
data_tools Documentation

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Collection of Python functions and classes designed to make a Computational Biologist's life easier.

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DEPENDENCIES

In module **data_tools.plots**:

- NumPy
- Matplotlib
- Pandas

In module **data_tools.Lasso**:

- NumPy
- Matplotlib
- Pandas
- Scikit-learn

INSTALLATION

First download/clone `data_tools` from the [GitHub repository](#). From the terminal:

```
git clone https://github.com/Nic-Nic/data_tools.git
cd data_tools
```

Then you can install it by running `setup.py` as follows:

```
python setup.py sdist
```

Or using `pip`:

```
pip install .
```


REFERENCE

3.1 `data_tools.Lasso`

Class for logistic regression models with L1 regularization.

```
class data_tools.Lasso.Lasso(Cs=array([ 0.01, 0.01009272, 0.01018629, 0.01028073,
0.01037605, 0.01047225, 0.01056935, 0.01066734, 0.01076624,
0.01086606, 0.01096681, 0.01106848, 0.01117111, 0.01127468,
0.01137921, 0.01148472, 0.0115912, 0.01169866, 0.01180713,
0.0119166, 0.01202708, 0.01213859, 0.01225114, 0.01236472,
0.01247936, 0.01259506, 0.01271184, 0.0128297, 0.01294865,
0.0130687, 0.01318987, 0.01331216, 0.01343558, 0.01356015,
0.01368587, 0.01381276, 0.01394083, 0.01407008, 0.01420053,
0.01433219, 0.01446507, 0.01459919, 0.01473454, 0.01487115,
0.01500903, 0.01514819, 0.01528864, 0.01543038, 0.01557345,
0.01571784, 0.01586357, 0.01601064, 0.01615909, 0.01630891,
0.01646012, 0.01661273, 0.01676675, 0.0169222, 0.0170791,
0.01723745, 0.01739726, 0.01755856, 0.01772136, 0.01788566,
0.01805149, 0.01821885, 0.01838777, 0.01855825, 0.01873032,
0.01890397, 0.01907924, 0.01925614, 0.01943467, 0.01961486,
0.01979672, 0.01998026, 0.02016551, 0.02035248, 0.02054117,
0.02073162, 0.02092383, 0.02111783, 0.02131362, 0.02151123,
0.02171068, 0.02191197, 0.02211512, 0.02232016, 0.02252711,
0.02273597, 0.02294676, 0.02315951, 0.02337424, 0.02359095,
0.02380968, 0.02403043, 0.02425323, 0.02447809, 0.02470504,
0.02493409, 0.02516527, 0.02539859, 0.02563407, 0.02587174,
0.02611161, 0.02635371, 0.02659804, 0.02684465, 0.02709354,
0.02734474, 0.02759826, 0.02785414, 0.02811239, 0.02837304,
