

Analysis of submitted breeding values and applied methods

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Common task

- Predict breeding values for young individuals without phenotypes
- 11 teams provided results for quantitative (QT) and 6 for binary trait (BT)

True Breeding Value

- QT:

$TBV = \sum[30 \text{ additive QTLs}] +$
+ haplotype effects (QTLs pairs 31-32 and 33-34) +
+ the effects of imprinted QTLs (for males only).

- BT:

$TBV = \sum 22 \text{ additive QTLs.}$

Methods used by Participants

- BLUP:
 - PBLUP(pedigree; univariate and bivariate),
 - TA-BLUP(specific relationship matrix),
 - G(enomic)BLUP (univariate and bivariate)
 - RR-BLUP – ridge regression
 - Spatial
- Bayes A,B,C (univariate and bivariate)
- PLSR
- Double Hierarchical Generalized Linear Models (DHGLM)
- Machine learning: Support vector , Boosting, Genome-wide Rapid Association

Comparison criteria

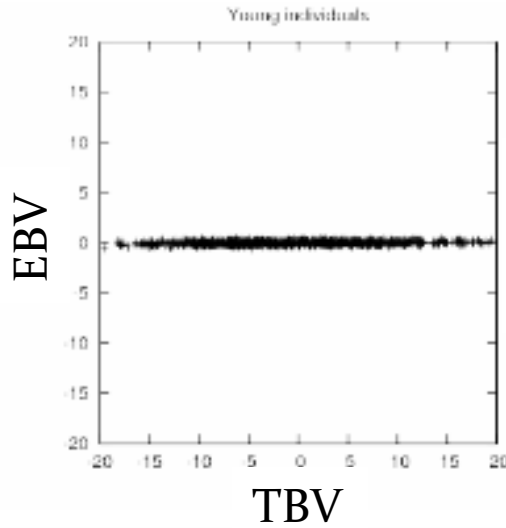
- **Accuracy** - Pearson correlation between TBV and EBV
- **Regression** - linear regression coefficient TBV on EBV
- **MSD** - mean square difference (TBV-EBV) after correcting for mean

Comparison criteria cont.

- **Ranking comparison:**
 - **Shared** - percentage of IDs shared between the groups of young individuals selected on TBV and EBV
 - **Loss** - % of loss of response to selection when 10% are selected based on EBV instead of TBV

