecular mechanism research because of its excellent ability to inhibit biofilm formation and the high rutin content compared with Z 0.

Furthermore, in order to locate the subcellular site of flavonoids accumulation in the tissues, histochemical analysis was conducted. Previous studies have shown that the key enzyme in flavonoids formation was found in the epidermal [53] and subepidermal mesophyll tissue that could absorb potentially harmful UV-B radiation [28]. The results showed that more accumulation of flavonoids was observed in Z 1 in the epidermal cell layers and vascular bundles compared with Z 0. This finding is in agreement with earlier findings previously reported [28]. Generally, the study on the accumulation of flavonoids in S. oblata under Z 0 and Z 1 was used to investigate the molecular mechanism that was beneficial to flavonoid production.

Up to date, metabolome and transcriptomic data of S. oblata are still not available in NCBI database, which is widely used in identifying novel genes that are involved in the biosynthesis of secondary metabolites. Thus, a metabolite and RNA-sequencing analysis without a reference genome was used to elucidate the differential regulation involved in the different light intensity groups for S. oblata. In the current study, the results indicated that the differential metabolites and DEGs which were annotated and classified were mainly related to the flavonoid biosynthesis pathway. It has been reported that the components involved in flavonoid biosynthesis are the main functional components in many species. For example, the flavonoid biosynthesis related genes in Tricyrtis sp were isolated and characterized and they consist of CHI, F3H, F3′H, FLS, DFR, ANS. These genes vary with the flower developmental stages [45]. 16 genes were different in the regulation of flavonoid biosynthesis in Camellia sinensis under different shading stages. It was observed that F3′H and FLS significantly decreased throughout the shading stages while the others (PAL, CHS, DFR, ANS, ANR and LAR, etc.) temporally decreased in the early or late shading stages [48]. Interestingly, compared with Z 0, this study showed that 13 putative structural genes related to the flavonoid pathway were up-regulated in Z 1, indicating that these putative genes may be the key target genes regulating flavonoid biosynthesis and metabolism.

Furthermore, 3 putative genes including 4CL1 (TRINITY_DN35155_c0_g1), CYP73A (TRINITY_DN29851_c0_g2) and CYP75B1 (TRINITY_DN31579_c2_g5) were positively correlated with rutin by the integration of metabolites and DEGs analysis and up-regulated to 19.31, 19.16, 11.70, 11.23 and 8.84 respectively by quantitative real-time PCR analysis. This suggests that our results provide the first accurate and relevant gene information for S. oblata in the flavonoid biosynthetic pathway.

4CL1 is the first main branch point enzyme that controls the metabolism of rutin through the phenylpropa- noid metabolic pathway. 4CL enzymes catalyzes the conversion of several hydroxycinnamic acids into their corresponding thioesters, conserving the evolution of vascular plants [54]. In addition, the 4CL enzymes have been characterized from various plants for their role in plant physiology or in biotic and abiotic stresses [55]. In this study, 4CL1 (TRINITY_DN35155_c0_g1) was positively correlated with Z 1 which helped to promote most of the enzymes including CYP73A (TRINITY_DN29851_c0_g2) and CYP75B1 (TRINITY_DN31579_c2_g5). In addition, de novo transcriptome sequencing was performed to reveal that genes such as curcumin synthase and CYP73A were the differentially expressed genes in the different species of gingers being investigated [56]. At the same time, expression of chalcone reductase, flavonoid 3′,5′-hydroxylase (F3′5′H) and CYP75B1 were all reported to be up-regulated under cold treatment in C. microphyllum, which was as a potential source of abiotic stress resistant germplasm for chickpea breeding programs [57]. This study has confirmed that the presence of CYP73A (TRINITY_DN29851_c0_g2) and CYP75B1 (TRINITY_DN31579_c2_g5) enhanced promote the production of rutin.

