# Meta-analysis of transcriptomes in insects showing density-dependent polyphenism

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Simple Summary: Population density can be an environmental cue to induce modification of insect 14 morphology, physiology, and behavior. This phenomenon is called density-dependent plasticity. 15 Aphids and locusts exhibit textbook examples of density-dependent plasticity but there is a lack of 16 integrative understanding for insect density-dependent plasticity. To address this problem, we com-17 bined public gene expression data obtained from multiple studies and re-analyzed them (this pro-18 cess is called meta-analysis). The present study provides additional insight into the regulatory 19 mechanisms of density-dependent plasticity, demonstrating the effectiveness of meta-analyses of 20 public transcriptomes. 21

Abstract: With increasing public data, a statistical analysis approach called meta-analysis, which 22 combines transcriptome results obtained from multiple studies, has succeeded in providing novel 23 insights into targeted biological processes. Locusts and aphids are representative of insect groups 24 that exhibit density-dependent plasticity. Although the physiological mechanisms underlying den-25 sity-dependent polyphenism have been identified in aphids and locusts, the underlying molecular 26 mechanisms remain largely unknown. In this study, we performed a meta-analysis of public tran-27 scriptomes to gain additional insights into the molecular underpinning of density-dependent plas-28 ticity. We collected RNA sequencing data of aphids and locusts from public databases and detected 29 differentially expressed genes (DEGs) between crowded and isolated conditions. Gene set enrich-30 ment analysis was performed to reveal the characteristics of the DEGs. DNA replication 31 (GO:0006260), DNA metabolic processes (GO:0006259), and mitotic cell cycle (GO:0000278) were 32 enriched in response to crowded conditions. To date, these processes have scarcely been the focus 33 of research. The importance of the oxidative stress response and neurological system modifications 34 under isolated conditions has been highlighted. These biological processes, clarified by meta-anal-35 ysis, are thought to play key roles in the regulation of density-dependent plasticity. 36

Keywords: Meta-analysis; density-dependent polyphenism; aphid; locust; RNA sequencing; gene37set enrichment analysis38

#### 1. Introduction

Phenotypic plasticity is the ability to respond adaptively to environmental variations 41 by producing more than one phenotype for a single genotype [1,2]. If two or more distinct 42 phenotypes are produced, this is called polyphenism [3]. Various environmental conditions can be cues to induce plastic phenotype changes, population density being one of 44

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those [3–6]. Locusts and aphids are representative insect groups for which density-de-45 pendent plasticity is considered to have contributed to their evolutionary success [7,8]. 46 The aphids exhibit wing polyphenotypes and locusts exhibit polyphenisms in body color 47 and behavior (e.g., gregarious and flight ability). 48

Physiological mechanisms underlying density-dependent polyphenism have been 49 revealed in aphids [7] and locusts [9]. Furthermore, a recent transcriptome analysis has 50 revealed gene expression profiles in response to different density conditions. RNA se-51 quencing (RNA-seq) revealed that a laterally transferred viral gene is responsible for 52 switching wing polyphonic phenotypes [10]. In locusts, expressed sequence tag (EST), mi-53 croarray, and RNA-seq revealed differentially expressed genes (DEGs), including juvenile 54 hormone (JH)-binding protein [11], heat-shock proteins [12], oxidative stress response 55 genes [12], neurotransmitter-related genes [13], catecholamine metabolic pathway genes 56 [14], and the microtubule cytoskeleton [15]. However, the common molecules involved in 57 all locust species exhibiting density-dependent polyphenism remain largely unknown. As 58 mentioned above, RNA sequencing has revealed the species-specific regulatory mecha-59 nisms in each species, and the common genes involved in the regulation of density-de-60 pendent polyphenism are predicted to be present. However, an inclusive analysis of 61 aphids and locusts was not performed to determine the common genes expressed in all or 62 multiple species that show density-dependent polyphenism. 63

