

Index Selection

Sweetpotato breeders Meeting, Malawi

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CIP



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Outline

- 1 Two selection indices
- 2 Computations
- 3 Example
- 4 Exercise
- 5 Conclusions

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- 2 Computations
- 3 Example
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Why to use a selection index?

- In breeding usually several traits have to be improved simultaneously.
- Very often intuitive procedures are used.
- Intuition usually fails when several correlated traits are involved:
 - Positive correlations: Improving one trait improves the other.
 - Negative correlations: Improving one trait can diminish the other.
- And do not forget the variability:
 - High variability: A lot of room to improve.
 - Low variability: No so much room to improve.

Two selection indices

- Elston, R. C. (1963). A weight-free index for the purpose of ranking or selection with respect to several traits at a time. *Biometrics*. 19(1): 85-97.
- Pesek, J. and R.J. Baker.(1969). Desired improvement in relation to selection indices. *Can. J. Plant. Sci.* 9:803-804.

Elston

The Elston index is a weight free index. Given p traits, for each genotype the Elston index is computed as

$$I_E = \prod_{i=1}^p (x_i - k_i)$$

where x_i is the value of the genotype for trait i and k_i is some lower bound. Two options for k :

- $k_i = \min x_i$
- $k_i = \frac{n \min x_i - \max x_i}{n-1}$.

Pesek-Baker

The Pesek Baker is an index where weights are given in the form of desired gains. Given p traits, this index is defined by

$$I_{PB} = \sum_{i=1}^p b_i x_i$$

where x_i is the value of the genotype for trait i . The coefficients, b_i , are computed from:

$$\mathbf{b} = \mathbf{V}^{-1}\mathbf{g}$$

with

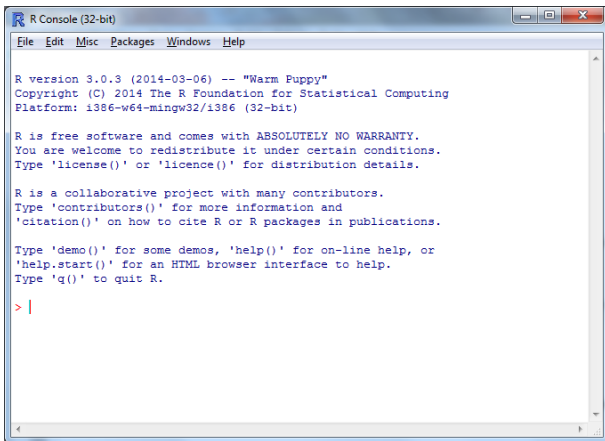
- \mathbf{b} the vector of index coefficients,
- \mathbf{V} the genetic variance-covariance matrix, and
- \mathbf{g} the vector of desired genetic gains to be specified by the breeder.

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Open R

We will use R for the example. If you have Clone Selector, you have R.
Open R.



```
R Console (32-bit)
File Edit Misc Packages Windows Help

R version 3.0.3 (2014-03-06) -- "Warm Puppy"
Copyright (C) 2014 The R Foundation for Statistical Computing
Platform: i386-w64-mingw32/i386 (32-bit)

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

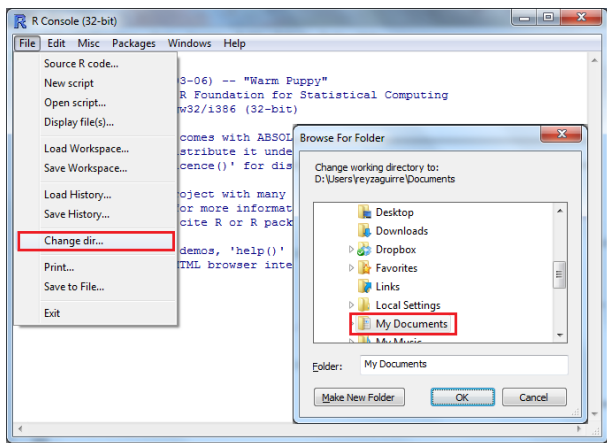
Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

> |
```

Define your working directory

Define your working directory.

My Documents by default.



GitHub account

`https://github.com/SweetPotatoImprov`

The screenshot shows the GitHub profile page for user Raul Eyzaguirre. At the top, there is a navigation bar with the GitHub logo, a search bar, and links for Explore, Gist, Blog, and Help. The user's profile header includes their name, a bio, and a profile picture of a sweet potato. Below the header, there are tabs for Contributions, Repositories, and Public activity, along with an Edit profile button. The main content area features a 'Popular repositories' section with the repository 'StatTools' listed. Below that is a 'Contributions' section with a heatmap showing activity from June to May. At the bottom, there are three summary cards: 'Year of contributions' (53 total), 'Longest streak' (5 days), and 'Current streak' (0 days).