In addition, studies have shown that transcriptional regulation such as bHLH is the most important tool for modulating flavonoid biosynthesis when plants are under stress conditions [58]. In plants, bHLH transcription factors have a wide range of functions, such as regulation of marginal pollen tract tissues growth [59], morphogenesis [60] plant growth, development and defence [61] and flavonoid biosynthesis [62]. The study
observed that three bHLH (bHLH79, bHLH147, and bHLH48-like) transcription factors were up-regulated at the budding and flowering stages, and another two transcription factors (bHLH3 and bHLH48) were down-regulated at the budding and flowering stages in S. oblata [46]. The over-expression of bHLH1 gene from grape was chemically synthesized and significantly increased the accumulation of flavonoids and enhanced salt and drought tolerance in transgenic Arabidopsis thaliana plants [63]. This experiment showed that under light intensity groups, the bHLH transcription factor was up-regulated to 10.48. At the same time, it was observed to be the main factor that regulated the flavonoid biosynthetic process, seed coat development, protein dimerization activity, seed development, regulation of proanthocyanidin biosynthetic process and so on.

**Conclusions**

The synthesis of the active ingredient content of S. oblata will be affected under different light intensity groups. Therefore, it was speculated that a smoggy environment would affect the synthesis of the secondary metabolic components in medicinal plants. Based on our work view, we successfully screened S. oblata under Z1 and it was observed to have the best rutin and flavonoids content and a better efficacy in inhibiting S. xylosus biofilm formation. The result of metabolite and transcriptome analysis revealed significant metabolite-genes that correlated with the flavonoids biosynthesis pathways such as 4CL1 (TRINITY_DN35155_c0_g1), CYP73A (TRINITY_DN29851_c0_g2) and CYP75B1 (TRINITY_DN31579_c2_g5), and therefore present a better understanding of the molecular mechanism behind the accumulation of flavonoids by S. oblata in response to different light intensity.

**Materials and methods**

**Plant growth and materials collection**

S. oblata, identified by Professor Xiuju Wu (College of Life Sciences, Northeast Agricultural University, Harbin, China) were grown under natural sunlight environment in the campus of Northeast Agricultural University (N 45°44’33.64″, E 126°43’22.07″) in Harbin, Heilongjiang Province of China. They were grown under natural environmental conditions without additionally watered/fertilized. The experiment was divided into two groups of different months and different light intensity groups. The first batch of experimental samples were collected from the S. oblata grouped under different months and sample collection started on the 11th of May and ended on 11th of October, 2017. The other experiment was conducted by using different black coloured nets to cover the S. oblata and thus different light intensity treatments were created as follows: S. oblata under natural growth condition (Z0), S. oblata with 40% ~ 50% shading treatment (Z1, 50% ~ 60% of natural sunlight can be transmitted through the shade nets) and S. oblata with 65% ~ 75% shading treatment (Z2, 25% ~ 35% of natural sunlight can be transmitted) (Fig. 1a and Fig. 2a). The nets were placed over the plants on 11th of May in 2017 (Fig. 2a). S. oblata leaves were collected from all the different light intensity treatments on 11th of September in 2017. At the same time, environmental parameters were measured among each treatment to monitor the growth conditions of S. oblata during light intensity experiment, including the PAR (Light Scout® Quantum Light Meters, Item#3415F, Spectrum Technology Inc. USA) and light reflection by spectrometer (UVCHR768, SVC, America). Above all, in each of the experiment, five trees were selected from each treatment group, and a large number of leaves were collected from each tree randomly [51]. All samples were collected on the 11th of each month, frozen in liquid nitrogen and stored at ~ 80 °C until they were analyzed.

**Biofilm formation ability of S. xylosus**

S. xylosus ATCC 700404 strain was cultured in Trypticase Soy Broth (TSB: Summus Ltd., Harbin, Heilongjiang, China) at 37 °C for 12 h with constant shaking. The methanolic extracts of S. oblata and standard rutin were used for MIC assays using the protocol described previously [64]. Briefly, the overnight cultures of S. xylosus were diluted to a density of 1 × 10^5 CFU/mL using sterile TSB, then 100 μL samples were added to each well in a 96-well plate (Corning Costar® 3599 Corning, NY, USA) containing serial dilutions of compounds in 100 μL culture medium. Control bacteria were cultivated in the absence of extracts of S. oblata. The MICs were determined as the lowest concentration of extracts of S. oblata after incubation for 24 h at 37 °C. The biofilm formation assay was carried out by the extracts of S. oblata using 96-well microtiter plates [64]. Negative control wells contained broth only. Positive control wells contained culture medium and bacterial suspension. Biofilms were treated as described above [64]. The OD of sample was measured at 595 nm using a micro-titer plate reader (DG5033A, Huadong Ltd., Nanjing, Jiangsu, China). Three independent sample analysis were performed for each tissue.

