1	Supplementary	<b>Information</b> for	or
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<sup>2</sup> Diogeographic paranels in thermal tolerance and gene expression variation und	2	<b>Biogeographic</b>	parallels in	thermal	tolerance and	l gene ex	pression	variation	unde
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- 3 temperature stress in a widespread bumble bee
- 4
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- 21 **This PDF file includes:**
- 22 Supplementary text
- 23 Fig. S1 to S2
- 24 Tables S1 to S4
- 25 Captions for Datasets S1 through S3
- 26 References for SI citations
- 27

# 28 Other supplementary materials for this manuscript include the following:

- 29 Datasets S1 to S3
- 30 31

### 32 Supplementary Information Text

#### 33 Methods

34 Queen Collection and Colony Foundation.

35 Queen Bombus vosnesenskii bees were collected by net from multiple sites between 36 February and July 2016 (Table S1). To increase chances of successful laboratory colony 37 foundation, we focused on queen bees which had recently emerged from hibernation but 38 had not yet started a colony by collecting queen bees visiting flowers. As is typical of 39 many bumble bees, hibernating queen emergence varies seasonally with latitude and 40 elevation<sup>1</sup>, and it was not possible to collect queens from all regions in the same month. 41 Oueens were collected while foraging or flying using aerial insect nets, assigned a unique 42 identification code, transferred to shipping vials and placed in a cooler for transport to the 43 USDA-ARS Pollinating Insect Research Unit (PIRU) bumble bee rearing facility.

44 At PIRU, queens produced colonies following a protocol based on procedures by 45 Evans et al.<sup>2</sup>, with modifications to maximize *B. vosnesenskii* production. Queens placed 46 in plastic queen initiation boxes (2.25L) (Biobest, Learnington, ON), provided with a 500 47 mg pollen provision (homogenous paste made of honey bee corbicular loads and sugar 48 solution) dipped in honey bee wax to encourage nesting behavior. Each queen was also 49 provided lab-made nectar solution in a plastic reservoir ad libitum. The queen initiation 50 boxes were maintained at 0:24 L:D,  $27^{\circ} \pm 1^{\circ}$  C and 55-60% relative humidity (RH)<sup>3</sup>. 51 Queens were checked daily for signs of nesting behavior, including wax secretion, honey 52 pot construction, or presence of brood. After five workers had eclosed, the small colony 53 was moved to a 7.75 L plastic hive box (Biobest, Learnington, ON). Additional pollen 54 provisions (without a wax coating) were provided as needed. At 20+ workers, colonies 55 were transported to the University of Wyoming for physiological experiments. During the 56 course of nest initiation through the transport to the University of Wyoming, the queens 57 and colonies were checked daily for presence of disease and pests and managed to 58 maximize colony growth. Propagation, mating, and overwintering of queens for second-59 generation laboratory reared colonies was attempted but not possible for this species.

60

61 <u>Physiology Data Analysis.</u>

62 We tested for statistical differences between queen collection locations for thermal 63 maxima and minima in two separate pairs of analyses. All data were analyzed in R 64 v3.5.1<sup>4</sup>. First, we tested for broad patterns based on region of origin used a mixed effect 65 linear model approach, defining colony of origin as a random effect, using *lmer* from 66 lme4 v1.1-18-1<sup>5</sup> and lmerTest v.3.1-1<sup>6</sup>. Next, we also tested for a relationship between 67 either CT<sub>MAX</sub> or CT<sub>MIN</sub> and specific queen collection location using *lmer* and calculated the  $R^2$  of the relationship using *r.squaredGLMM* from MuMIn v1.42.1<sup>7</sup>, against the 68 69 following WorldClim<sup>8</sup> variables: Annual Mean Temperature, Maximum Temperature of 70 the Warmest Month, Minimum Temperature of the Coldest Month, Mean Diurnal Range, 71 and Isothermality.

72

73 <u>Weighted Gene Co-Expression Network Analysis.</u>

74 To study broad patterns of gene expression, data were analyzed using WGCNA

v1.64.1<sup>9</sup> following best practices as outlined in the package tutorials to identify modules

of co-expressed genes that correlated with physiological and environmental variables,