Search or type a command

Explore Gist Blog Help

SweetPotatoImprov

Contributions Repositories Public activity Edit profile

Popular repositories

StatTools
R code for the analysis of field experiments data

Contributions

Summary of Pull Requests, issues opened, and commits. [Learn more.](#)

Year of contributions
53 total
Jun 11 2013 - Jun 11 2014

Longest streak
5 days
May 26 - May 30

Current streak
0 days
Rock - Hard Place

GitHub repo

<https://github.com/SweetPotatoImprov/StatTools>

SweetPotatoImprov / StatTools

Unwatch 2 Star 0 Fork 0

R code for the analysis of field experiments data — Edit

55 commits 1 branch 0 releases 1 contributor

branch: master StatTools / +

edited

SweetPotatoImprov authored on May 30 latest commit 35176a46ee

AMMI	as.character deleted	16 days ago
CheckConsis	Update CheckConsis.R	14 days ago
CheckData	typo	22 days ago
GoodDataPractices	edited	12 days ago
IndexSelection	as.character deleted	16 days ago
MissValEst	as.character deleted	16 days ago
Resp2Selection	README file	21 days ago
Tai	as.character deleted	16 days ago
README.md	Good data practices included	14 days ago

README.md

Code

- Issues 0
- Pull Requests 0
- Wiki
- Pulse
- Graphs
- Network
- Settings

HTTPS clone URL

<https://github.com/SweetPotatoImprov/StatTools>

You can clone with [HTTPS](#), [SSH](#), or [Subversion](#).

Clone in Desktop

Download ZIP

Two selection indices

The screenshot shows a GitHub repository page for 'SweetPotatoImprov / StatTools'. At the top, there are interaction buttons: 'Unwatch' (with a dropdown arrow), '2' (indicating 2 watchers), 'Star' (with a dropdown arrow), '0' (indicating 0 stars), and 'Fork' (with a dropdown arrow), '0' (indicating 0 forks). Below this, the repository path is 'branch: master' and 'StatTools / IndexSelection / +'. A file list table is displayed with the following entries:

File Name	Type	Time
data	data	4 hours ago
Elston	edited	8 hours ago
PesekBaker	edited	8 hours ago
README.md	new folder	a month ago

The 'Elston' and 'PesekBaker' entries are highlighted with a red box. Below the file list is a preview of the 'README.md' file, which contains the following text:

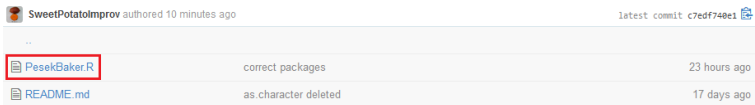
IndexSelection

Proposed Selection Indices for Applied Breeding Programs.

Download the files

Download the R functions (Elston.R and PesekBaker.R) into your working directory.

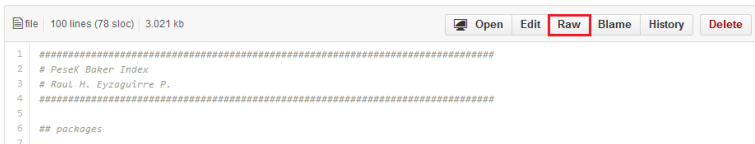
- Click on the function.



SweetPotatoImprov authored 10 minutes ago latest commit c7edf740e1

...		
PesekBaker.R	correct packages	23 hours ago
README.md	as.character deleted	17 days ago

- Click on Raw.



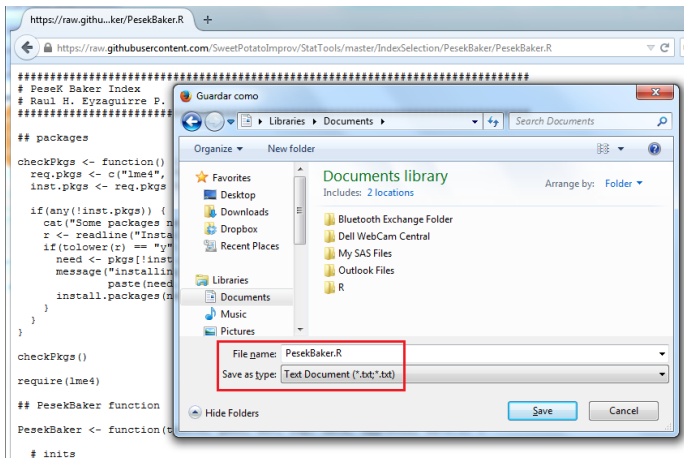
file | 100 lines (78 sloc) | 3.021 kb

Open Edit **Raw** Blame History Delete

```
1 #####  
2 # Pesek Baker Index  
3 # Raul H. Eyzaguirre P.  
4 #####  
5  
6 ## packages  
7
```

Download the files

- Right click on the mouse and *Save as ...* or go to *Save web page as ...* Beware of browsers, some are not so smart. Make sure to save the file with the real name (Elston.R or PesekBaker.R)



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Data

Data is in the file SI_example.csv.