**Determination of rutin content**

High performance liquid chromatography (HPLC) analysis was performed on methanolic extracts of S. oblata using Waters Alliance HPLC system (Shimadzu, Corporation, Kyoto, Japan) that is equipped with a binary pump and a UV/V detector. The fresh tissues obtained from the different months and light intensity groups which were frozen in liquid nitrogen were dried under room temperature. A total of 20 to 50 mg for each dry sample.
was used to extract flavonoids by adding 2 mL of 50% (v/v) of methanol (HPLC grade) in H$_2$O. Then, the mixture was placed in an ultrasonic cleanser (Ningbo Scientz Biotechnology Co. Ltd) for 20 min and centrifuged for 10 min at 13000 rpm [9]. The supernatant was filtered through a 0.45 mm membrane filter and loaded for HPLC analysis. The chromatographic separation was carried out on a Diamosil C18 column (4.6 mm × 250 mm, 5 μm) with a gradient solvent A (0.1% formic acid aqueous solution) and solvent B (acetonitrile) as mobile phase at a flow rate of 1 mL/min. The gradient conditions was 0 min, 5% solvent B; 30 min, 53% solvent B; 35 min, 5% solvent B [15]. Rutin quantity was estimated based on the linear calibration curve of standard rutin (Sigma-Aldrich, Germany) under a detection wavelength of 355 nm. Three independent sample analysis were performed for each tissue.

Histochemical analysis of flavonoids with different light intensity groups

Small pieces of fresh *S. oblata* leaves (Z$_0$ and Z$_1$) were embedded in medium before cutting for histolocalisation as described above [65, 66]. The 20-μm embedded tissue were obtained using a LEICA CM 1850P vibrating blade microtome. These sections were labeled with saturated (0.25%, w/v) 2-aminoethyl diphenylborinate (DPBA) (Macklin, Shanghai Macklin Biochemical Co.,Ltd.) that dissolved in 80% methanol for 15 min. Then, 80% methanol was used to wash away the excess DPBA dye and xylene was used to make the tissues transparent. The DPBA-labelled sections were viewed by laser confocal scanning microscopy (LCSM, Germany).

Metabolites extraction and LC-MS analysis with different light intensity groups

**Metabolites Extraction**

Frozen samples from Z$_0$ and Z$_1$ were ground into fine powder in liquid nitrogen and then stored at −80°C for RNA extraction. Total RNA was extracted using Trizol reagent (Invitrogen, CA, USA) by following the manufacturer’s procedure. The total RNA quantity and purity were analyzed by Bioanalyzer 2100. Two biological repeats were performed.

**RNA library construction and transcriptomic analysis**

Approximately 10 μg of total RNA were prepared. The methods adopted for library preparation, de novo strategy and transcriptomic analysis were the same with our previous study [72]. Sequencing was carried out using an Illumina Hiseq 4000 platform (LC-Bio, Hangzhou, China) according to the manufacturer’s protocol. De novo assembly of the transcriptome was performed with Trinity 2.4.0 [73]. For gene identification and expression analysis, the reads from different species were co-assembled, and for gene sequence analysis, the reads from different species were assembled separately.
Unigenes annotation and functional classification
All assembled Unigenes were aligned against the non-redundant (Nr) protein database (http://www.ncbi.nlm.nih.gov/), Gene ontology (GO) (http://www.geneontology.org), SwissProt (http://www.expasy.ch/sprot), KEGG (http://www.genome.jp/kegg/) and eggNOG (http://egg nogdb.embl.de/) databases using DIAMOND [32] with a threshold of E-value < 0.00001.

Differentially expressed Unigenes analysis
Salmon [74] was used to calculate the expression level of Unigenes (TPM) [75]. The differentially expressed Unigenes (DEGs) were selected with log2 (fold change) > 1 or log2 (fold change) < -1 and with statistical significance (p value < 0.05) by R package edgeR [76]. Next, GO and KEGG enrichment analysis were again performed on DEGs by perl scripts in-house.