0.0286361, 0.0289016, 0.02916956, 0.02944, 0.02971296,
0.02998844, 0.03026648, 0.0305471, 0.03083031, 0.03111616,
0.03140465, 0.03169582, 0.03198969, 0.03228628, 0.03258562,
0.03288774, 0.03319266, 0.03350041, 0.03381101, 0.03412449,
0.03444087, 0.03476019, 0.03508247, 0.03540774, 0.03573602,
0.03606735, 0.03640175, 0.03673925, 0.03707988, 0.03742366,
0.03777064, 0.03812083, 0.03847427, 0.03883098, 0.039191,
0.03955436, 0.03992109, 0.04029122, 0.04066478, 0.04104181,
0.04142232, 0.04180637, 0.04219398, 0.04258518, 0.04298001,
0.0433785, 0.04378069, 0.0441866, 0.04459628, 0.04500975,
0.04542706, 0.04584824, 0.04627332, 0.04670234, 0.04713535,
0.04757236, 0.04801343, 0.04845859, 0.04890787, 0.04936132,
0.04981898, 0.05028087, 0.05074705, 0.05121755, 0.05169242,
0.05217169, 0.0526554, 0.05314359, 0.05363631, 0.0541336,
0.0546355, 0.05514206, 0.05565331, 0.0561693, 0.05669007,
0.05721568, 0.05774615, 0.05828155, 0.0588219, 0.05936727,
0.0599177, 0.06047322, 0.0610339, 0.06159978, 0.0621709,
0.06274732, 0.06332908, 0.06391624, 0.06450884, 0.06510694,
0.06571058, 0.06631981, 0.0669347, 0.06755528, 0.06818162,
0.06881377, 0.06945178, 0.0700957, 0.07074559, 0.07140151,
0.07206351, 0.07273165, 0.07340598, 0.07408657, 0.07477346,
0.07546673, 0.07616642, 0.0768726, 0.07758532, 0.07830465,
0.07903066, 0.07976339, 0.08050292, 0.0812493, 0.08200261,
0.08276289, 0.08353023, 0.08430468, 0.08508632, 0.08587519,
0.08667139, 0.08747496, 0.08828599, 0.08910453, 0.08993067,
0.09076446, 0.09160598, 0.09245531, 0.09331251, 0.09417766,
0.09505083, 0.09593209, 0.09682153, 0.09771921, 0.09862522,
0.09953962, 0.10046251, 0.10139395, 0.10233402, 0.10328281,
0.1042404, 0.10520687, 0.10618229, 0.10716676, 0.10816036,
0.10916317, 0.11017528, 0.11119677, 0.11222774, 0.11326826,
0.11431843, 0.11537833, 0.11644806, 0.11752771, 0.11861737,
0.11971713, 0.12082709, 0.12194734, 0.12307798, 0.1242191,
0.1253708, 0.12653318, 0.12770633, 0.12889036, 0.13008537,
0.13129146, 0.13250873, 0.13373729, 0.13497706, 0.13622816,
0.13749172, 0.13876648, 0.14005305, 0.14135156, 0.1426621,
0.1439848, 0.14531975, 0.14666709, 0.14802691, 0.14939935,
0.1507845, 0.15218251, 0.15359347, 0.15501751, 0.15645476,
```

