
si4ul

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SI4UL.CHANGEPOINT_SI MODULE

`si4ul.changepoint_si.optseg_si(x, sigma=1, beta=1.5)`
computing valid p-values for optimal changepoint by Selective Inference using Dynamic Programming.

Parameters

- **x** (*array-like of shape (point-num)*) – time series data.
- **sigma** (*float*) – standard deviation in distribution of time series data.
- **beta** (*float*) – regularization factor.

Returns changepoint list and p-values of each changepoint.

Return type (array-like of shape (changepoint-num), array-like of shape (changepoint-num))

Examples

```
>>> seg, p_list = changepoint_si.optseg_si(x)
>>> print(seg)
[25, 40, 59, 80, 100, 125, 140, 159, 180]
>>> print(p_list)
[6.439051311970013e-06, 5.451462826100586e-07, 3.589542368255502e-07, 7.
↪ 74132481356728e-05, 0.11719543493338598, 6.4387982662212394e-06, 1.
↪ 1776732917248327e-06, 3.589542368255725e-07, 1.8180641069769048e-09]
```

`si4ul.changepoint_si.optseg_si_oc(x, sigma=1, beta=1.5)`
computing valid p-values for optimal changepoint by over-comconditioning Selective Inference.

Parameters

- **x** (*array-like of shape (point-num)*) – time series data.
- **sigma** (*float*) – standard deviation in distribution of time series data.
- **beta** (*float*) – regularization factor.

Returns changepoint list and p-values of each changepoint.

Return type (array-like of shape (changepoint-num), array-like of shape (changepoint-num))

Examples

```
>>> seg, p_list = changepoint_si.optseg_si(x)
>>> print(seg)
[25, 40, 59, 75, 80, 100, 125, 140, 159, 175, 180]
>>> print(p_list)
[0.013528694810211626, 0.05567852225768205, 0.329173206633249, 0.36014111520463954, ↵
↵0.4101971284236369, 0.6981932657047393, 0.16996861755494708, 0.3719537702838958, ↵
↵0.3592986205789473, 0.36014111520464676, 0.1196678816881915]
```

```
si4ul.changepoint_si.plot_changepoint_detection(x, sg_results, p_value_list, alpha, underlying=None,
                                                segment_size=0, title='OptSeg-SI')
```

the result of optimal changepoint after SI.

Parameters

- **x** (*array-like of shape (point-num)*) – time series data.
- **sg_results** (*array-like of shape (changepoint-num)*) – changepoint list.
- **p_value_list** (*array-like of shape (changepoint-num)*) – p-values of each changepoint.
- **alpha** (*float*) – significance level.
- **underlying** (*array-like of shape (true changepoint-num)*) – mean values of true segment.
- **segment_size** (*float*) – true segment size.
- **title** (*string*) – plot title.

SI4UL.KMEANS_SI MODULE

`si4ul.kmeans_si.all_clusters_combination_test(obs_model, test_gene=None, sigma=1.0)`

post clustering inference for all cluster combinations. If `test_gene` is set, test is `PCI_gene`. otherwise, test is `PCI_cluster`.

Parameters

- **obs_model** (`KMeans`) – refer to document of `KMeans`.
- **test_gene** (`int`) – feature to compare.
- **sigma** (`float`) – standard deviation in distribution.

Returns matrix of test statistics, matrix of homotopy `PCI` p-value and matrix of naive p-value.

Return type (array-like of shape(3, cluster_num, cluster_num))

Examples

```
>>> print(kmeans_si.all_clusters_combination_test(obs_model))
pci_cluster
(array([[nan, 5.24520442, 4.53785448], [nan, nan, 5.38552569], [nan, nan, nan]]),
array([[nan, 0.01863257, 0.00028306], [nan, nan, 0.00359755], [nan, nan, nan]]),
array([[nan, 1.06122705e-06, 3.37658148e-05], [nan, nan, 5.03368420e-07], [nan, nan,
↪ nan]]))
```

`si4ul.kmeans_si.kmeans(X, n_clusters)`

k-means clustering algorithm.

Parameters

- **X** (array-like of shape (n, d)) – data matrix.
- **n_clusters** (`int`) – number of cluster.

Returns refer to document of `KMeans`.

Return type `KMeans`

Examples

```
>>> print(kmeans_si.kmeans(X, K))
<si4ul.si.kmeans_si.KMeans at 0x102aef700>
```

```
si4ul.kmeans_si.pci_cluster(obs_model, comparison_clusters, sigma=1.0, max_iter=1000, random_seed=0,
                             z_max=20)
```

post clustering inference for test between clusters.

