# Transcriptominc divergence of a parasite-populations: two common garden experients in two hosts

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Abstract

Background:

Results:

Conclusions: Yeh!

### Background Results

The populations differ

We sampled worms The gene expression does too

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Worm Sex

scvPlot(cds.mf)

residualsEcdfPlot( cds.mf, "male" ) residualsEcdfPlot( cds.mf, "female" )

label	sex	host	population	intensity	worms in prep	conc in prep
AA/T20F	female	AA	T	1	1	5.60
AA/T12F	female	AA	${ m T}$	14	1	6.80
AA/T45F	female	AA	${ m T}$	5	1	8.00
AA/T24M	$_{\mathrm{male}}$	AA	${ m T}$	6	3	4.80
AA/T42M	male	AA	${ m T}$	11	1	5.60
AA/T3M	male	AA	${ m T}$	5	4	4.88
AA/R18F	female	AA	R	4	1	4.80
AA/R28F	female	AA	R	10	1	5.20
AA/R8F	female	AA	R	27	1	5.20
AA/R16M	male	AA	R	10	4	5.20
AA/R11M	$_{\mathrm{male}}$	AA	R	25	14	6.40
AA/R2M	$_{\mathrm{male}}$	AA	R	10	4	6.60
AJ/T8F	female	AJ	${ m T}$	10	1	5.91
AJ/T5F	female	AJ	${ m T}$	2	1	4.80
AJ/T26F	female	AJ	${ m T}$	2	1	2.40
AJ/T25M	$_{\mathrm{male}}$	AJ	${ m T}$	24	5	4.05
AJ/T19M	$_{\mathrm{male}}$	AJ	${ m T}$	24	7	3.50
AJ/T20M	male	AJ	${ m T}$	20	8	3.80
AJ/R1F	female	AJ	R	3	1	5.92
AJ/R3F	female	AJ	R	3	1	6.90
AJ/R5F	female	AJ	R	10	1	4.04
AJ/R1M	male	AJ	R	3	1	2.50
AJ/R3M	male	AJ	R	3	2	2.60
AJ/R5M	male	AJ	R	10	1	2.23

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\label{eq:policy} $\operatorname{plot}(\ \operatorname{res.mf} baseMean, res.mf \log 2 \operatorname{FoldChange}, \ \log = "x", \ \operatorname{pch} = 20, \ \operatorname{cex} = .4, \ \operatorname{col} = \operatorname{ifelse}(\ \operatorname{res.mf} padj < .01, "red", "black"), main = "Mydata")$$ Eel species $\operatorname{scvPlot}(\operatorname{cds.eel})$$ residualsEcdfPlot( cds.eel, "Aa") residualsEcdfPlot( cds.eel, "Aj")$$ plot( res.eelbaseMean, res.eellog2FoldChange, log="x", pch=20, cex=.4, col = ifelse( res.eelpadj < .01, "red", "black"), main = "Mydata")$$ Populations $\operatorname{scvPlot}(\operatorname{cds.pop})$$ residualsEcdfPlot( cds.pop, "EU") residualsEcdfPlot( cds.pop, "TW") $$ plot( res.popbaseMean, res.poplog2FoldChange, log="x", pch=20, cex=.4, col = ifelse( res.poppadj < .01, "red", "black"), main = "Differencesbetweenpopulations")$$
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library	raw.reads	raw.mapped	tax.mapped	screened
AA_R11M	11986442	8628520	7868814	6889551
$AA_R16M$	10810349	6858585	6217540	5276284
$AA_R18F$	9227615	6552527	5933235	5200958
$AA_R28F$	10135670	6665381	6005399	5171806
$AA_R2M$	12469746	7628428	6929651	5906422
$AA_R8F$	15270570	11527867	10758535	9453468
$AA_T12F$	11299438	7842479	7195621	6332396
$AA_T20F$	11740839	7744179	7114349	6323422
$AA_{-}T24M$	8552723	5254194	4662053	3969305
$AA_{-}T3M$	11031751	6460836	5800042	4993726
$AA_T42M$	11573501	7567845	6787375	5694801
$AA_T45F$	10646847	7714472	7173709	6283585
$AJ_R1F$	9855005	6400558	5890748	5167912
$AJ_R1M$	10211903	5851063	5313544	4506254
$AJ_R3F$	9897937	6425201	5948079	5124077
$AJ_R3M$	8775211	4562324	4073621	3422526
$AJ_R5F$	11949105	8442537	7830247	6882280
$AJ_R5M$	11231532	7504494	6772010	5913016
$AJ_T19M$	9195576	4798404	4293123	3635843
$AJ_T20M$	10862591	6880937	6251674	5280529
${ m AJ\_T25M}$	11195315	7162880	6480185	5645097
$AJ_T26F$	11195335	7439917	6641973	6031374
$AJ_{-}T5F$	10357569	7413685	6794507	6007930
$AJ_{-}T8F$	14196382	10275074	9496489	8364594

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## Discussion Conclusions

#### Methods

#### General coding methods

The bulk of analysis (unless otherwise cited) presented in this paper was carried out in R [1] using custom scripts. We used a method provided in the R-packages Sweave [2] and Weaver [3] for "reproducible research" combining R and TEXcode in a single file. All intermediate data files needed to compile the present manuscript from data-sources are provided upon request. For visualistation we used the R-packages lattice [4] and ggplot2 [5].

#### **Competing interests**

The authors declare no competing interests.

### Authors contributions Acknowledgments

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- Figures
  Figure 1 Figure 2 Figure 3 Figure 4 Figure 5 -

**Tables** 

Table 1 -

Table 7 -

**Additional Files**