Meta-analysis that combines the transcriptome results from multiple studies is 64 thought to be effective in providing additional insights into density-dependent polyphen-65 isms because meta-analysis can uncover new information that could not be achieved with 66 conventional hypothesis-driven research methods. Meta-analysis of transcriptomes is ef-67 fective in revealing novel gene expression profiles [16-20]. As mentioned above, the 68 amount of transcriptome data related to density-dependent polyphenism is sufficient to 69 execute the meta-analysis. In the present study, we conducted a meta-analysis of public 70 transcriptomes using RNA-seq data obtained from five Schistocerca species, the pea aphid 71 Acyrthosiphon pisum, and the migratory locust Locusta migratoria. We identified DEGs be-72 tween crowded and isolated conditions and performed gene set enrichment analysis to 73 understand the characteristics of the DEGs. 74

#### 2. Materials and Methods

#### 2.1. Curation of public gene expression data

Expression data were searched for in the public databases Gene Expression Omnibus (GEO) [21] and DBCLS SRA (https://sra.dbcls.jp/), and RNA-seq data archived in the Se-78 quence Read Archive (SRA) [22] using the keywords including crowding, density, poly-79 phenism, and gregarious. The datasets examined in this study were selected based on the 80 following criteria: RNA-seq data obtained from crowded or isolated insect species show-81 ing density-dependent polyphenism. 82

#### 2.2. RNA-seq data retrieval, processing, and quantification

SRA retrieval and conversion to FASTQ files was performed using the prefetch and 84 fasterq-dump programs in the SRA Toolkit (v2.9.6) [23], respectively. To decrease disk 85 space usage, the obtained FASTQ files were immediately compressed using pigz (v. 2.4). 86 For the trimming and quality control, Trim Galore! (v.0.6.6) [24] and Cutadapt (v.3.4) [25] 87 software was used. Trim Galore! [24] was run with the parameters -fastqc --trim1--paired 88 (if data were paired-end). Trim Galore! [24] includes Cutadapt (v.3.4) [25] to trim low-89 quality base calls. Since all data were passed through this quality check, low-quality data 90 were not included in this study. 91

#### 2.3. Quantification of gene expression level

In species whose genomic information is available (the pea aphid Acyrthosiphon pisum 93 and the migratory locust Locusta migratoria), RNA-seq reads were mapped to reference 94

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genome sequences using HISAT2 (v.2.2.1) [26] with parameters -q --dta -x (*Acyrthosiphon* 95 *pisum*: GCF\_005508785.1 or *L. migratoria*: GCA\_000516895.1). Quantification of the 96 mapped data was performed using StringTie (2.1.5) [27], and the values of transcripts per 97 million (TPM) were used as quantitative values of gene expression. 98

In species with genomic information not available (all species of grasshopper 99 Schistocerca), TPM values were calculated using Salmon (v.1.5.0) [28] with parameters -i 100 index -l IU. Transcriptome assemblies required for Salmon were retrieved from the Tran-101 scriptome Shotgun Assembly (TSA) database (Table S1). If there were no available assem-102 blies in TSA, de novo transcriptome assembly was performed using Trinity (v2.13.1) pro-103 gram with Docker environment utilizing scripts in the Systematic Analysis for Quantifi-104 cation of Everything (SAQE) repository [16]. Trinity was performed with the default pa-105 rameters. 106

#### 2.4. The detection of differentially expressed genes

To evaluate the changes in expression of each gene, the expression ratio was calculated using TPM paired with crowded and isolated transcriptomes (CI ratio). The CI ratio for each gene was calculated using Equation (1): 110

$$CI ratio = \log_2(TPM_{crowded} + 0.01) - \log_2(TPM_{isolated} + 0.01),$$
(1)

This calculation was performed for all pairs of crowded and isolated transcriptomes. 111 Pairs of transcriptome data used for comparison are shown in Table S2. The value "0.01" 112 was added to the TPM for each gene to convert zero into a logarithm. Crowded and iso-113 lated ratios (CI-ratio) were classified as "upregulated," "downregulated," or "unchanged" 114 according to the thresholds. Genes with a CI ratio greater than 1 (2-fold expression 115 change) were treated as "upregulated". Genes whose CI ratio was under -1 (0.5-fold ex-116 pression changes) were treated as "downregulated". Others were treated as "unchanged". 117 To reveal the DEGs between the crowded and isolated conditions, the crowded and iso-118 lated score (CI score) of each gene was calculated by subtracting the number of sample 119 pairs with "downregulated" from those of "upregulated," as shown in Equation (2): 120

$$CI \ score = count \ number_{upregulated} - count \ number_{downregulated},$$
(2)