This file contains data for:

- 2 locations: La Molina and Satipo.
- 2 replications in each location.
- 1041 genotypes.
- 25 traits.
- 3 plants per plot.

Small data

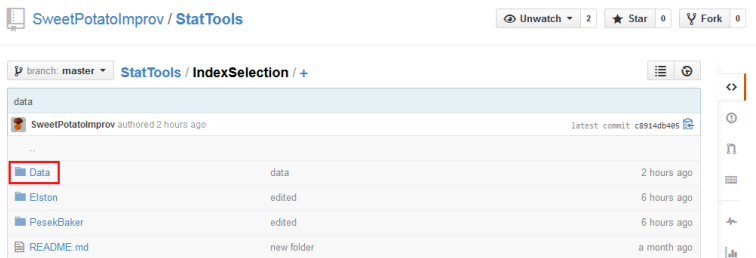
Let us start with a small subset in the file `SI_example_small.csv`.

This file contains data for:

- 2 locations: La Molina and Satipo.
- 2 replications in each location.
- 8 genotypes.
- 5 traits: RYTHA, BC, DM, STAR, NOCR.
- 3 plants per plot.

Download the data

Data are on GitHub, so you can proceed in the same way as we did to download the functions.



Screenshot of a GitHub repository page for `SweetPotatoImprov / StatTools`. The page shows the file list for the `data` directory. The file `Data` is highlighted with a red box. Other files include `Elston`, `PesekBaker`, and `README.md`.

File Name	Author	Time Ago
data	SweetPotatoImprov	2 hours ago
..		
Data		2 hours ago
Elston		6 hours ago
PesekBaker		6 hours ago
README.md		a month ago

Download the data

- Click on the file.

SweetPotatoImprov authored 3 minutes ago latest commit 6440184e39

SI_example_small.csv data 4 hours ago

- Click on Raw.

file 34 lines (33 sloc) 1.845 kb

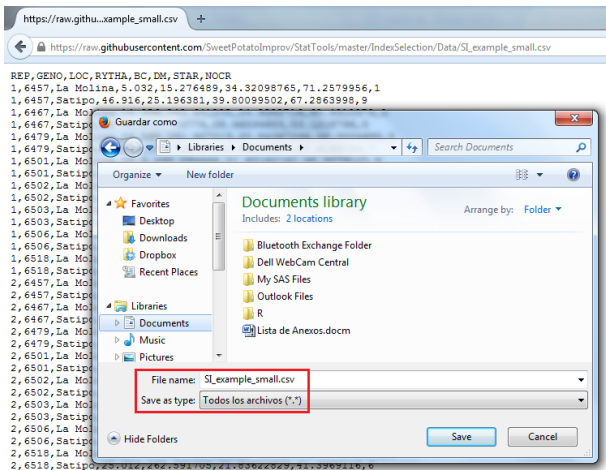
Open Edit **Raw** Blame History Delete

Search this file...

	REP	GENO	LOC	RYTHA	BC	DM	STAR	NOCR
1	1	6457	La Molina	5.032	15.276489	34.32098765	71.2579956	1
2	1	6457	Satipo	46.916	25.196381	39.80099502	67.2863998	9
3	1	6467	La Molina	14.356	243.641205	24.9382716	60.4310379	2
4	1	6467	Satipo	26.64	201.799774	28.46534653	53.1218796	5
5	1	6479	La Molina	24.568	151.407013	20.84367246	58.3554688	3
6	1	6479	Satipo	37.296	104.716797	25.86206897	57.6193733	7
7	1	6501	La Molina	29.6	498.594666	21.62162162	48.3273125	8

Download the data

- Right click on the mouse and *Save as ...* or go to *Save web page as ...* Make sure to save the file with the real name and format.



Load the data directly from GitHub

You can also load the data directly from GitHub into R.

