

Integrating gene expression data into Genomic prediction

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My project overview

- (1) to integrate transcriptome data with whole-genome sequence data for genomic-transcriptomic prediction
- (2) to assess whether our statistical models can capture substantial proportions of phenotypic variance explained by transcriptome data
- (3) to test whether accounting for transcriptome data can improve phenotype prediction.



Materials

***Drosophila* Genetic Reference Panel (DGRP)**

- **Availability of Supporting Data:** <http://dgrp2.gnets.ncsu.edu>
- *Drosophila* Whole-Genome Sequence Data
- *Drosophila* Gene Expression Data
- *Drosophila* Phenotype Data



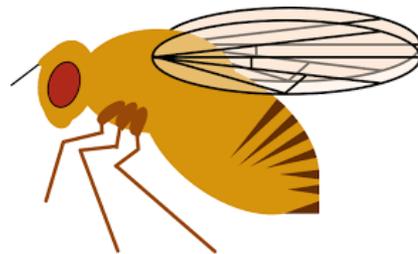


Whole-Genome Sequence Data

- 2'863'909 SNPs from 185 fully sequenced inbred lines

Expression Data

- The abundances of RNA products of 18'140 genome-wide annotated genes and novel transcribed regions (NTRs) in 185 DGRP lines





Phenotype data

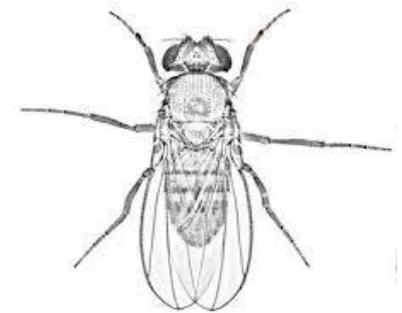
(all phenotypes are based on the line mean of several observations)

9 traits

1. Startle response (SR)
2. Starvation resistance (SRE)
3. Food intake (FI)
4. Alcohol sensitivity and tolerance (AST)

Olfactory perceptions to 5 chemical odorants:

5. Olfactory perceptions to Ethyl Butyrate (OPEB)
6. Olfactory perceptions to 2-Heptanone (OP2H)
7. Olfactory perceptions to Methyl Salicylate (OPMS)
8. Olfactory perceptions to l-Carvone (OPIC)
9. Olfactory perceptions to 1-Hexanol (OP1H)





Statistical models

methods	data
GBLUP	Whole-Genome Sequence Data
TBLUP (linear kernel)	Gene Expression data
Reproducing Kernel Hilbert Space Regression (RKHSR) (Gaussian kernel)	Gene Expression data
GTBLUP (GBLUP + TBLUP)	Whole-Genome Sequence Data Gene Expression data
GRBLUP (GBLUP + RKHSR)	Whole-Genome Sequence Data Gene Expression data



Comparison of predictive abilities

- Pearson correlation coefficients between predicted phenotypes and observed phenotypes.
- The different approaches were assessed using 20 replicates of a 5-fold cross-validation (CV).
- The final predictive ability of each model was the mean of the prediction accuracies across 100 estimates.



Estimation of omics-augmented broad sense heritability (based on the between line effects)

The omics-augmented broad sense heritability is defined as the proportion of phenotypic variance explained by whole genome SNP marker effects and other utilized omics data effects.

$$\hat{H}_o^2 = \frac{\hat{\sigma}_g^2 + \hat{\sigma}_{omics}^2}{\hat{\sigma}_g^2 + \hat{\sigma}_{omics}^2 + \hat{\sigma}_e^2}$$



(1) SNP-based genomic narrow sense heritability

for GBLUP (\hat{h}_G^2)

$$\hat{h}_G^2 = \frac{\hat{\sigma}_g^2}{\hat{\sigma}_g^2 + \hat{\sigma}_e^2}$$

(2) SNP and gene expression data-augmented broad sense heritability

for GTBLUP (\hat{H}_{GT}^2)