similar to previous studies<sup>10,11</sup>. The data were prepared by quantile normalizing the 77 filtered *logCPM* data used in the MDS and gene-by-gene (edgeR/limma<sup>12,13</sup>) analyses. 78 79 Next, these data were used construct a signed-hybrid network and classify modules using 80 the step-by-step approach using pairwise Pearson correlations and a soft-threshold power 81 of 6, as identified by *pickSoftThreshold*. Next, modules were identified with 82 cutreeDynamic and merged if they demonstrated a correlation >75%. The resultant 83 modules were then compared to environmental and physiological traits of the individual 84 bees using Pearson correlation coefficients. The traits tested were: ColonyID, Latitude, 85 Longitude, Elevation, Region, Treatment, Region by Treatment interaction, Annual Mean 86 Temperature of queen collection location (from WorldClim) and CT<sub>TEMP</sub> (temperature of 87 the individual bee at time of collection with control bees assigned 25°C). The modules 88 with statistically significant correlation ( $P \le 0.05$ , *corPvalueStudent*) to at least one trait 89 variable was retained for further analysis. We then tested for a significant association 90 between the modules and location of origin (using WorldClim Annual Mean 91 Temperature<sup>8</sup> as a proxy) and temperature at time of sacrifice (using  $CT_{TEMP}$ ). We 92 compared four linear mixed models of decreasing complexity (ColonyID as a random 93 variable) using the eigengene value of each member of the module as the response 94 variable. The model with the best fit to the module members was identified by iteratively comparing nested models using likelihood ratio tests<sup>14</sup> as implemented by ImerTest v3.0-95  $1^6$  using a P  $\leq$  0.05 of a Kenward-Rogers Approximation as our cutoff<sup>15</sup>. The best model 96 97 was then compared to the null model to determine goodness of fit. Modules were also 98 evaluated to identify the hub gene (gene with the highest level of connectivity to all other 99 module members; chooseTopHubInEachModule), which gene ontology groups were 100 enriched (following methods described above), and level of overlap with gene-by-gene 101 differential expression results.





105 Fig. S1. Histogram of Benjamini-Hochberg false discovery rate corrected p-values for all

106 genes in all three pairwise treatment contrasts. While the distribution of FDR corrected

107 values are similar in distribution for contrasts with  $CT_{MAX}$  and both control and  $CT_{MIN}$ ,

108 very few genes are differentially expressed between control and  $CT_{MIN}$  treatment bees at 109 both the  $P \le 0.05$  level (left of the dashed line) and the  $P \le 0.1$  level (left of the dash-dot

110 line).



111 112 Fig. S2. Annual Mean Temperature is correlated with the other tested BioClim variables, but not with latitude illustrating how our

113 sampling design decouples space and climate.

Colony	State	Latitude	Longitude	Elevation (m)	Region
160019	CA	35.68882	-121.28838	7	L-CA
160034	CA	36.39965	-118.99074	224.9	L-CA
160035	CA	37.03608	-119.5266	350.5	L-CA
160095	OR	45.696075	-121.33883	70.1	L-OR
160102	OR	45.696075	-121.33883	70.1	L-OR
160131	OR	45.535039	-121.20785	441.4	L-OR
160137	OR	45.535039	-121.20785	441.4	L-OR
160457	CA	36.63117	-118.80432	2153.7	H-CA
160460	CA	36.63117	-118.80432	2153.7	H-CA
160462	CA	36.63117	-118.80432	2153.7	H-CA
160637	OR	44.22465	-121.8717	1474.6	H-OR
160652	OR	45.3257	-121.65949	1594.7	H-OR
160664	OR	44.22465	-121.8717	1474.6	H-OR

114 Table S1. Collection sites for queen bees, with colony IDs matching those of queens

115

- 117 Table S2.  $CT_{MIN}$  and  $CT_{MAX}$  modeled by region. Region of origin as a categorical
- 118 variable (repeated measures ANOVA, including colony as a random effect) had a
- 119 significant effect on  $CT_{MIN}$  but not  $CT_{MAX}$ .
- 120

ANOVA*						
Treatment	Error strata	Df	SumSq	MeanSq	F	$P(\mathbf{F})$
CT <sub>MAX</sub>	Error: Colony					
	Region	3	216.9	72.31	3.009	0.095
	Residuals	8	192.2	24.03		
	Error: Within					
	Residuals	12	199.3	16.61		
CT <sub>MIN</sub>	Error: Colony					
	Region	3	246.4	82.12	8.616	0.00691
	Residuals	8	76.25	9.54		
	Error: Within					
	Residuals	12	35.58	2.965		

121 \* Generalized ANOVA model: CT<sub>TEMP</sub> ~ Region + Error(ColonyID)

123 Table S3. CT<sub>MIN</sub> and CT<sub>MAX</sub> modeled by five BioClim variables representing aspects of the thermal environment including colony as

124 a random effect using a generalized linear mixed model approach<sup>†</sup>. Four of the tested variables had a significant effect on  $CT_{MIN}$  but