This works well if you use RStudio:

```
> urlfile <- 'https://raw.githubusercontent.com/SweetPotatoImprov/StatTools/  
+           master/IndexSelection/Data/SI_example_small.csv'  
> mydata <- read.csv(urlfile)
```

This works well directly from R:

```
> require(RCurl)  
> mydata <- getURL(urlfile, ssl.verifypeer = FALSE)  
> mydata <- read.csv(textConnection(mydata))
```

Load the data and functions

```
> mydata <- read.csv("SI_example_small.csv")
> mydata$REP <- factor(mydata$REP)
> mydata$GENO <- factor(mydata$GENO)
> str(mydata)

'data.frame':      32 obs. of  8 variables:
 $ REP  : Factor w/ 2 levels "1","2": 1 1 1 1 1 1 1 1 1 1 ...
 $ GENO : Factor w/ 8 levels "6457","6467",...: 1 1 2 2 3 3 4 4 5 5 ...
 $ LOC  : Factor w/ 2 levels "La Molina","Satipo": 1 2 1 2 1 2 1 2 1 2 ...
 $ RYTHA: num  5.03 46.92 14.36 26.64 24.57 ...
 $ BC   : num  15.3 25.2 243.6 201.8 151.4 ...
 $ DM   : num  34.3 39.8 24.9 28.5 20.8 ...
 $ STAR : num  71.3 67.3 60.4 53.1 58.4 ...
 $ NOCR : int   1 9 2 5 3 7 8 11 6 7 ...

> source('Elston.R')
> source('PesekBaker.R')
```

Elston index

```
> Elston(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', mydata)
$Elston.Index
  6457   6467   6479   6501   6502   6503   6506   6518
0.09313 0.00000 1.18126 4.82440 33.84316 0.28160 0.00000 0.00000

$Sorted.Elston.Index
  6502   6501   6479   6503   6457   6467   6506   6518
33.84316 4.82440 1.18126 0.28160 0.09313 0.00000 0.00000 0.00000
```

Remember that by default it uses $k_i = \min x_i$ for the index

$$I_E = \prod_{i=1}^P (x_i - k_i)$$

Elston index

We can use the second lower bound $k_i = \frac{n \min x_j - \max x_j}{n-1}$.

```
> Elston(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', lb=2, mydata)
```

```
$Elston.Index
```

6457	6467	6479	6501	6502	6503	6506	6518
4.862	0.710	6.790	19.822	84.154	11.965	18.425	3.439

```
$Sorted.Elston.Index
```

6502	6501	6506	6503	6479	6457	6518	6467
84.154	19.822	18.425	11.965	6.790	4.862	3.439	0.710

Pesek Baker index

Let us compute the Pesek Baker index using these traits (means between brackets):

- RYTHA (31.3 tons/ha)
- BC (188 ppm)
- DM (30.05%)
- STAR (60.3%)
- NOCR (6.66)

Which are your desired genetic gains?

My guess: 5, 100, 2, 2, 5.

Compute the index

```
PesekBaker(traits, geno, env, rep, data, dgg=NULL, sf=0.1)
```

- `traits` is a list of traits,
- `geno` is the index for genotypes,
- `env` is the index for environments,
- `rep` is the index for replications,
- `data` is the data frame containing the data,
- `dgg` is the vector of desired genetic gains, standard deviations by default, and
- `sf` is the selected fraction, 0.1 by default.

```
> output <- PesekBaker(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', 'LOC', 'REP',  
+ mydata, c(5, 100, 2, 2, 5))
```

Results

The PesekBaker function returns the following elements:

- `$Genetic.Variance`: The estimated genetic variances.
- `$Correlation.Matrix`: The estimated correlation matrix.
- `$Covariance.Matrix`: The estimated covariance matrix.
- `$Index.Coefficients`: The index coefficients.
- `$Std.Response.to.Selection`: The standardized response to selection.
- `$Response.to.Selection`: The response to selection.
- `$Pesek.Baker.Index`: The Pesek-Baker index value.
- `$Sorted.Pesek.Baker`: The Pesek-Baker index value sorted in descending order.

Index coefficients

```
> output$Index.Coefficients
```

	coef
RYTHA	-0.29365
BC	0.01197
DM	0.21252
STAR	0.01919
NOCR	2.24846

Why is the RYTHA coefficient negative?

- It does not mean that we are going to select to diminish root yield.
- We must see the correlations.
- We must see the variances.

See the correlations

We want to improve all the traits, but some have a negative correlation.

There are also traits with high positive correlations.