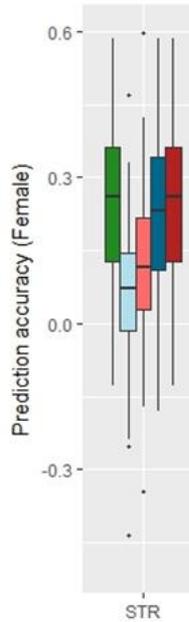
$$\hat{H}_{GT}^2 = \frac{\hat{\sigma}_g^2 + \hat{\sigma}_t^2}{\hat{\sigma}_g^2 + \hat{\sigma}_t^2 + \hat{\sigma}_e^2}$$

for GRBLUP (\hat{H}_{GR}^2)

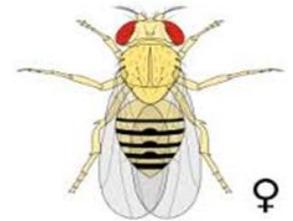
$$\hat{H}_{GR}^2 = \frac{\hat{\sigma}_g^2 + \hat{\sigma}_v^2}{\hat{\sigma}_g^2 + \hat{\sigma}_v^2 + \hat{\sigma}_e^2}$$

The variance components $\hat{\sigma}_g^2$, $\hat{\sigma}_t^2$, $\hat{\sigma}_v^2$, $\hat{\sigma}_e^2$ were estimated from the entire data sets.

Results

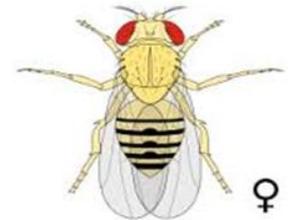
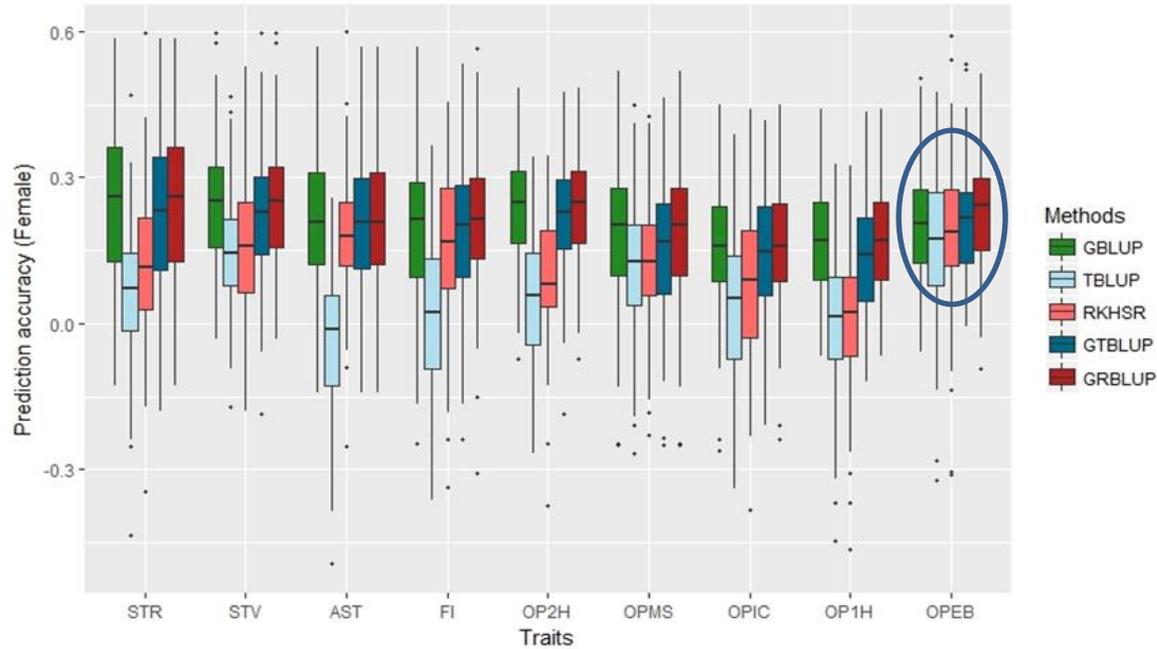


