# Comparing genomic prediction methods on wheat data

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September 11, 2015

This vignette is designed to illustrate the application of dscr package in genomic prediction.

We will assess four methods for genotype-based prediction on a public wheat dataset. This dataset is available in BLR package. We will compare four methods: Bayes Lasso (from BLR package), BSLMM (from GEMMA software), rrBLUP and rrBLUP with a Gaussian kernel (from rrBLUP pacakge).

### Quick start

1. Download the repository from GitHub.

```
wget https://github.com/xiangzhu/dscr_blm/archive/master.zip
unzip master.zip
rm master.zip
```

After this step, there is a local dscr\_blm folder.

```
[xiangzhu@office] ls -1
drwxrwxr-x 3 xiangzhu xiangzhu
                                 4096 Sep 10 17:39 datamakers
drwxrwxr-x 2 xiangzhu xiangzhu
                                 4096 Sep 10 17:37 methods
-rw-rw-r-- 1 xiangzhu xiangzhu
                                 505 Sep 11 14:29 methods.R
                                 4798 Sep 10 13:47 README.md
-rw-rw-r-- 1 xiangzhu xiangzhu
-rw-rw-r-- 1 xiangzhu xiangzhu
                                 151 Sep 11 10:38 run_dsc.R
                                  160 Sep 11 14:30 scenarios.R
-rw-rw-r-- 1 xiangzhu xiangzhu
                                  549 Sep 11 13:57 score.R
-rw-rw-r-- 1 xiangzhu xiangzhu
-rw-rw-r-- 1 xiangzhu xiangzhu 177728 Sep 11 15:16 vignette.pdf
-rw-rw-r-- 1 xiangzhu xiangzhu
                                 6503 Sep 11 15:17 vignette.Rmd
```

2. Install the required R packages.

```
# for dsc
install.packages("devtools")
devtools::install_github("stephens999/ashr") # required by one vignette of dscr
devtools::install_github("stephens999/dscr",build_vignettes=TRUE)

# for genomic prediction
install.packages("BLR")
install.packages("rrBLUP")
install.packages("BGLR")
```

3. Go to the local dscr\_blm directory. Start an R console and run run\_dsc.R.

```
source('run_dsc.R')
```

After this step, there is a new folder dsc\_blm containing all results of the comparisons.

```
[xiangzhu@office] ls -1
drwxrwxr-x 3 xiangzhu xiangzhu
                                 4096 Sep 10 17:39 datamakers
drwxrwxr-x 6 xiangzhu xiangzhu
                                 4096 Sep 11 14:31 dsc_blm
drwxrwxr-x 2 xiangzhu xiangzhu
                                 4096 Sep 10 17:37 methods
-rw-rw-r-- 1 xiangzhu xiangzhu
                                  505 Sep 11 14:29 methods.R
-rw-rw-r-- 1 xiangzhu xiangzhu
                                 4798 Sep 10 13:47 README.md
-rw-rw-r-- 1 xiangzhu xiangzhu
                                  151 Sep 11 10:38 run_dsc.R
-rw-rw-r-- 1 xiangzhu xiangzhu
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                                  549 Sep 11 13:57 score.R
-rw-rw-r-- 1 xiangzhu xiangzhu 177728 Sep 11 15:16 vignette.pdf
-rw-rw-r-- 1 xiangzhu xiangzhu
                                 6503 Sep 11 15:17 vignette.Rmd
```

## Step-by-step illustration

If dscr package is used for the first time, the following installation is necessary.

First load the library, and initialize a new dsc using new\_dsc. This object will be used to store the details of the dsc.

```
library(dscr)
dsc_blm = new_dsc("genomic_prediction","dsc_blm")
```

Now define the function datamaker to create data, which consists of input and meta data. The input is what the methods will be given. The meta data will be used when scoring the methods. Here, the input is a list with three elements,

- input\$data.name: the name of data set (character string)
- input\$phenotype.id: the identifier of phenotype for multiple-phenotype dataset (integer)
- input\$test.subject: the identifiers of individuals that are assigned to the test set (integer vector)

and the meta is a list with one element, meta\$true.value, the observed phenotypes in the test set. The datamaker function takes a single parameter args, which is a list with three elements: args\$data.name, args\$phenotype.id (same as input) and args\$test.size (the sample size of test set, integer). The function datamaker.R is in folder datamaker/.

Now use datamaker to define scenarios. Each scenario is determined by the datamaker, its arguments and the random seeds it uses. The name wheat is used for naming the scenario in results and also output directories.

```
source_dir("datamakers") # need datamaker to create scenarios

# use the first kind of phenotype in wheat data
# set the sample size of test set as 100
args_eg1 = list(data.name="wheat",test.size=100,phenotype.id=1)

# repeat the random partition of data for 20 times, using seed 1-20
add_scenario(dsc_blm,name="wheat",datamaker,args=args_eg1,seed=1:20)
```

Now define the methods. They share the same form where they take input data and produce output in a specified format. Here the output is a list with one component, output\$predict, the predicted phenotypes in the test set. For each method, we have to write a "wrapper" function to guarantee the input-output requirement. We also allow for additional arguments of each method (e.g. the tunning/user-specified parameters of the method). The wrapper functions (\*.wrapper.R) of all four methods are in folder methods/.