The calculation method for CI ratios and CI scores was based on a previous study121[17] (https://github.com/no85j/hypoxia\_code/blob/master/CodingGene/HN-score.ipynb).122Visualization of CI ratio was performed with heatmap2 (R ver 4.2.0).123

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#### 2.5. Gene annotation and gene set enrichment analysis

In species whose genomic information is available, GFFread (v0.12.1) [29] was used 126 with the reference genome (Acyrthosiphon pisum: GCF\_005508785.1 and L. migratoria: 127 GCA\_000516895.1) to extract the transcript sequence from the genome data. In species 128 whose genomic information is not available (all species of the Schistocerca genus), tran-129 script sequences were collected from the TSA database (Table S1). However, transcript 130 sequences of Schistocerca gregaria were obtained by de novo assembly using Trinity, as de-131 scribed above (2.3. Quantification of gene expression level). TransDecoder (v5.5.0) 132 (https://transdecoder.github.io/: Accessed on 22 June2021) was used to identify coding re-133 gions and translate them into amino acid sequences of all species. Open reading frames 134 were extracted by TransDecoder.Longorfs with parameters -m 100, and coding regions 135 were predicted by TransDecoder.Predict with default parameters. Then, the protein 136 BLAST (BLASTP) program with parameters -evalue 0.1 -outfmt 6 -max\_target\_seqs 1 in 137 the NCBI BLAST software package (v2.6.0) was used to determine the orthologous gene 138 relationship among species. Ensembl Biomart was used to obtain stable protein IDs and 139 gene names from *D. melanogaster* gene set (BDGP6.32). 140

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Gene set enrichment analysis was performed by Metascape [30] using the DEGs obtained in section 2.4. 141

#### 3. Results

#### *3.1. Data collection of transcriptomes in density-dependent polyphenism*

To compare gene expression under crowded and isolated conditions, we retrieved 145 transcriptome data acquired under different density conditions. Consequently, 66 pairs of 146 transcriptome data were collected from the public databases (Table 1). Full information 147 about the dataset used in this study is provided in Table S2 [10, 31, 32, 33, 34, 35]. These 148 data comprised transcriptomes from seven insect species, including aphids and locusts. 149 The tissues used to obtain these transcriptomes are shown in Table 1. 150

Table 1. Summary of datasets used in this study.

Species	Number of RNA sequencing files (under crowded condi- tion)	Number of RNA sequencing files (under isolated condi- tion)	Tissue
Schistocerca gregaria	1	1	CNS
Acyrthosiphon pisum	8	8	whole
Schistocerca americana	10	10	head, thorax
Schistocerca nitens	10	10	head, thorax
Schistocerca piceifrons	10	10	head, thorax
Schistocerca serialis cubense	10	10	head, thorax
Locusta migratoria	17	17	brain, Integument, tho- racic ganglion

#### 3.2. Gene set enrichment analysis

After the expression ratio (CI ratio) was calculated for each species, as described in 154 the Materials and Methods section 2.4, we integrated the CI ratio table and obtained the 155 CI ratio values of 2652 genes (Table S3: https://doi.org/10.6084/m9.figshare.20174255.v1). 156 During this process, orthologous relationships among species were required to be deter-157 mined using BLAST, because the gene names of the expression ratio are species-specific 158 depending on the annotation file or transcriptome assemblies of each species. First, we 159 performed a BLASTP search using all genes in *S. gregaria* as queries against those in each 160 species examined in this study. Since density-dependent polyphenisms have been exten-161 sively studied using this species, this species was selected as a reference for the determi-162 nation of orthologous relationships among species. Next, we performed searches for all 163 genes vs. all BLASTP between D. melanogaster and S. gregaria to annotate the gene names 164 of each species into those of D. melanogaster. 165