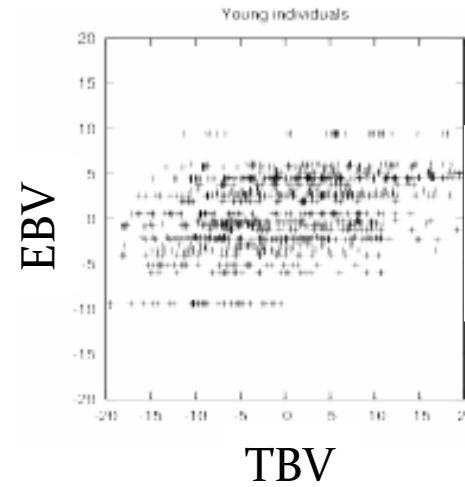
Accuracy of EBV - examples

QT



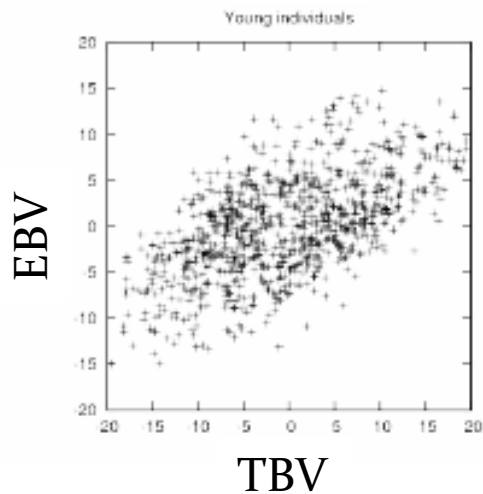
PLSR

$r = 0.24$



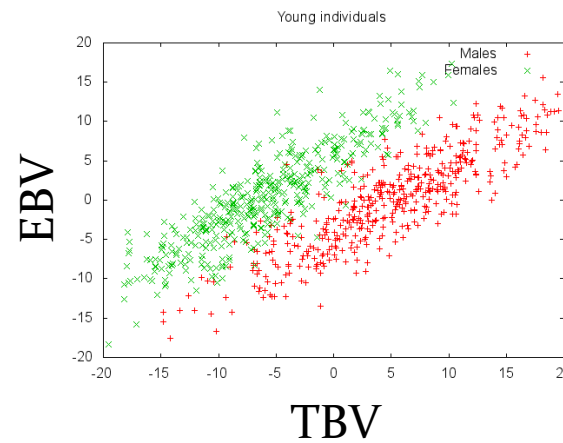
PBLUP

$r = 0.39$



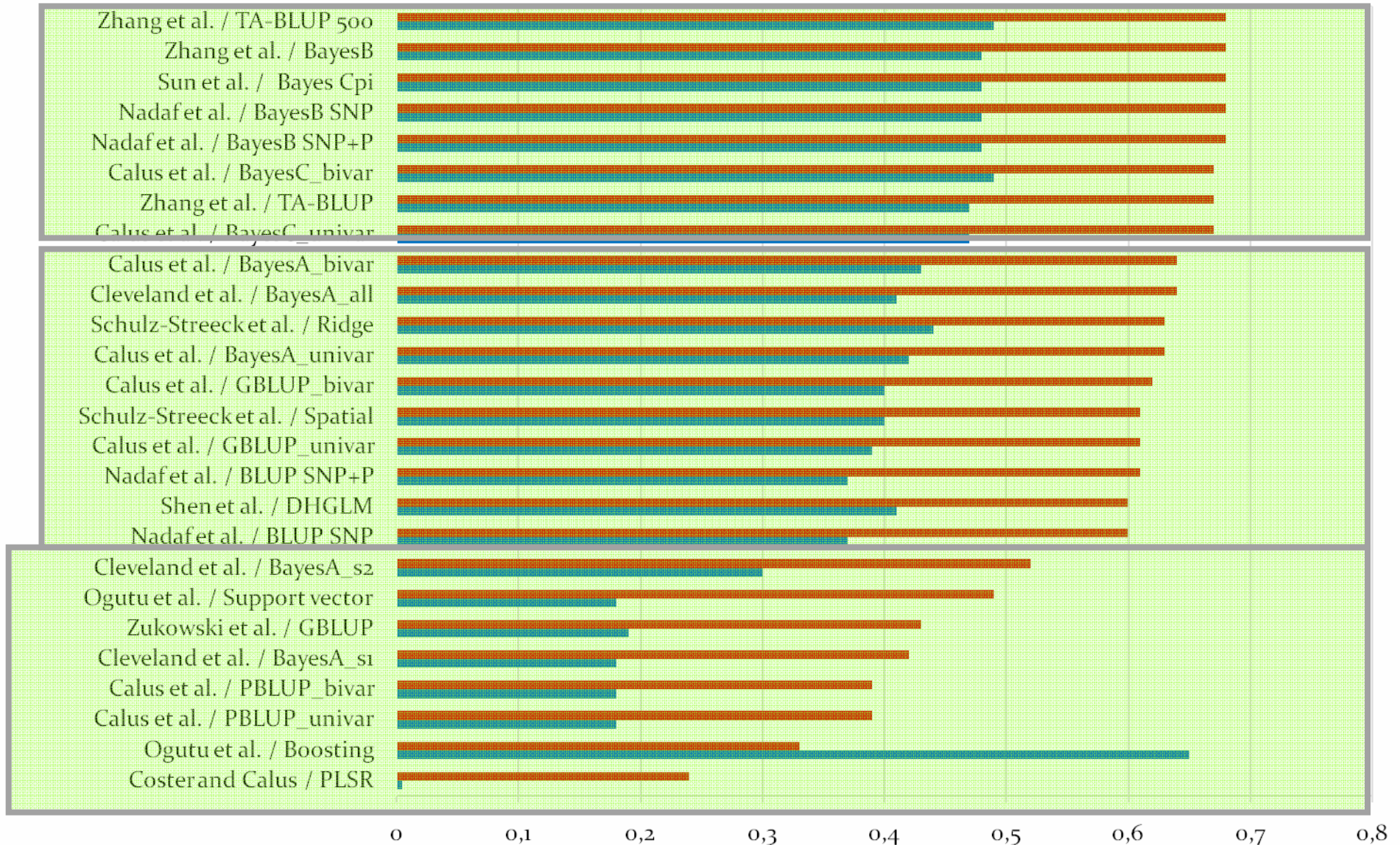
GBLUP Bivar

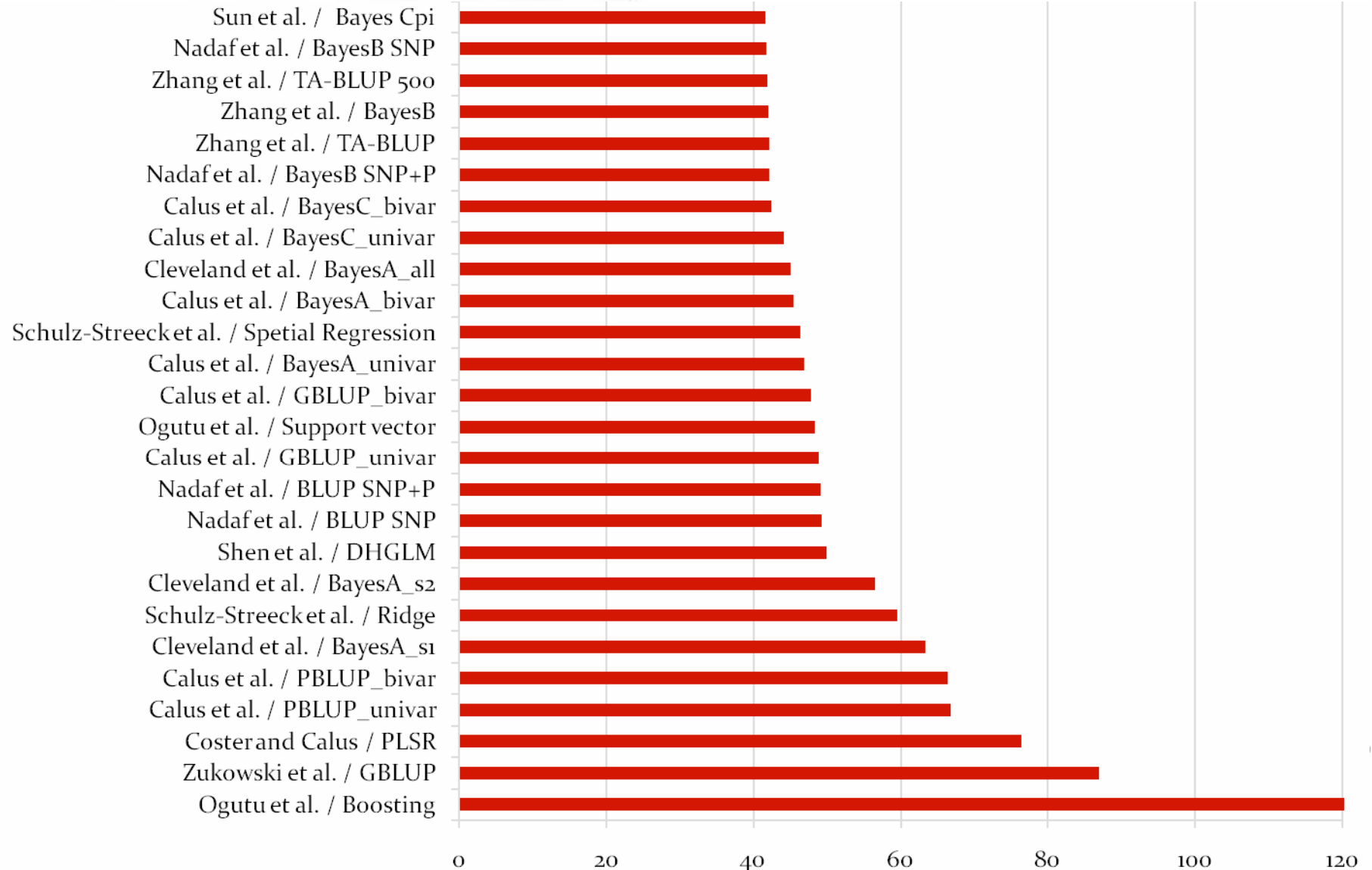
$rr = 0.62$