		Annual Mea	an Temperature	Maximum Temperature of the Warmest Month		Minimum Temperature of the Coldest Month		Mean Diurnal Range		Isothermality	
		(Int.)	AMT	(Int.)	MTWM	(Int.)	MTCM	(Int.)	MDR	(Int.)	Iso
	Est.	-9.44	0.76	-19.68	0.65	-0.11	0.74	-26.9	1.92	-6.35	0.92
	Stat.	-7.12	5.86	-7.76	6.95	-0.11	3.34	-4.6	4.22	-0.88	0.56
	df	10	10	10	10	10	10	10	10	10	10
	р	<0.001	<0.001	<0.001	<0.001	0.917	0.007	0.001	0.002	0.401	0.59
NIN	$\sigma^2$	2.97		2.97		2.97		2.97		2.97	
CT	$ au_{00}$	2.15 ColonyID		1.28 ColonyID		6.14 ColonyID		4.31 ColonyII	D	14.16 ColonyID	
	ICC	0.42		0.3		0.67		0.59		0.83	
	Ν	12 ColonyID		12 ColonyID		12 ColonyID		12 ColonyID		$12 _{ColonyID}$	
	Obs. Marg.R <sup>2</sup> Cond.R <sup>2</sup>	24 0.680 0.814		24 0.732 0.813		24 0.448 0.820		24 0.553 0.818		24 0.024 0.831	
	Est.	54.82	-0.37	59.64	-0.31	50.16	-0.4	60.43	-0.71	51.49	-0.02
	Stat.	19	-1.3	9.48	-1.33	31.79	-1.2	5.72	-0.86	6.3	-0.01
	df	9.95	10.04	10	10.11	10.56	9.99	10.09	10.04	10.43	10.29
	р	<0.001	0.222	<0.001	0.211	<0.001	0.256	<0.001	0.409	<0.001	0.992
CT <sub>MAX</sub>	$\sigma^2$	16.78		16.85		16.69		16.55		16.6	
	$ au_{00}$	8.84 ColonyID		8.61 ColonyID		9.32 ColonyID		10.68 Colony	yID	11.94 ColonyID	
	ICC	0.34		0.34		0.36		0.39		0.42	
	Ν	13 ColonyID		13 ColonyID		13 ColonyID		13 ColonyID		13 ColonyID	
	Obs.	24		24		24		24		24	

125 none had a significant relationship with  $CT_{MAX}$ .

Marg.R <sup>2</sup>	0.090	0.094	0.079	0.043	0.000
Cond.R <sup>2</sup>	0.404	0.400	0.409	0.418	0.418

 $\dagger$ : Generalized linear mixed model: CT<sub>TEMP</sub> ~ 1+ Variable + (1|ColonyID)

- 127 Table S4. Numbers of differentially expressed genes (DEG) as a function of thermal
- 128 treatment and/or region of origin from linear models in limma. The total number of genes
- assigned to a model are based on either the *selectModel* default results or simplest model
- 130 within 2 of the absolute lowest AIC for colony blocked data. The numbers and
- 131 percentages of genes within each best-fitting model that are differentially expressed in at
- 132 least one region, treatment, or region:treatment contrast in the individual gene analyses
- 133 (FDR cutoff of  $\leq 0.05$ ) are provided in parentheses. Given that the gene assignment to a
- 134 linear model necessitates exclusion from other possible models, any individual gene may 135 be significant in contrasts that don't align with its linear model categorization.
- 136

Model Name	Model Formula	selectModel	Simplest Model
		(DEG)	(DEG)
Total Genes =	9328		
None	~ 1	3654 (0, 0%)	5032 (1, 0%)
Region	$\sim 0 + \text{Region}$	2855 (57, 2%)	2581 (66, 3%)
Treatment	$\sim 0 + Treatment$	1393 (201, 14%)	1049 (242, 23%)
Treatment +	$\sim 0 + \text{Treatment} + \text{Region}$	1056 (122, 11%)	494 (101, 20%)
Region			
Full	$\sim 0 + \text{Treatment} + \text{Region} +$	370 (62, 17%)	172 (32, 19%)
	Treatment*Region		

## 139 Dataset S1 (separate file)

Results of physiological test of thermal maximums and minimums of *Bombus vosnesenskii* bumble bee workers raised in a common laboratory environment from
queens collected from altitudinally and latitudinally distinct populations.

## 143 Dataset S2 (separate file)

144 Results of edgeR/Limma individual gene-by-gene analysis of differential gene

145 expression for thermally stressed *Bombus vosnesenskii* bumble bee workers raised in a

146 common laboratory environment from queens collected from altitudinally and

147 latitudinally distinct populations, including gene ontology results.

# 148 Dataset S3 (separate file)

149 Results of WGCNA analysis of patterns of differential gene expression for thermally

- 150 stressed *Bombus vosnesenskii* bumble bee workers raised in a common laboratory
- 151 environment from queens collected from altitudinally and latitudinally distinct
- 152 populations, including gene ontology results.
- 153
- 154
- 155

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