Statistical model developer. Uses LASSO Logistic regression with cross validation (CV) for penalty parameter “C”. Inherits from `sklearn.linear_model.LogisticRegressionCV`.

compute_second_line()

Method that computes second-line predictors. This is, initially selected predictors’ data is dropped out, and another model is trained with the remaining data.

fit_data(*x*, *y*, *silent=False*)

Fits the passed data, assumed to follow pandas data structures [`pandas.DataFrame`] or [`pandas.Series`]. Where *x* is the normalized MS intensity [float] and *y* the disease status category [int].

plot_coef()

Plots the non-zero coefficients for the fitted predictors.

plot_sample_probs(*x=None*, *y=None*)

plot_score()

Plots the mean score across all folds obtained during CV. Highlights the optimum C parameter chosen (NOTE: Since C represents the inverse regularization parameter, it’s chosen so that maximizes the score).

3.2 data_tools.plots

Plotting functions module.

`data_tools.plots.volcano`(*logfc*, *logpval*, *thr_pval=0.05*, *thr_fc=2.0*, *c=('C0', 'C1')*, *legend=True*, *title=None*, *filename=None*, *figsize=None*)

Generates a volcano plot from the differential expression data provided.

- **Arguments:**

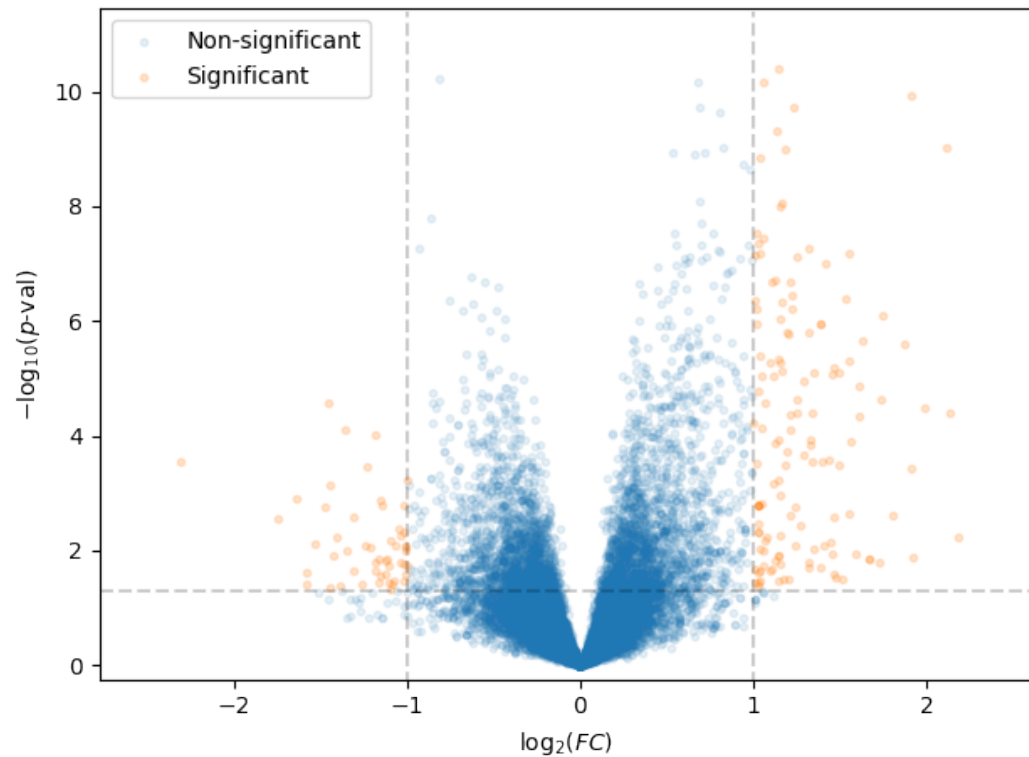
- *logfc* [list]: Or any iterable type. Contains the log (usually base 2) fold-change values. Must have the same length as *logpval*.
- *logpval* [list]: Or any iterable type. Contains the -log p-values (usually base 10). Must have the same length as *logfc*.
- *thr_pval* [float]: Optional, 0.05 by default. Specifies the p-value (non log-transformed) threshold to consider a measurement as significantly differentially expressed.
- *thr_fc* [float]: Optional, 2. by default. Specifies the FC (non log-transformed) threshold to consider a measurement as significantly differentially expressed.
- *c* [tuple]: Optional, ('C0', 'C1') by default (matplotlib default colors). Any iterable containing two color arguments tolerated by matplotlib (e.g.: ['r', 'b'] for red and blue). First one is used for non-significant points, second for the significant ones.
- *legend* [bool]: Optional, True by default. Indicates whether to show the plot legend or not.
- *title* [str]: Optional, None by default. Defines the plot title.
- *filename* [str]: Optional, None by default. If passed, indicates the file name or path where to store the figure. Format must be specified (e.g.: .png, .pdf, etc)
- *figsize* [tuple]: Optional, None by default (default matplotlib size). Any iterable containing two values denoting the figure size (in inches) as [width, height].

- **Returns:**

- [`matplotlib.figure.Figure`]: Figure object containing the volcano plot.

- **Examples:**

```
>>> volcano(my_log_fc, my_log_pval)
```



`data_tools.plots.piano_consensus` (*df*, *nchar*=40, *boxes*=True, *title*=None, *filename*=None, *figsize*=None)

Generates a GSEA consensus score plot like R package `piano`'s `consensusScores` function, but prettier. The main input is assumed to be a `pandas.DataFrame` whose data is the same as the `rankMat` from the result of `consensusScores`.