- Methods
- GBLUP
 - TBLUP
 - RKHSR
 - GTBLUP
 - GRBLUP

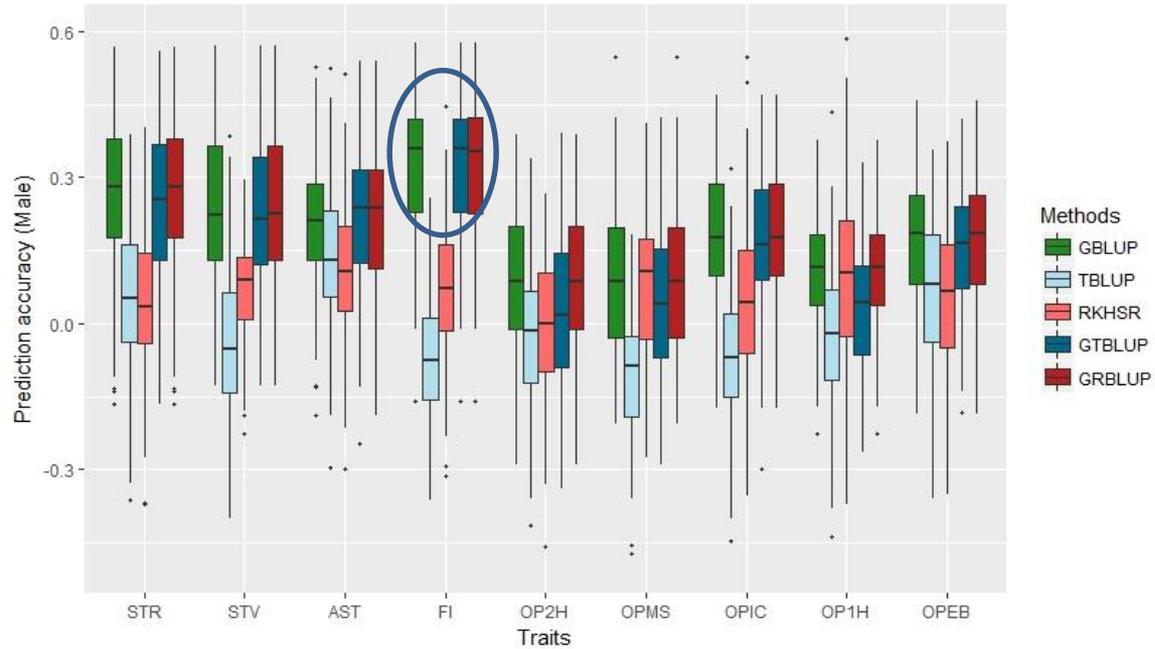


Predictive abilities for 9 traits with 5 statistical models in females

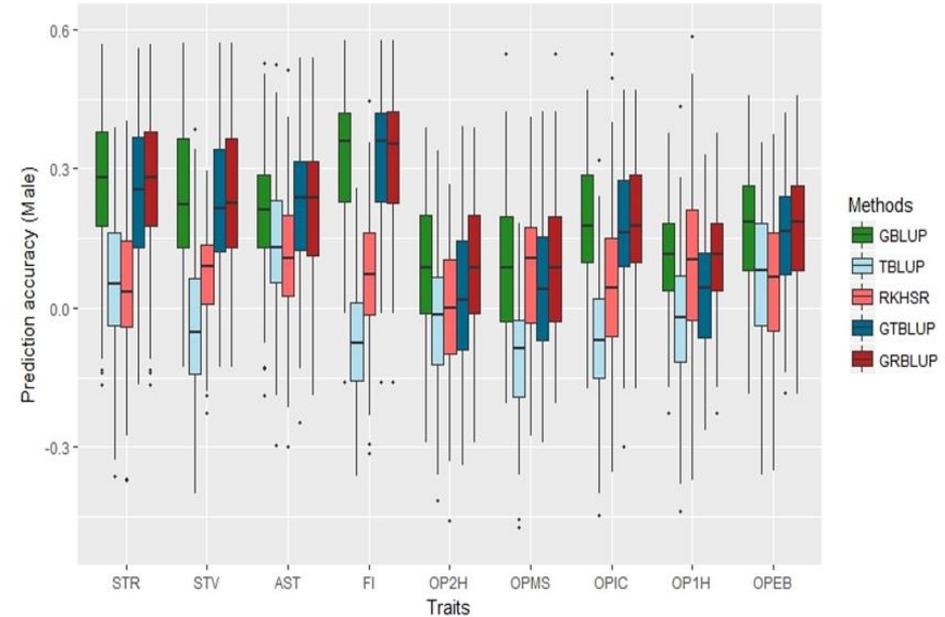
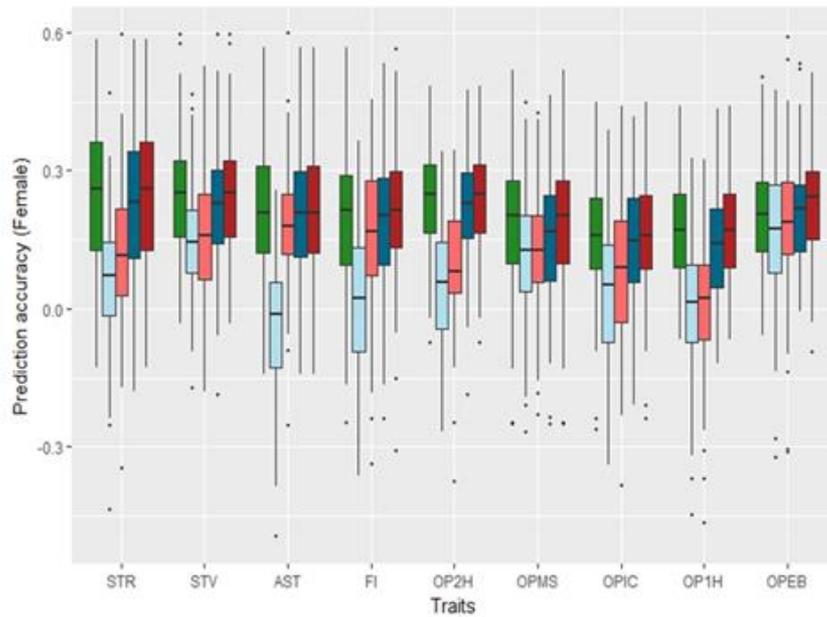