CI scores were calculated based on the CI ratio table (Table S4: 166 https://doi.org/10.6084/m9.figshare.20174255.v1). The previous studies showed that ebony, 167 Mcm2, and Mcm7 were upregulated in response to crowded condition, and Duox, Mdr49 168 and Cyp6a14 were downregulated in response to crowded condition (i. e. upregulated in 169 response to isolated condition) [14, 33, 34]. In this study, CI score of these genes showed 170 high or low values (Figure 1). This means that the expression patterns of these genes were 171 consistent with previous results. Previous transcriptome analyses also showed that 172 GO:0008061 [cuticle binding] and GO:0006979 [oxidative stress] were enriched in crowded 173 and isolated condition, respectively [14, 33, 34]. CI score of genes that was included in 174

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these GO terms (GO:0008061; obst-E, GO:0006979; spz, Glaz, srl) showed high or low values 175 (Figure 1). These results show the suitability of the CI score method to detect the DEGs. 176 We listed the top 1% of genes with the highest and lowest CI scores (Table 2 and Table 3, 177 respectively). ebony (e) showed a high CI score for the upregulated DEGs (Tenth rank in 178 Table 2). This gene, encoding a protein that converts dopamine to N- $\beta$ -alanyl dopamine, 179 is highly expressed in gregarious L. gregaria, leading to the production of the yellow color 180 [14]. In the top 1% of genes with the highest and lowest CI scores, we assessed the expres-181 sion pattern in each species and/or tissue (Table S5: 182 https://doi.org/10.6084/m9.figshare.20174255.v1 and Table S6: 183 https://doi.org/10.6084/m9.figshare.20174255.v1). Many genes with unchanged expres-184 sion levels were observed in A. pisum and L. migratoria. Although this tendency was also 185 observed the heatmap of expression ratio (Figure S1: 186 https://doi.org/10.6084/m9.figshare.20174255.v1), the genes that showed a drastic expres-187 sion change were observed in *L. migratoria* samples (SmD3 and CG34461). 188

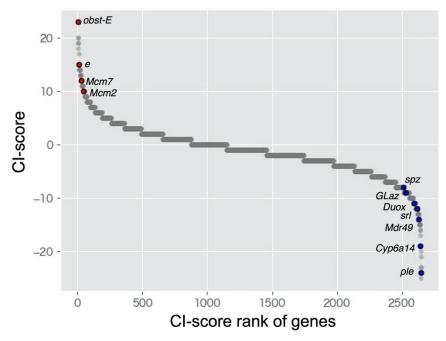


Figure 1. Scatter plots of CI-score of all genes (2652 genes) identified in this study.

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We selected 97 upregulated genes and 199 downregulated genes as DEGs (CI score 193 thresholds:8 and -8). Using these DEGs, we performed gene set enrichment analysis using 194 Metascape to identify their characteristics. Metascape showed that "DNA replication" 195 (GO:0006260; CG14803, CG2990, Psf3, CG8142, RnrL, Psf1, Pol31, Mcm7, Mcm2, E2f1, 196 DNApol-alpha60, dup), "DNA metabolic process" (GO:0006259; CG14803, CG2990, Psf3, 197 CG8142, RnrL, FANCI, pds5, Psf1, Pol31, Gen, Mcm7, Mcm2, pch2, Fancl, CG5285, E2f1, 198 DNApol-alpha60, dup), and "mitotic cell cycle" (GO:0000278; Psf3, sofe, borr, Incenp, pds5, 199 SmD3, tum, Klp61F, Psf1, Pol31, eco, polo, Mcm2, E2f1, jar, Cap-D2, DNApol-alpha60, dup, 200 Zwilch) were enriched significantly in upregulated DEGs (Figure 2A). In addition, we 201 searched for genes that were likely to be associated with density-dependent polyphenism 202 (Table 2). JHEH encodes for juvenile hormone epoxide hydrolase involved in JH inactiva-203 tion, as reported by Iga and Kataoka 2012 [36]. JH plays a central role in the regulation of 204 polyphenism. 205

Table 2. Top 1% of genes with the highest CI scores.