Parameters

- **obs_model** ([KMeans](#)) – refer to document of KMeans.
- **comparison_clusters** (*array-like of shape(2)*) – set of clusters to compare.
- **sigma** (*float*) – standard deviation in distribution.
- **max_iter** (*int*) – upper limit count of iteration in k-means algorithm.
- **random_seed** (*int*) – seed of random for determine initial cluster.
- **z_max** (*float*) – upper limit of parameter z on test statistics vector.

Returns test statistics, homotopy PCI p-value and naive p-value.

Return type (float, float, float)

Examples

```
>>> print(kmeans_si.pci_cluster(obs_model, comparison_clusters))
(5.245204424402314, 0.018632573868904267, 1.0612270479959528e-06)
```

```
si4ul.kmeans_si.pci_gene(obs_model, comparison_clusters, test_gene, sigma=1.0, max_iter=1000,
                          random_seed=0, z_max=20)
```

post clustering inference for test between clusters about a feature.

Parameters

- **obs_model** ([KMeans](#)) – refer to document of KMeans.
- **comparison_clusters** (*array-like of shape(2)*) – set of clusters to compare.
- **sigma** (*float*) – standard deviation in distribution.
- **test_gene** (*int*) – feature to compare.
- **max_iter** (*int*) – upper limit count of iteration in k-means algorithm.
- **random_seed** (*int*) – seed of random for determine initial cluster.
- **z_max** (*float*) – upper limit of parameter z on test statistics vector.

Returns test statistics, homotopy PCI p-value and naive p-value.

Return type (float, float, float)

Examples

```
>>> print(kmeans_si.pci_gene(obs_model, comparison_clusters, gene_id))
(0.26352301450242993, 0.4212956294190716, 0.20005529456703786)
```

`si4ul.kmeans_si.plot_histogram(obs_model, comparison_clusters, test_gene, is_plot_norm=False)`
 plot histogram of distribution per cluster using test.

Parameters

- **obs_model** ([KMeans](#)) – refer to document of KMeans.
- **comparison_clusters** (*array-like of shape(2)*) – set of clusters to compare.
- **test_gene** (*int*) – feature to plot.
- **is_plot_norm** (*bool*) – whether plot normal distribution in background.

`si4ul.kmeans_si.plot_p_matrix(matrix, digit=3, alpha=0.05)`
 plot matrix of p-value that is calculated by each cluster combinations.

Parameters

- **matrix** (*array-like of shape(cluster_num, cluster_num)*) – matrix of p-value.
- **digit** (*int*) – digit number to display.
- **alpha** (*float*) – significant level.

`si4ul.kmeans_si.plot_scatter(obs_model, comparison_clusters, show_dims)`
 plot scatter data in inputted 2-dims per cluster using test.

Parameters

- **obs_model** ([KMeans](#)) – refer to document of KMeans.
- **comparison_clusters** (*array-like of shape(2)*) – set of clusters to compare.
- **show_dims** (*array-like of shape(2)*) – set of dims to show.

`si4ul.kmeans_si.plot_statistics_matrix(matrix, digit=3)`
 plot matrix of statistics that is calculated by each cluster combinations.

Parameters

- **matrix** (*array-like of shape(cluster_num, cluster_num)*) – matrix of test statistics.
- **digit** (*int*) – digit number to display.

`si4ul.kmeans_si.plot_violin(obs_model, test_gene)`
 plot violin per cluster using test and other.

Parameters

- **obs_model** ([KMeans](#)) – refer to document of KMeans.
- **test_gene** (*int*) – feature to plot.

2.1 si4ul.si.kmeans_si module

class si4ul.si.kmeans_si.KMeans(*X, n_clusters, max_iter=1000, random_seed=0*)

Bases: object

this class returns the results of k-means clustering. we can get from kmeans_si.kmeans(). other kmeans_si APIs need this object to input.

cluster_centers_

cluster center matrix.

Type array-like of shape(*n_clusters, d*)

count

count of iteration in k-means algorithm.

Type int

labels_

label vector that each data join.

Type array-like of shape(*n*)

label_num_list

list of the number of data contained in the cluster.

Type array-like of shape(*n_clusters*)

max_iter

upper limit count of iteration in k-means algorithm.

Type int

n_clusters

number of cluster.

Type int

random_seed

seed of random for determine initial cluster.

Type int

X

data matrix.