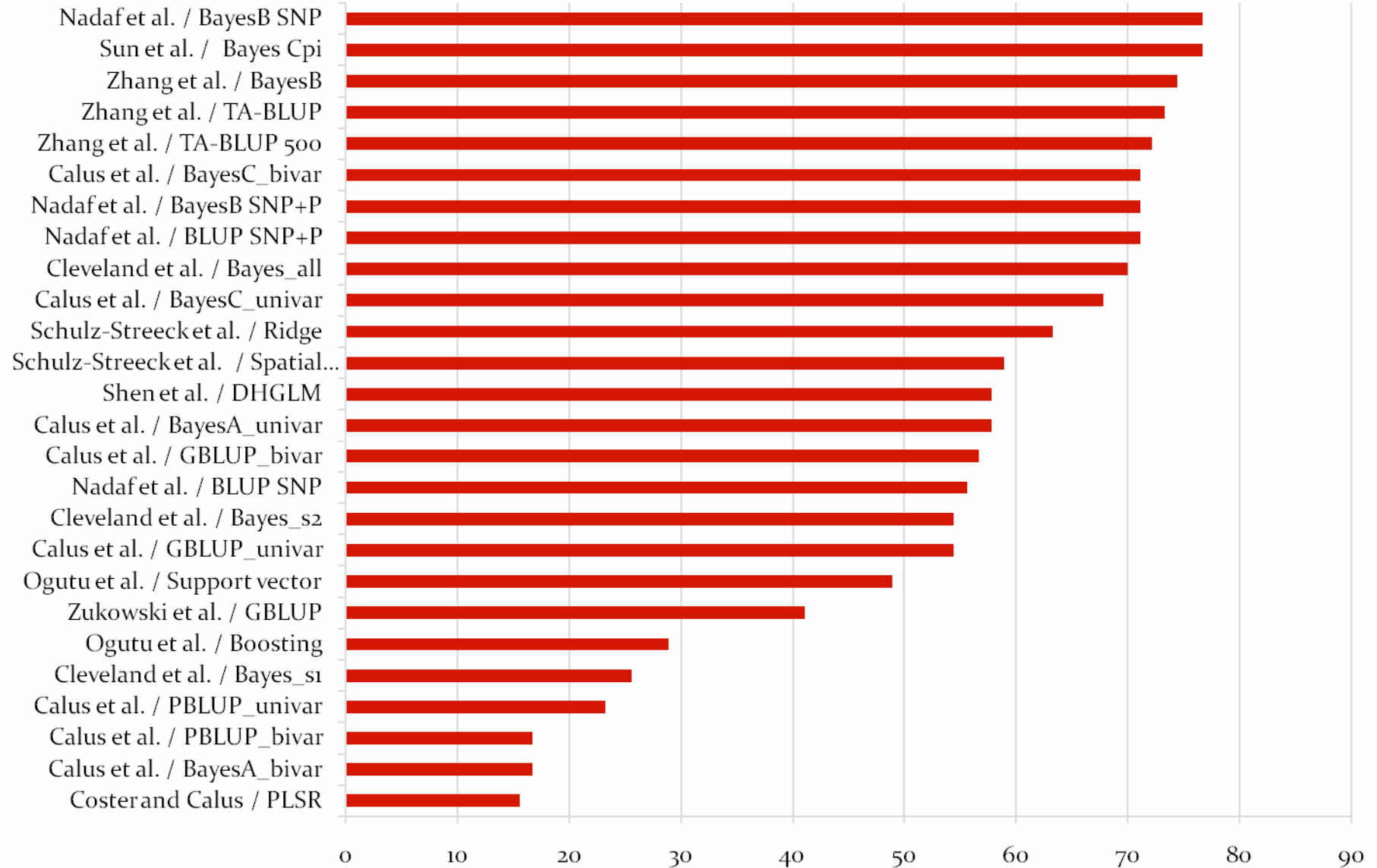


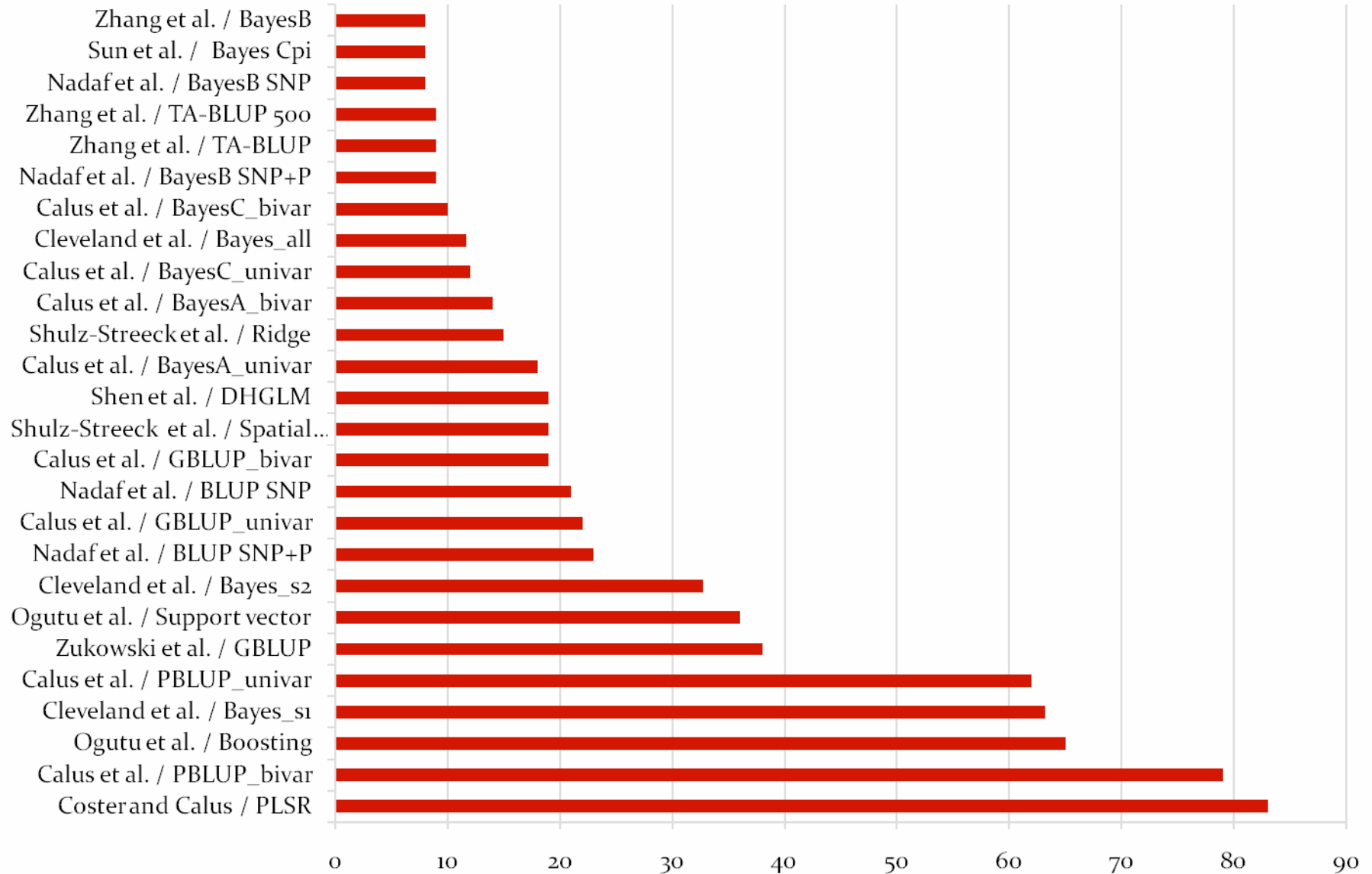
BayesC Bivar

$r = 0.67$

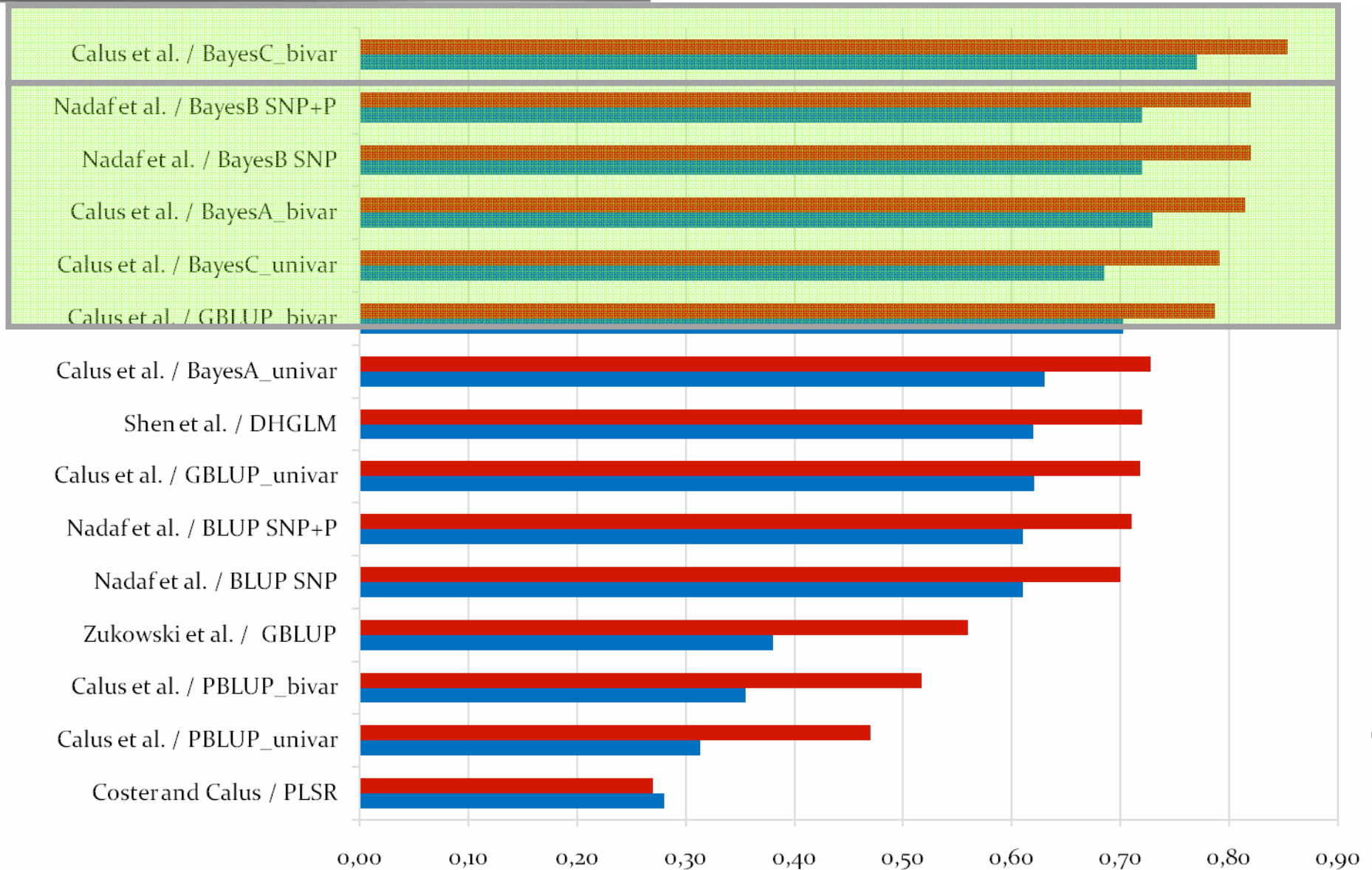


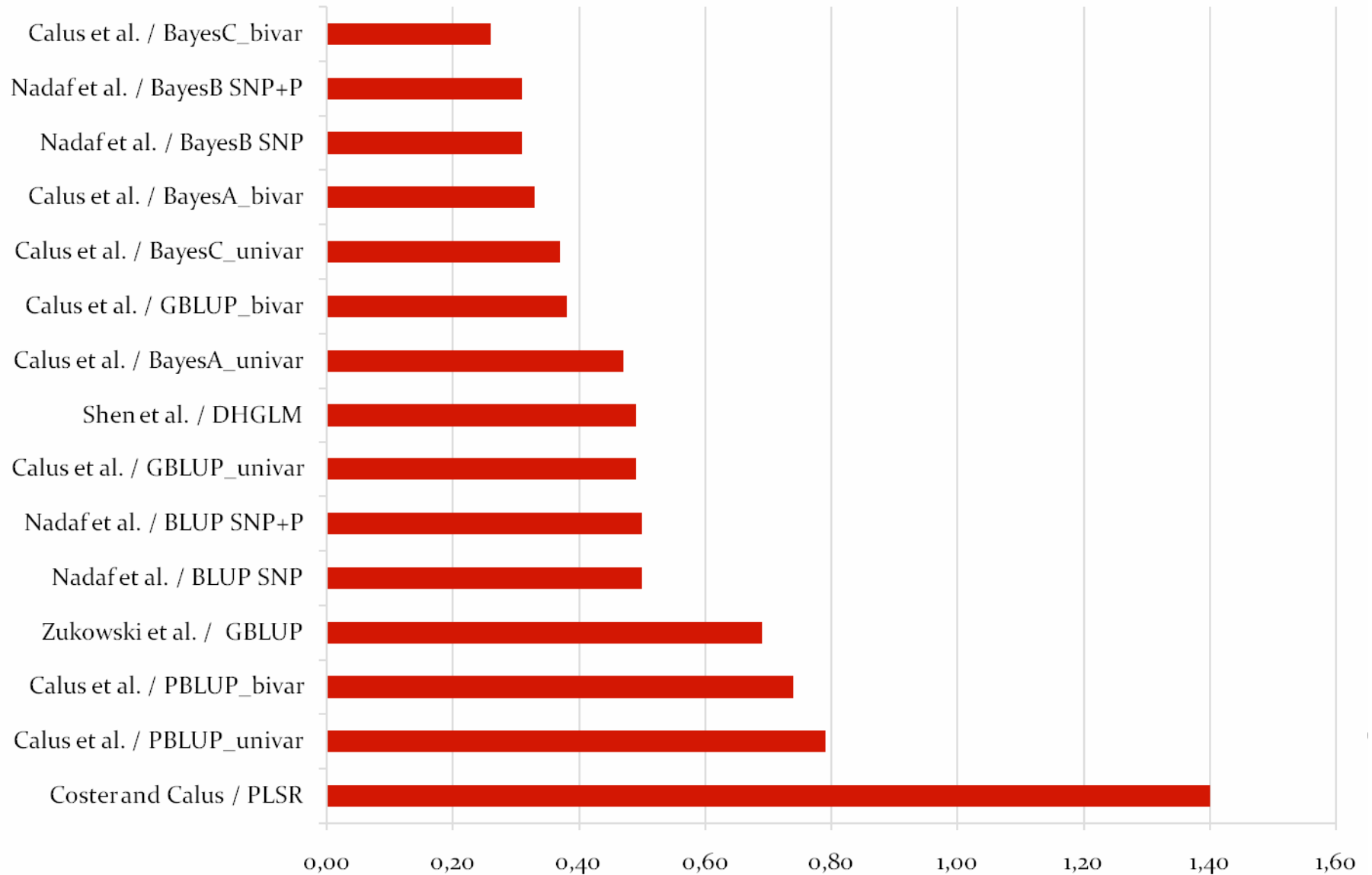


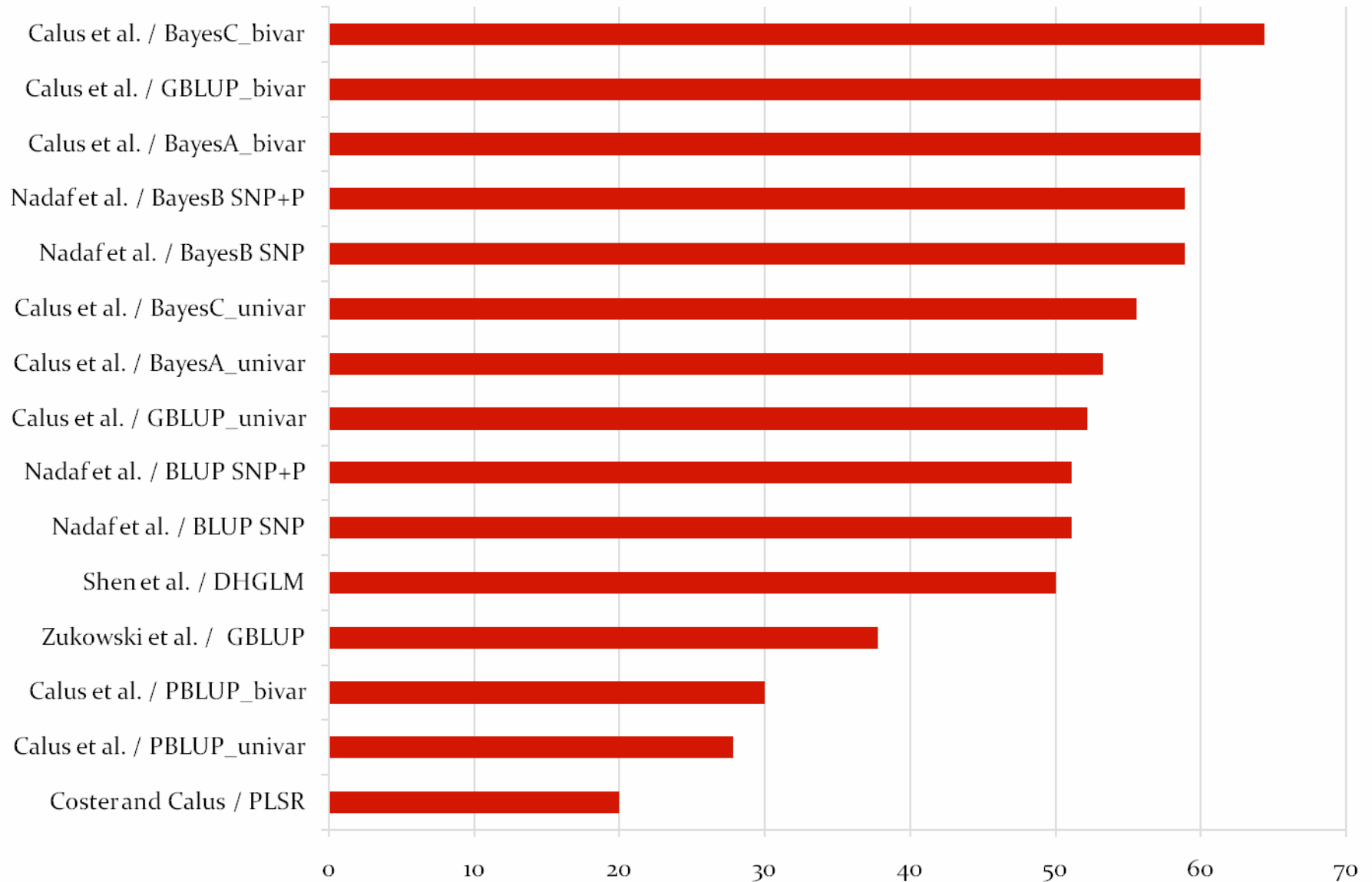


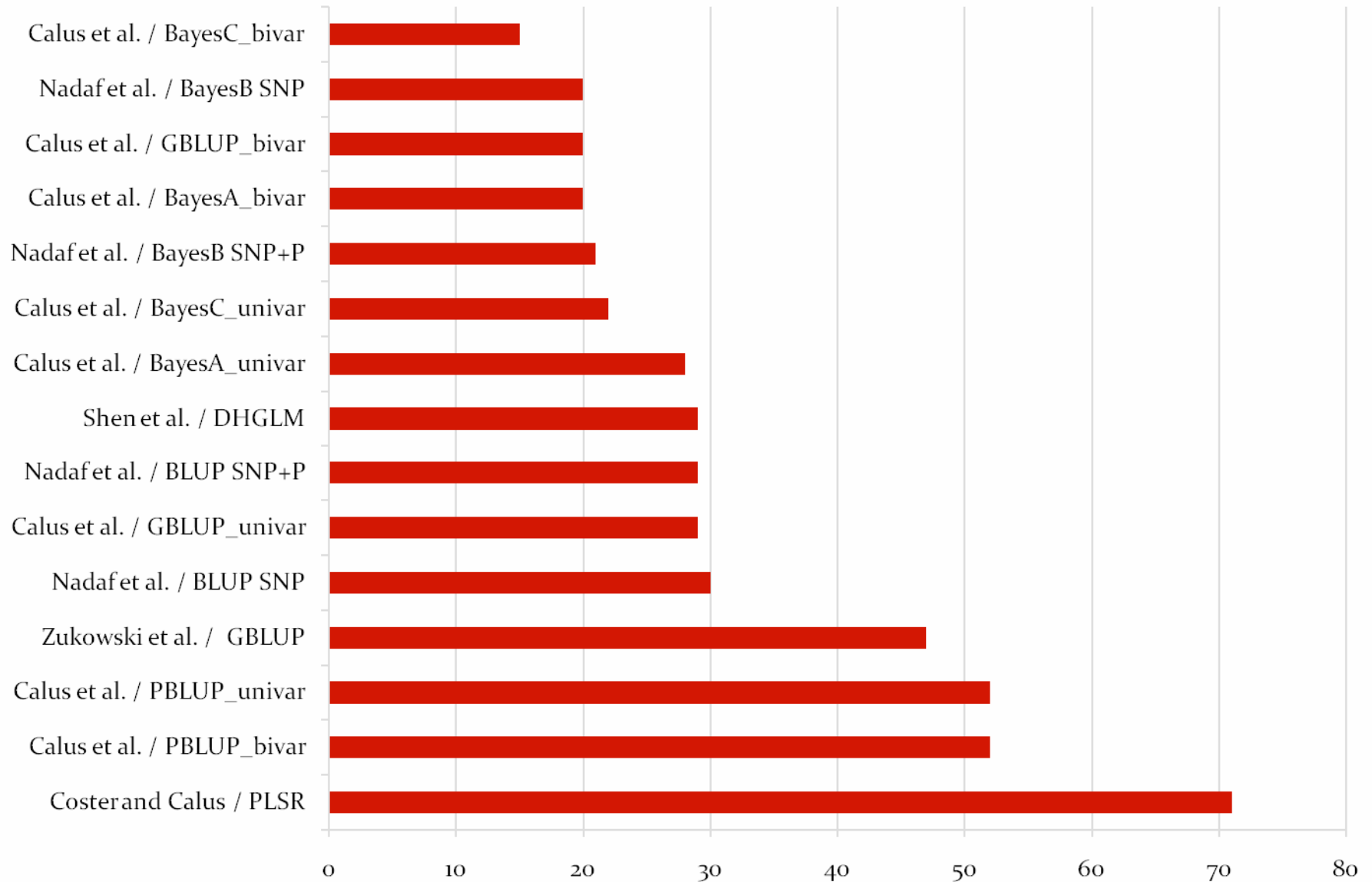


1. Nadaf et al. - BayesB SNP