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	crowaed condition				
Gene_name	upregulated	downregulated	unchanged	CI score	
obst-E	31	8	27	23	
CG14803	23	3	40	20	
His2A:CG33835	30	10	26	20	
Jheh2	22	3	41	19	
nw	27	8	31	19	
HDAC1	21	3	42	18	
Ugt36D1	24	7	35	17	
CG5321	22	7	37	15	
Cdep	21	6	39	15	
e	17	2	47	15	
pds5	19	4	43	15	
CG4572	21	7	38	14	
CG6765	16	2	48	14	
Incenp	23	9	34	14	
Psf3	21	7	38	14	
ft	21	7	38	14	
polo	19	5	42	14	
tou	22	8	36	14	
tum	19	5	42	14	
CG10175	22	9	35	13	
CG2990	20	7	39	13	
CG8173	18	5	43	13	
DNApol-alpha60	22	9	35	13	
SmD3	18	5	43	13	
Ts	20	7	39	13	
CG8646	20	8	38	12	

# Number of samples with expression patterns responding to the crowded condition

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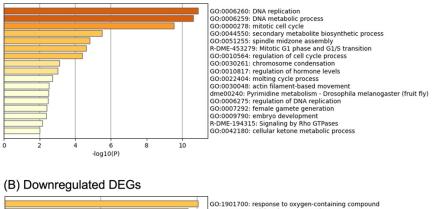
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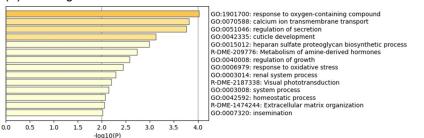
	crowded condition				
Gene_name	upregulated	downregulated	unchanged	CI score	
CG14301	3	28	35	-25	
ple	7	31	28	-24	
CG9657	6	29	31	-23	
Lrp4	3	26	37	-23	
GluClalpha	5	26	35	-21	
CG34461	9	29	28	-20	
Cyp6a14	9	28	29	-19	
CG10211	10	27	29	-17	
Ggt-1	8	24	34	-16	
Syt4	6	22	38	-16	
Cralbp	9	24	33	-15	
KaiR1D	3	18	45	-15	
RpS5a	3	18	45	-15	
Shmt	6	21	39	-15	
Tcs3	4	19	43	-15	
tyn	2	17	47	-15	
CG13744	7	22	37	-15	
CG15449	5	19	42	-14	
Est-6	9	23	34	-14	
KaiR1D	11	25	30	-14	
Mdr49	7	21	38	-14	
dpr12	6	20	40	-14	
CG13643	7	20	39	-13	
CG6006	5	18	43	-13	
CG7246	2	15	49	-13	
Cnb	5	18	43	-13	

### Number of samples with expression patterns responding to the crowded condition

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#### (A) Upregulated DEGs





**Figure 2.** Results of gene set enrichment analysis in (A) upregulated DEGs and (B) downregulated 212 DEGs. 213

The response to oxygen-containing compounds (GO:1901700; G9a, hfw, Shmt, Clic,214Duox, FASN1, Kr-h1, GLaz, ple, Rab8, srl, Syn, CASK, spz, and dco) was significantly en-215riched in the downregulated DEGs (Figure 2B). Multiple genes involved in the response216to oxidative stress (GO:0006979: Duox, GLaz, ple, srl, and spz) were included in GO: 1901700217

The CI score of the *pale (ple)* gene, encoding tyrosine hydroxylase which is involved 218 in pigmentation of the cuticle and catecholamine biosynthesis [37], was the second lowest 219 (Table 3). This indicates that the expression of *ple* was higher in many isolated transcriptomes than in crowded transcriptomes. *ple* genes are known to be highly expressed under 221 crowded conditions, inducing gregarious color and behavior [14]. 222

Among the downregulated DEGs, the genes that are reported to function in the nervous system were conspicuous (*CG9657*: [38], *Lrp4*: [39], *GluClalpha*: [40], *Syt4*: [41]).

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#### 4. Discussion

As various data acquired under different conditions are registered, manual curations 227 are needed for meta-analysis. In this study, we collected RNA sequencing data (66 pairs) 228 from public databases and compared their expression levels in various insect species. We 229 aimed to identify the common genes that can explain the regulation of the density-de-230 pendent polyphenism of all or multiple species. However, in the gene list with highest or 231 lowest CI\_score (Table 2, 3 and Table S5, S6), almost all genes showed no expression 232 changes in A. pisum. This result may reflect the use of different developmental stages. The 233 aphids used in this study are derived from adults, whereas the other samples are derived 234 from larval stages. However, we believe that our meta-analysis of public RNA sequencing 235 using this vast amount of data has led to important insights into the gene expression pro-236 file underlying density-dependent polyphenism, as discussed below. 237