Type array-like of shape(*n, d*)

Examples

```
>>> kMeans.X
array([[ 1.32473015,  0.12584954], [ 0.17229311,  2.01200624], [ 1.47662313, -1.
↪ 28557418], [ 0.1302503 , -0.43927367], [-1.41546232,  0.13654847], [-1.05260842,
↪ 1.20597647], [-0.14717876, -0.15950429], [-0.61262757,  0.05772617], [ 0.92854842,
↪ -0.49440228], [-0.80456805, -1.15935248]])
>>> kMeans.n_clusters
3
>>> kMeans.cluster_centers_
array([[ 1.47662313, -1.28557418], [ 0.55908753, -0.24183267], [-0.74259465,  0.
↪ 45058097]])
```

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```
>>> kMeans.labels_  
array([1, 2, 0, 1, 2, 2, 1, 2, 1, 2])  
>>> kMeans.label_num_list  
[1, 4, 5]  
>>> kMeans.count  
2
```


SI4UL.SEGMENTATION_SI MODULE

`si4ul.segmentation_si.plot_histogram(local_white)`
plot histogram of pixel values in input image.

Parameters `local_white` (`LocalWhite`) – segmetation object.

`si4ul.segmentation_si.plot_histogram_region(local_white, display_statistics=False)`
plot histogram of pixel values per region.

Parameters `local_white` (`LocalWhite`) – segmetation object.

`si4ul.segmentation_si.psegi_thresholding(local_white, sigma=10)`
post segmentation inference for local white segmentation.

Parameters

- `local_white` (`LocalWhite`) – segmetation object.
- `sigma` (`float`) – standard deviation in distribution of input image.

Returns test statistics, PSegI p-value, naive p-value.

Return type (`float`, `float`, `float`)

Examples

```
>>> print(segmentation_si.psegi_thresholding(local_white))
(31.08397312484288, 0.08085241645874175, 0.0)
```

`si4ul.segmentation_si.thresholding(image_path, result_dir_path, window_size=50, bias=1.1, is_blur=True, ksize=(11, 11), sigma_x=0, sigma_y=0, is_output_regions=False)`

local white segmentation algorithm. If bias is bigger than 1, algorithm can detect black object. If bias is smaller than 1, algorithm can detect white object.

Parameters

- `image_path` (`string`) – image path to segmentation.
- `result_dir_path` (`string`) – directory to store the segmentation results.
- `window_size` (`int`) – length of the side of the square in the nearest pixel range.
- `bias` (`float`) – bias used in threshold determination. If it's bigger than 1, algorithm can detect black object. If it's smaller than 1, algorithm can detect white object.
- `isBlur` (`bool`) – whether to use Gaussian smoothing for image preprocessing.
- `ksize` (`Tuple(int, int)`) – the side of the convolution window for Gaussian smoothing.

- **sigma_x** (*float*) – standard deviation of x-coordinate for Gaussian smoothing.
- **sigma_y** (*float*) – standard deviation of y-coordinate for Gaussian smoothing.
- **is_output_regions** (*bool*) – wheter to make images of each regions.

Returns refer to document of LocalWhite.

Return type *LocalWhite*

Examples

```
>>> print(segmentation_si.thresholding('./image.jpg', './result/image_18'))
<si4ul.si.segmentation_si.LocalWhite at 0x10e33efa0>
```

3.1 si4ul.si.segmentation_si module

class si4ul.si.segmentation_si.**LocalWhite**(*image_path, result_dir_path, window_size, bias, is_blur, ksize, sigma_x, sigma_y*)

Bases: object

this class returns the results of local white segmentation. we can get from segmentation_si.thresholding(). other segmentation_si APIs need this object to input.

bias

bias used in threshold determination. If it's bigger than 1, algorithm can detect black object. If it's smaller than 1, algorithm can detect white object.

Type float

image

input image of segmentation.

Type array-like of shape (image_height, image_width)

image_gaussian

image after Gaussian smoothing.

Type array-like of shape (image_height, image_width)

image_height

height of image.

Type int

image_original

image matrix read in from the input image.

Type array-like of shape (image_height, image_width)

image_path

image path to segmentation.

Type string

image_width

width of image.

Type int

result_dir_path

directory to store the segmentation results.

Type string

window_size

length of the side of the square in the nearest pixel range.

Type int

Examples

```
>>> localWhite.image
array([[147, 147, 147, ..., 155, 157, 157],
       [147, 147, 148, ..., 154, 156, 156],
       [148, 148, 148, ..., 153, 154, 154],
       ...,
       [159, 158, 157, ..., 152, 151, 150],
       [159, 159, 157, ..., 152, 150, 150],
       [159, 159, 157, ..., 152, 150, 149]], dtype=uint8)
>>> localWhite.image_height, localWhite.image_width
(88, 73)
```


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