Now define a list of the methods that will run in our dsc. Each method is defined by its name (name), the function used to implement it (fn), and any additional arguments (args).

```
source_dir("methods") # need method wrappers

add_method(dsc_blm, name="bayes_lasso", fn=bayeslasso.wrapper, args=list(nIter=1.1e4,burnIn=1e3))
add_method(dsc_blm, name="gemma_bslmm", fn=gemma.bslmm.wrapper, args=list(w=1.1e4,s=1e3))
add_method(dsc_blm, name="rrblup", fn=rrblup.wrapper, args=list())
add_method(dsc_blm, name="rrblup_kernel", fn=rrblup.gkernel.wrapper, args=list())
```

To assess the prediction performance of each method, we define a score function; see score.R. The score function uses data (from datamaker.R) and output (from \*.wrapper.R) to compute a variety of prediction quality metrics, and save them in a list score. Currently, there are four metrics used in our score function:

- score\$mse: mean square error between predicted and observed phenotypes in the test set
- score\$rmse: root mean square error between predicted and observed phenotypes in the test set
- score\$pcor: Pearson correlation between predicted and observed phenotypes in the test set
- score\$slope: slope of simple regression of observed on predicted phenotypes in the test set

We have completed defining datamaker, methods and score for our dsc. Now we will run all methods on all the scenarios.

```
res=run_dsc(dsc_blm)
```

This returns a data frame res with the results of running all the methods on all scenarios.

```
> head(res[sample(1:80,replace = FALSE), 1:8])
     .id seed scenario
                              method
                                            mse
                                                     rmse
                                                               pcor
                                                                        slope
49 wheat
                 wheat
                              rrblup 0.7254980 0.8517617 0.5583381 1.1970995
42 wheat
            2
                 wheat
                              rrblup 0.8531229 0.9236465 0.5631131 1.2921754
34 wheat
                         gemma_bslmm 0.6087290 0.7802109 0.5353176 0.8697284
           14
                 wheat
                         bayes_lasso 0.7785023 0.8823278 0.5645790 1.1769306
4 wheat
           4
                 wheat
74 wheat
                 wheat rrblup kernel 0.8097204 0.8998447 0.6310907 0.9797745
           14
56 wheat
           16
                 wheat
                              rrblup 0.5883878 0.7670644 0.6604027 1.3740188
```

Now we can summarize the comparison results. For example, look at the median rmse and pcor.

```
> aggregate(rmse~method+scenario,res,median)
         method scenario
   bayes_lasso
                   wheat 0.8770236
1
2
    gemma_bslmm
                   wheat 0.8881155
                   wheat 0.8716839
         rrblup
4 rrblup kernel
                   wheat 0.9682903
> aggregate(pcor~method+scenario,res,median)
         method scenario
                              pcor
   bayes_lasso
                   wheat 0.5267583
1
2
    gemma_bslmm
                   wheat 0.5021476
         rrblup
3
                   wheat 0.5337660
4 rrblup_kernel
                   wheat 0.5947725
```

## Update the comparison

We call this framework "dynamic comparison" because one can add new data, new methods and new score metrics to the existing comparisons.

#### Add a dataset

In my current implementation, this part is not automatic. For each dataset, I first manually pre-process it such that the wrapper functions can use the data conviniently. For example, in the directory <code>/datamakers/wheat</code>, there are two R scripts to prepare the wheat data for comparison.

```
-rw-rw-r-- 1 xiangzhu xiangzhu 935 Sep 10 13:47 dataprep_gemma.R
-rw-rw-r-- 1 xiangzhu xiangzhu 356 Sep 10 13:47 dataprep_rtool.R
```

The function dataprep\_gemma.R generates \*.txt and \*.txt.gz that are used by gemma.bslmm.wrapper.R. The function dataprep\_rtool.R generates \*.RData that are used by methods implemented as R packages. This one-time step requires manual scripting, so it might be sub-optimal when there are many datasets for comparison. After this step, the dataset is added and can be used directly in the future.

#### Add a method

Write a new method wrapper, save it in the folder methods, add the new method to dsc\_blm, and run dsc. For example, I add a G-BLUP method (available in BGLR package) in our comparison.

```
source_dir("methods")
library(BGLR)
add_method(dsc_blm, name="g_blup", fn=gblup.wrapper, args=list(nIter=1.1e4,burnIn=1e3))
res=run_dsc(dsc_blm)
```

Now the new method and its results will show up in the data frame res.

#### Add a score

Simply modify the script score. R by adding a new score metric as an element of the list.