Gene set enrichment analysis showed that DNA replication, DNA metabolic processes, and the mitotic cell cycle were enriched in response to crowded conditions (Figure 1). DNA replication and cell cycle have rarely been focused on as regulatory mechanisms of density-dependent polyphenism [33]. A recent study [42] showed that the regulation 241

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of DNA replication and the cell cycle is involved in the density-dependent wing poly-242 phenism of the planthopper Nilaparvata lugenes. Together with this study, our results em-243 phasize the importance of DNA replication and the cell cycle as the regulatory mecha-244 nisms of density-dependent polyphenism. Although wing polyphenism was also ob-245 served in A. pisum, it was not observed in locusts. Therefore, the regulation of DNA rep-246 lication and cell cycle is expected to be involved in other developmental processes in lo-247 custs. Although the role of JH in aphid wing polyphenism is controversial, JH content is 248 lower under crowded conditions than under isolated conditions in locusts [9, 43]. There-249 fore, Jheh2 expression in response to crowded conditions may play an important role in 250 JH degradation. 251

The high expression of *ple* gene under isolated conditions was not consistent with a 252 previous study. Although *ple* genes are known to be responsible for gregarious pigmen-253 tation in L. migratoria, the relationship between ple genes and body-color polyphenism is 254 controversial [44]. Several downregulated DEGs (Duox, GLaz, ple, srl, and spz) classified 255 under response to oxygen-containing compounds were related to the response to oxida-256 tive stress, and *ple* was included in that category. A previous microarray study showed 257 that the expression of transcripts encoding proteins against oxidative stress damage 258 (peroxiredoxin, 5-oxoprolinase, microsomal glutathione-S-transferase, and transaldolase) 259 was higher in isolated locusts than in crowded locusts [12]. Rapid accumulation of oxida-260 tive stress inhibits flight sustainability in solitary L. migratoria, leading to variations in 261 flight traits between solitary and gregarious locusts [45]. Therefore, the high ple expression 262 under the isolated conditions in this study may result from the response to oxidative stress. 263

Neurological system modifications may play an important role in inducing density-264 dependent phenotypic changes in S. gregaria [12] and L. migratoria [15]. We found that the 265 expression of several genes functioning in the nervous system was increased under iso-266 lated conditions (CG9657 [38], Lrp4 [39], GluClalpha [40], and Syt4 [41]). CG9657 is an 267 SLC5A transporter expressed in glial cells and is involved in sleep behavior in D. melano-268 gaster [38]. Presynaptic Lrp4 functions to ensure normal synapse numbers in D. melano-269 gaster [39]. GluClalpha plays an important role in the ON/OFF selectivity of the visual sys-270 tems in D. melanogaster [40]. Syts are Ca2+-binding proteins involved in the presynaptic 271 transmitter release [41]. These genes are also related to density-dependent behavioral 272 changes. 273

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#### 5. Conclusions

We identified novel genes related to density-dependent polyphenism in insects using a meta-analysis of public transcriptomes. Reliable and general principles should be derived from meta-analysis because this method integrates the results of a number of studies. Therefore, our results can be applied to other species that exhibit density-dependent polyphenisms. 281

Supplementary Materials: The following supporting information can be downloaded at283www.mdpi.com/xxx/s1, Table S1: IDs of transcriptome assemblies used in this study, Table S2: Transcriptome dataset used in this study.284

The following tables downloaded figshare can be at 286 (https://doi.org/10.6084/m9.figshare.20174255.v1), Table S3: Gene expression ratios when compar-287 ing crowded and isolated conditions (CI ratio), Table S4: Score calculated based on the CI ratio 288 between crowded and isolated conditions, Table S5: Number of samples with expression patterns 289 responding to the crowded condition (Top 1% of genes with the highest CI scores), Table S6: Num-290 ber of samples with expression patterns responding to the crowded condition (Top 1% of genes with 291

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	the lowest CI scores), Figure S1: Expression ratio of Top 1% of genes with the highest and lowest CI scores in each species. Each color box indicates the expression ratio of each sample.	292 293
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