It is difficult to have this kind of correlation structures on mind.

```
> output$Correlation.Matrix
```

	RYTHA	BC	DM	STAR	NOCR
RYTHA	1.0000	0.26732	-0.231076	-0.15266	0.615558
BC	0.2673	1.00000	-0.775152	-0.85198	0.091521
DM	-0.2311	-0.77515	1.000000	0.89555	0.004778
STAR	-0.1527	-0.85198	0.895548	1.00000	0.077349
NOCR	0.6156	0.09152	0.004778	0.07735	1.000000

See the standard deviations

Maybe we want to improve too much a trait with a very low variability.

If a trait has high variability, then there is a lot of room to improve.

Let us remember my desired gains: 5, 100, 2, 2, 5.

```
> output$Genetic.Variance^0.5
[1] 5.068 151.583 5.859 8.231 1.663
> output$Response.to.Selection
[1] 1.765 35.298 0.706 0.706 1.765
> output$Std.Response.to.Selection
[1] 0.34827 0.23286 0.12049 0.08577 1.06140
```

Compute the index - second try

I will relax a little bit my ambition about NOCR.

```
> output <- PesekBaker(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', 'LOC', 'REP',
+                       mydata, c(5, 100, 2, 2, 2.5))
> output$Index.Coefficients

      coef
RYTHA 0.01853
BC     0.01669
DM     0.20259
STAR  0.15350
NOCR  0.66803

> output$Response.to.Selection

[1] 3.605 72.102 1.442 1.442 1.803

> output$Std.Response.to.Selection

[1] 0.7114 0.4757 0.2461 0.1752 1.0841
```


Using the defaults

By default the function gives weights so that the desired genetic gains are one standard deviation for each trait. Same relative weight for each trait.

```
> output <- PesekBaker(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', 'LOC', 'REP',
+                       mydata)
> output$Index.Coefficients

      coef
RYTHA 0.12673
BC     0.04457
DM     0.18665
STAR  0.71823
NOCR  -0.28636

> output$Response.to.Selection

[1] 1.8160 54.3195 2.0995 2.9495 0.5959

> output$Std.Response.to.Selection

[1] 0.3583 0.3583 0.3583 0.3583 0.3583
```

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Load the data

Load the complete data set.

```
> mydata <- read.csv("SI_example.csv")
> mydata$REP <- factor(mydata$REP)
> mydata$GENO <- factor(mydata$GENO)
> names(mydata)

 [1] "REP"   "GENO"  "LOC"   "PROT"  "BC"    "FE"    "ZN"    "CA"    "MG"
[10] "STAR"  "FRUC"  "GLUC"  "SUCR"  "MALT"  "DM"    "VW"    "NOPS"  "NOCR"
[19] "NONC"  "CRW"   "NCRW"  "RYTHA" "TRW"   "BIOM"  "HI"    "CI"    "FYTHA"
[28] "CYTHA"
```

Choose a set of traits to improve

Choose a set of traits to improve and run Pesek-Baker using defaults. See the correlations, standard deviations and response to selection.

```
> output <- PesekBaker(c('RYTHA', 'BC', 'DM', 'STAR', 'NOCR'), 'GENO', 'LOC', 'REP',  
+                       mydata)  
> output$Correlation.Matrix  
  
      RYTHA      BC      DM      STAR      NOCR  
RYTHA 1.00000 0.06816 -0.1677 -0.0496 0.8300  
BC     0.06816 1.00000 -0.5271 -0.6703 0.1144  
DM     -0.16766 -0.52711 1.0000 0.8277 -0.1710  
STAR   -0.04960 -0.67032 0.8277 1.0000 -0.1026  
NOCR   0.83003 0.11437 -0.1710 -0.1026 1.0000  
  
> output$Genetic.Variance^0.5  
[1] 5.626 145.822 4.631 7.170 1.394  
  
> output$Response.to.Selection  
[1] 3.1484 81.5980 2.5916 4.0124 0.7798  
  
> output$Std.Response.to.Selection  
[1] 0.5596 0.5596 0.5596 0.5596 0.5596
```

Get your index

With an eye on the correlations and standard deviations, define your own desired genetic gains. Run the index again.

See the response to selection. Not satisfied? Modify your desired genetic gains and run the index again.

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Conclusions for the Pesek Baker index

- Keep an eye in the correlations.
- Keep both eyes in the standard deviations.
- Specify your desired genetic gains in relative terms,
- and even better, specify your desired genetic gains in relative terms and in standard deviation units.
- If a trait is important give it a weight a little bit higher than one standard deviation (somewhere between 1 and 2). If a trait is not so important give it a weight lower than 1 standard deviation (between 0 and 1).
- Do not ever try to improve a trait in more than 2 standard deviations (or maybe 1.5 could be a better and more conservative upper bound).
- Play with different values for the desired genetic gains and compare the response to selection that you get with each group of values.