- **Arguments:**

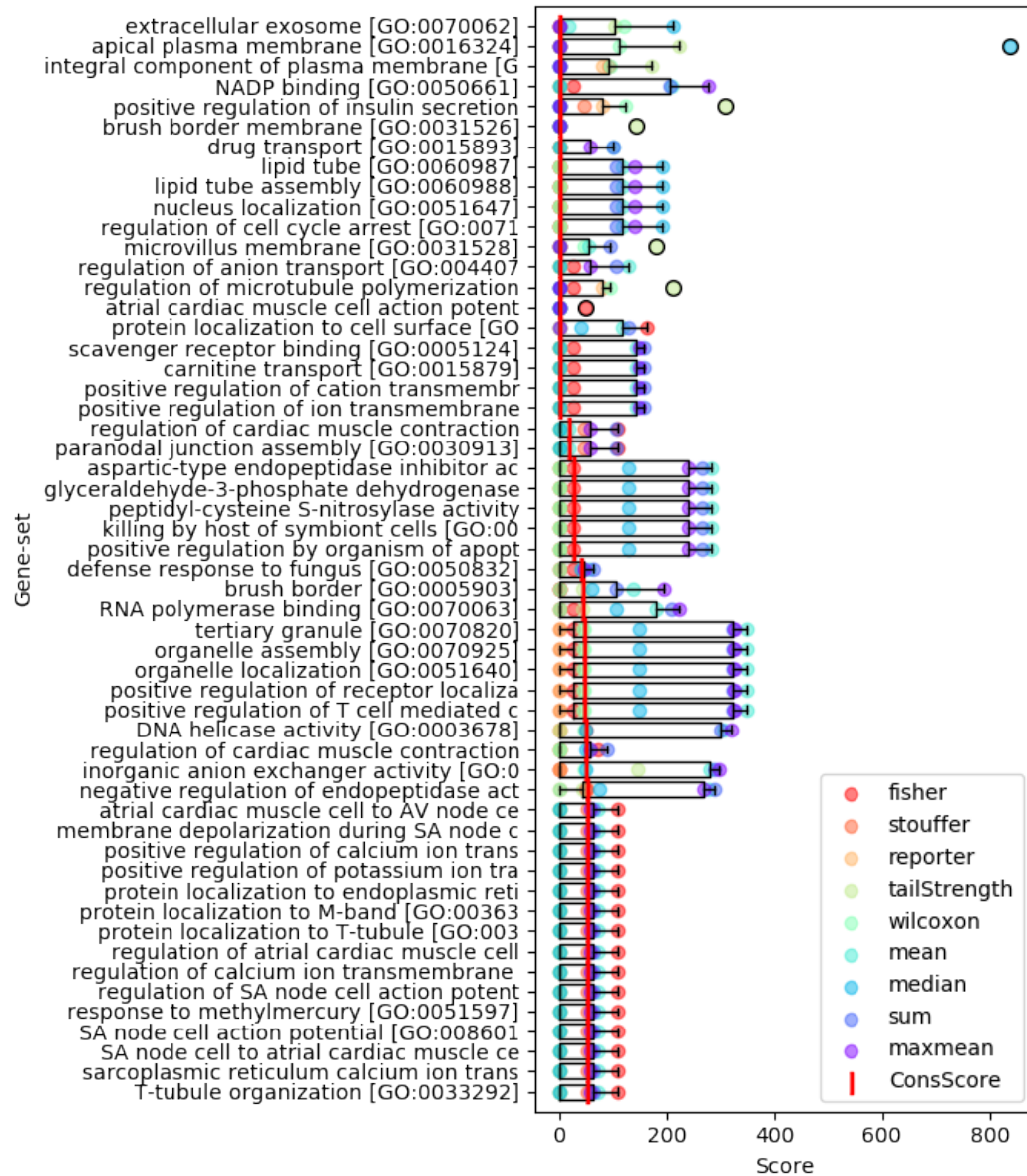
- *df* [`pandas.DataFrame`]: Values contained correspond to the scores of the gene-sets (consensus and each individual methods). Index must contain the gene-set labels. Columns are assumed to be `ConsRank` (ignored), `ConsScore` followed by the individual methods (e.g.: `mean`, `median`, `sum`, etc).
- *nchar* [`int`]: Optional, 40 by default. Number of string characters of the gene-set labels of the plot.
- *boxes* [`bool`]: Optional, True by default. Determines whether to show the boxplots of the gene-sets or not.
- *title* [`str`]: Optional, None by default. Defines the plot title.
- *filename* [`str`]: Optional, None by default. If passed, indicates the file name or path where to store the figure. Format must be specified (e.g.: `.png`, `.pdf`, etc)
- *figsize* [`tuple`]: Optional, None by default (default matplotlib size). Any iterable containing two values denoting the figure size (in inches) as [`width`, `height`].