Integration of metabolites and expressed Unigenes with different light intensity groups
Functional analysis of integration of metabolomics and transcriptomics data
Correlation between the expression levels of 13 putative genes and the profiles of flavonoids in *S. oblata* from Z0 and Z1 were carried out using the program R 2.10.1. RPMK values for genes and the peak values of metabolites were used as a matrix for pearson partial correlation analysis [77]. The metabolite correlation network was constructed for Z0 and Z1 using all metabolite accumulation profiles separately. Correlation pairs were deemed statistically significant when the [PCC] > 0.9 and p-value < 0.01. The resulting correlation networks were obtained and used for network visualization and analysis of network properties using Cytoscape software (Cytoscape 2.6.3) [78].

Verification of differentially expressed Unigenes in flavonoid biosynthesis pathway by quantitative real-time PCR
The total RNA for quantitative real-time PCR analysis was extracted using TRIZOL reagent and 1.0 μg RNA was used for reverse transcription using the PrimeScript™ RT reagent Kit with gDNA Eraser (Tiangen, Beijing, China) in 20 μL of reaming system. 13 putative structural genes and 1 putative regulatory gene were selected as DEGs. The CPR gene were selected as internal control. The specific primers were designed from Sangon Biotech (Shanghai) and listed in Additional file 1: Table S2. Quantitative real-time PCR was performed using the Roche Light Cycle 480 II sequence detection system (Roche, Switzerland) as previously described, with a few modifications [79]. The final volume for each reaction was 10 μL with the following components: 1 μL diluted cDNA template (1 mg/mL), 5 μL SYBR Green Master (ROX) (Indianapolis, IN, USA), 0.3 μL forward primer, 0.3 μL reverse primer and 3.4 μL ddH2O. The reaction was set at 95 °C for 10 min, followed by 40 cycles of denaturation at 95 °C for 15 s and annealing/extension at 60 °C for 1 min. Each quantitative real-time PCR analysis was performed with three biological replicates.

Student’s t-test
Values were expressed as means ± SDs. The statistical differences among different groups were compared by 1-way ANOVA. Significant means were separated using Tukey method and statistical significant level was set as p < 0.05. The data of quantitative real-time PCR were analyzed using repeated measurements in –ΔCt model [80].

Supplementary information
Supplementary information accompanies this paper at https://doi.org/10.1186/s12870-019-2100-8.

Additional file 1: Figure S1. Genes with significant differential expression in *Syringa oblata* Lindl. Figure S2. Volcano diagram of differentially expressed Unigenes between two groups in *Syringa oblata* Lindl. Figure S3. GO enrichment analysis of differentially expressed Unigenes in *Syringa oblata* Lindl. Table S1. Air pollution index for May to October in 2017. Table S2 Fluorescence quantitative real-time PCR primer sequence.

Abbreviations
4CL: 4-coumarate-CoA ligase; CHS: chalcone synthase; CYP73A: trans-cinnamate 4-monoxygenase; CYP73B1: flavonoid 3′-monoxygenase; DEGs: Differentially Expressed Genes; DLO2, DMR6 and SGR1: flavonol synthase; DPBA: 2-aminoethyl diphenylborinate; F3: flavonoid 3′′- hydroxylase; FHT: naringenin 3′-dioxygenase; GO: Gene ontology; HPLC: High performance liquid chromatography; HST: shikimate O-hydroxycinnamoyltransferase; KEGG: Kyoto Encyclopedia of Genes and Genomes; MIC: minimal inhibitory concentration; PAL: phenylalanine ammonia-lyase; PCA: principal component analysis; PLS-DA: partial least squares-discriminate analysis; PM: particulate matter; S. oblata: *Syringa oblata* Lindl.; S. suis: Streptococcus suis; S. xylosus: Staphylococcus xylosus

Author contributions
Y-HL designed the research. Y-YL and X-RC conducted the fieldwork, performed the experiments and analyzed the data. J-PW, X-XX, X-YC, W-YD, W-QC, GB, and NE were supportive during the experiment. Y-HL, Y-YL and X-RC wrote the manuscript. All authors have read and approved the manuscript for publication.

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Availability of data and materials
All raw sequence reads have been deposited in NCBI’s Gene Expression Omnibus and are accessible through GEO Series accession number GSE137862 (https://www.ncbi.nlm.nih.gov/geo/query/acc.cgi?acc=GSE137862).

Ethics approval and consent to participate
Not applicable.

Consent for publication
Not applicable.

Competing interests
The authors declare that they have no competing interests.
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