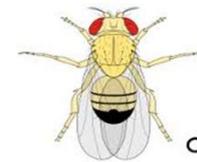
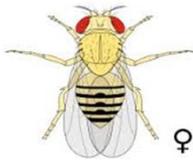
Results



Predictive abilities for 9 traits with 5 statistical models in females



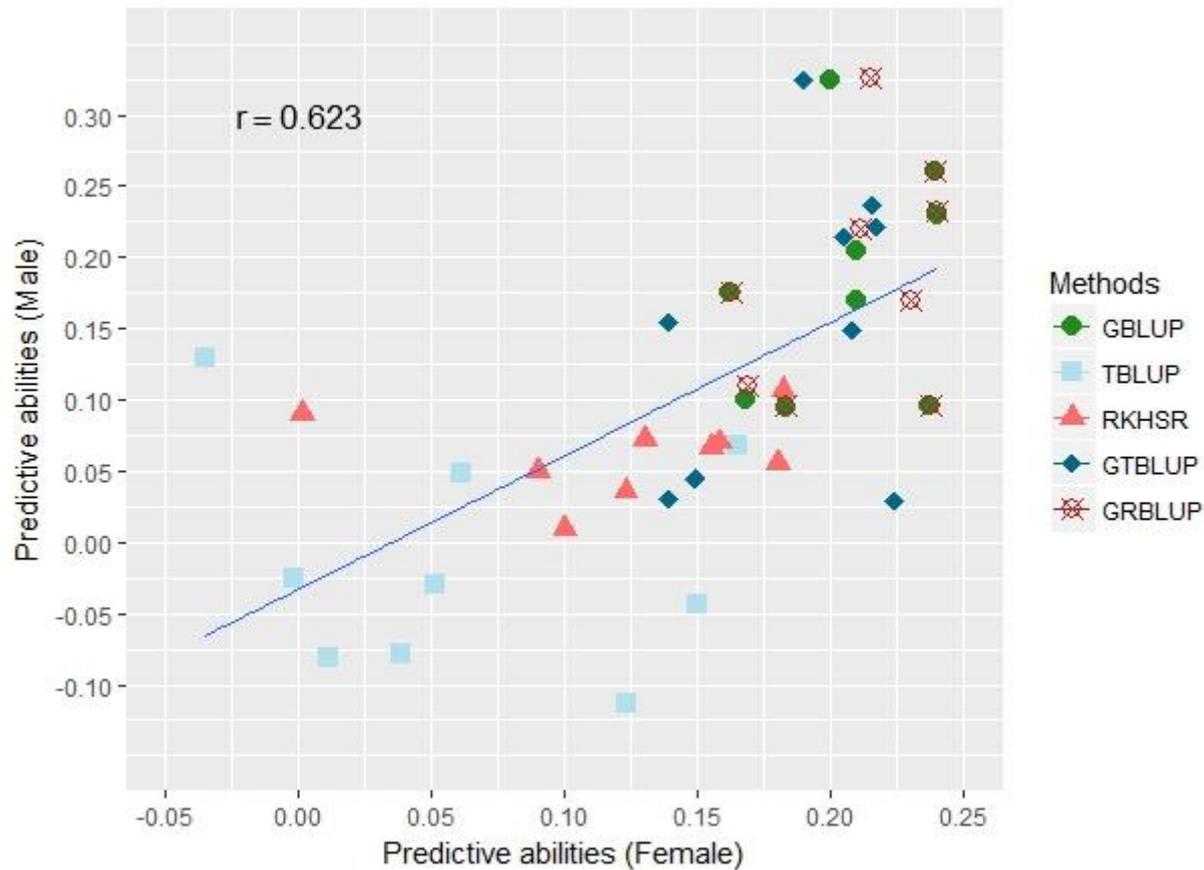
Predictive abilities for 9 traits with 5 statistical models in males