2. Sun et al. - Bayes Cpi
3. Nadaf et al. - BayesB SNP+P
4. Zhang et al. - BayesB
5. Zhang et al. - TA-BLUP 500
6. Zhang et al. - TA-BLUP
7. Calus et al. - BayesC_bivar
8. Calus et al. - BayesC_univar
9. Calus et al. - BayesA_bivar
10. Schulz-Streeck et al. - RR
11. Calus et al. - GBLUP_bivar
12. Schulz-Streeck et al. - Spatial
13. Calus et al. - BayeaA_univar
14. Calus et al. - GBLUP_univar
15. Shen et al. - DHGLM
16. Nadaf et al. - BLUP SNP+P
17. Nadaf et al. - BLUP SNP
18. Ogutu et al. - Boosting
19. Calus et al. - PBLUP_univar
20. Ogutu et al. - Support vector
21. Calus et al. - PBLUP_bivar
22. Zukowski et al. - GBLUP
23. Coster and Calus - PLSR









1. Calus et al. - BayesC_bivar
2. Calus et al. - BayesA_bivar
3. Nadaf et al. - BayesB SNP
4. Nadaf et al. - BayesB SNP+P
5. Calus et al. - GBLUP_bivar
6. Calus et al. - BayesC_univar
7. Calus et al. - BayesA_univar
8. Calus et al. - GBLUP_univar
9. Shen et al. - DHGLM
10. Nadaf et al. - BLUP SNP+P
11. Nadaf et al. - BLUP SNP
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13. Calus et al. - PBLUP_bivar
14. Calus et al. - PBLUP_univar
15. Coster and Calus - PLSR

- For QT:
 - BayesB, TA-BLUP, BayesC, RR-BLUP > GBLUP, Spatial, BayesA > PBLUP, Machine Learning
 - For BT:
 - BayesC, BayesB, BayesA, GBLUP > Other methods
 - Bivariate > Univariate for the same method
 - Bayes and TA-BLUP methods better for complex genetic architecture
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Thank you for sharing the results !!!

Congratulations to the authors !!!