- **Returns:**

- `[matplotlib.figure.Figure]`: the figure object containing a combination of box and scatter plots of the gene-set scores.

- **Examples:**

```
>>> piano_consensus(df, figsize=[7, 8])
```



3.3 data_tools.sets

Set operations module.

`data_tools.sets.in_all(x, N)`

Checks if a vector x is present in all sets contained in a list N .

- **Arguments:**

- x [tuple]: Or any hashable type as long as is the same contained in the sets of N .
- N [list]: Or any iterable type containing [set] objects.

- **Returns:**

- [bool]: True if x is found in all sets of N , False otherwise.

- **Examples:**

```
>>> N = [{(0, 0), (0, 1)}, # <- set A
...      {(0, 0), (1, 1), (1, 0)}] # <- set B
>>> x = (0, 0)
>>> in_all(x, N)
True
>>> y = (0, 1)
>>> in_all(y, N)
False
```

`data_tools.sets.bit_or(a, b)`

Returns the bit operation OR between two bit-strings a and b . NOTE: a and b must have the same size.

- **Arguments:**

- a [tuple]: Or any iterable type.
- b [tuple]: Or any iterable type.

- **Returns:**

- [tuple]: OR operation between a and b element-wise.

- **Examples:**

```
>>> a, b = (0, 0, 1), (1, 0, 1)
>>> bit_or(a, b)
(1, 0, 1)
```

`data_tools.sets.multi_union(N)`

Returns the union set of all sets contained in a list N .

- **Arguments:**

- N [list]: Or any iterable type containing [set] objects.

- **Returns:**

- [set]: The union of all sets contained in N .

- **Examples:**

```
>>> A = {1, 3, 5}
>>> B = {0, 1, 2}
>>> C = {0, 2, 5}
>>> multi_union([A, B, C])
set([0, 1, 2, 3, 5])
```

`data_tools.sets.find_min(A)`

Finds and returns the subset of vectors whose sum is minimum from a given set A .

- **Arguments:**

- A [set]: Set of vectors ([tuple] or any iterable).

- **Returns:**

- [set]: Subset of vectors in *A* whose sum is minimum.

- **Examples:**

```
>>> A = {(0, 1, 1), (0, 1, 0), (1, 0, 0), (1, 1, 1)}
>>> find_min(A)
set([(0, 1, 0), (1, 0, 0)])
```

3.4 data_tools.strings

String operations module.

`data_tools.strings.is_numeric(s)`

Determines if a string can be considered a numeric value. NaN is also considered, since it is float type.

- **Arguments:**

- *s* [str]: String to be evaluated.

- **Returns:**

- [bool]: True/False depending if the condition is satisfied.

- **Examples:**

```
>>> is_numeric('4')
True
>>> is_numeric('-3.2')
True
>>> is_numeric('number')
False
>>> is_numeric('NaN')
True
```

`data_tools.strings.join_str_lists(a, b, sep=")`

Joins element-wise two lists (or any 1D iterable) of strings with a given separator (if provided). Length of the input lists must be equal.

- **Arguments:**

- *a* [list]: Contains the first elements [str] of the joint strings.
- *b* [list]: Contains the second elements [str] of the joint strings.
- *sep* [str]: Optional ' ' (non separated) by default. Determines the separator between the joint strings.

- **Returns:**

- [list]: List of the joint strings.

- **Example:**

```
>>> a = ['a', 'b']
>>> b = ['1', '2']
>>> join_str_lists(a, b, sep='_')
['a_1', 'b_2']
```


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