Comparison of predictive abilities between females and males for 9 traits with 5 statistical models

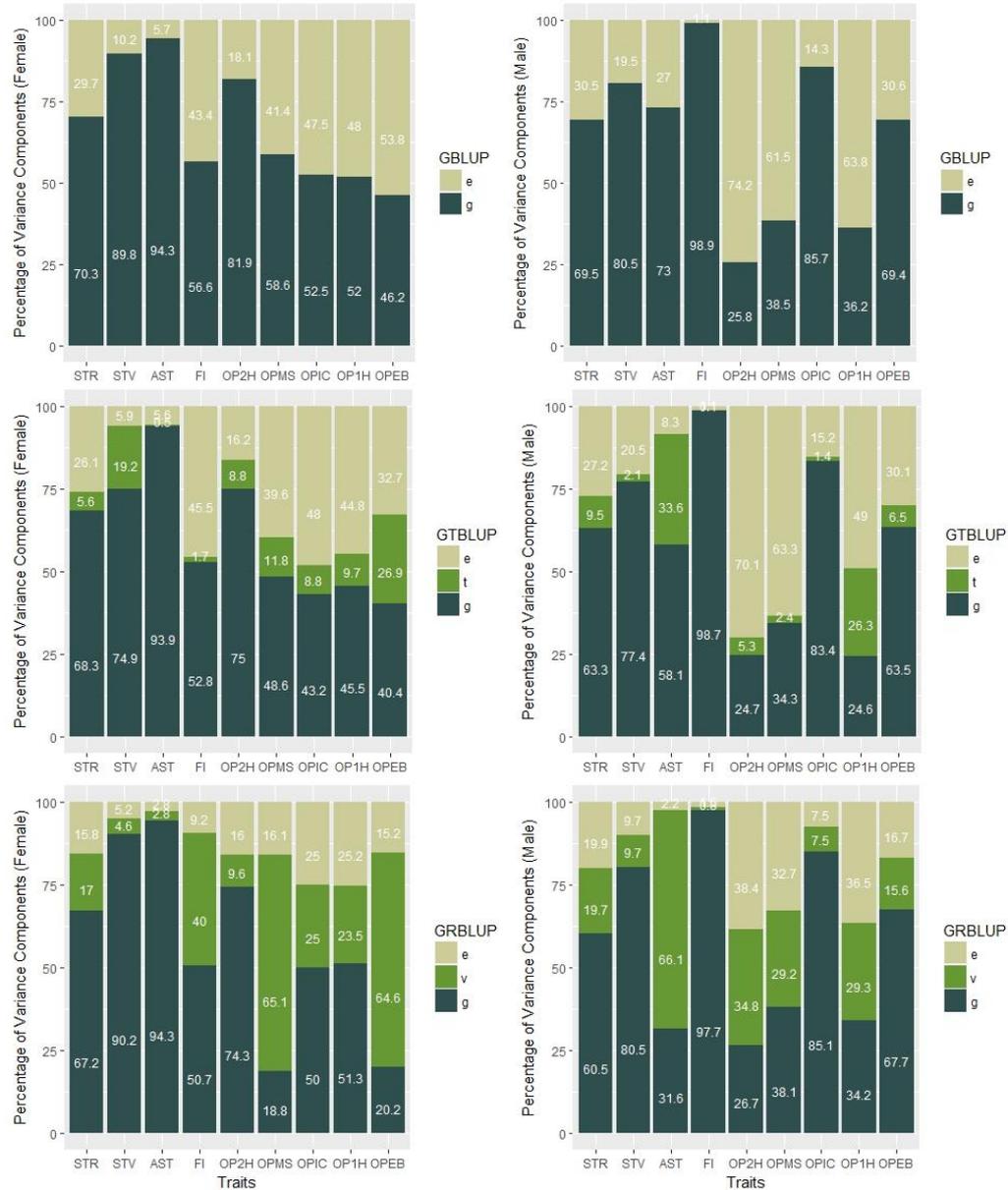


Correlation of predictive abilities between females and males across 9 traits and 5 statistical models

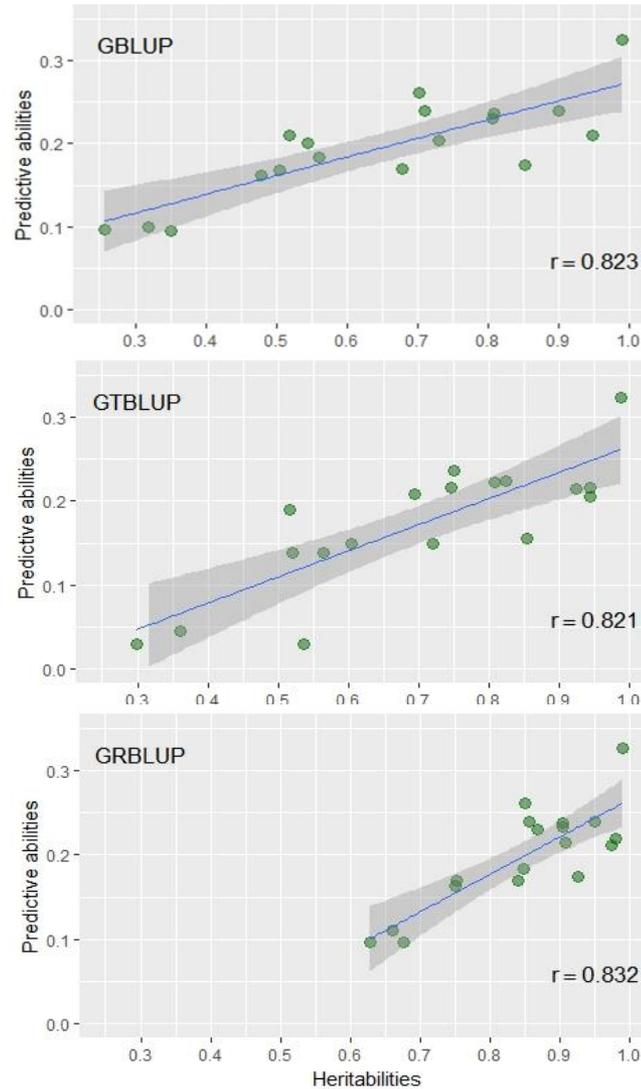




Percentage of variance components for GBLUP, GTBLUP and GRBLUP on 9 traits for females (left) and males (right)



The correlation between predictive abilities and heritabilities for GBLUP, GTBLUP, GRBLUP across all traits and both sex





Conclusion

- GRBLUP and GBLUP provided similar predictive ability, but GRBLUP could capture more phenotypic variance components explained by transcriptome data.
- GRBLUP has excellent goodness of fit with transcriptome data. The better goodness of fit of GRBLUP in general did not translate into a better predictive ability in most traits.
- The sample size was (too) small
- Gene expression was not measured at a time point and in a specific tissue which was functionally linked to the trait of interest.
- We suspect that adding more specifically collected transcriptome data has the potential to improve genomic predictions in larger scale applications.



Acknowledgements

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Thanks for listening