

`sklearn.metrics.DistanceMetric`

```
class sklearn.metrics.DistanceMetric
```

DistanceMetric class

This class provides a uniform interface to fast distance metric functions. The various metrics can be accessed via the [get_metric](#) class method and the metric string identifier (see below).

Examples

```
>>> from sklearn.metrics import DistanceMetric
>>> dist = DistanceMetric.get_metric('euclidean')
>>> X = [[0, 1, 2],
        [3, 4, 5]]
>>> dist.pairwise(X)
array([[ 0.,          5.19615242],
       [ 5.19615242,  0.]])
```

>>>

Available Metrics

The following lists the string metric identifiers and the associated distance metric classes:

Metrics intended for real-valued vector spaces:

identifier	class name	args	distance function
"euclidean"	EuclideanDistance	•	$\sqrt{\sum (x - y)^2}$
"manhattan"	ManhattanDistance	•	$\sum x - y $
"chebyshev"	ChebyshevDistance	•	$\max x - y $
"minkowski"	MinkowskiDistance	p, w	$\sum (w * x - y ^p)^{1/p}$
"wminkowski"	WMinkowskiDistance	p, w	$\sum w * (x - y) ^p^{1/p}$
"seuclidean"	SEuclideanDistance	V	$\sqrt{\sum (x - y)^2 / V}$
"mahalanobis"	MahalanobisDistance	V or VI	$\sqrt{(x - y)' V^{-1} (x - y)}$

Deprecated since version 1.1: `WMinkowskiDistance` is deprecated in version 1.1 and will be removed in version 1.3. Use `MinkowskiDistance` instead. Note that in `MinkowskiDistance`, the weights are applied to the absolute differences already raised to the p power. This is different from `WMinkowskiDistance` where weights are applied to the absolute differences before raising to the p power. The deprecation aims to remain consistent with SciPy 1.8 convention.

Metrics intended for two-dimensional vector spaces: Note that the haversine distance metric requires data in the form of [latitude, longitude] and both inputs and outputs are in units of radians.

identifier	class name	distance function
"haversine"	HaversineDistance	$2 \arcsin(\sqrt{\sin^2(0.5 * dx) + \cos(x1) \cos(x2) \sin^2(0.5 * dy)})$

Metrics intended for integer-valued vector spaces: Though intended for integer-valued vectors, these are also valid metrics in the case of real-valued vectors.

identifier	class name	distance function
"hamming"	HammingDistance	$N_{\text{unequal}}(x, y) / N_{\text{tot}}$
"canberra"	CanberraDistance	$\sum x - y / (x + y)$
"braycurtis"	BrayCurtisDistance	$\sum x - y / (\sum x + \sum y)$

Metrics intended for boolean-valued vector spaces: Any nonzero entry is evaluated to "True". In the listings below, the following abbreviations are used:

- N : number of dimensions
- NTT : number of dims in which both values are True
- NTF : number of dims in which the first value is True, second is False
- NFT : number of dims in which the first value is False, second is True
- NFF : number of dims in which both values are False
- NNEQ : number of non-equal dimensions, $NNEQ = NTF + NFT$
- NNZ : number of nonzero dimensions, $NNZ = NTF + NFT + NTT$

Toggle Menu

identifier	class name	distance function
"jaccard"	JaccardDistance	$NNEQ / NNZ$
"matching"	MatchingDistance	$NNEQ / N$
"dice"	DiceDistance	$NNEQ / (NTT + NNZ)$
"kulsinski"	KulsinskiDistance	$(NNEQ + N - NTT) / (NNEQ + N)$
"rogerstanimoto"	RogersTanimotoDistance	$2 * NNEQ / (N + NNEQ)$
"russellrao"	RussellRaoDistance	$(N - NTT) / N$
"sokalmichener"	SokalMichenerDistance	$2 * NNEQ / (N + NNEQ)$
"sokalsneath"	SokalSneathDistance	$NNEQ / (NNEQ + 0.5 * NTT)$

User-defined distance:

identifier	class name	args
"pyfunc"	PyFuncDistance	func

Here `func` is a function which takes two one-dimensional numpy arrays, and returns a distance. Note that in order to be used within the `BallTree`, the distance must be a true metric: i.e. it must satisfy the following properties

1. Non-negativity: $d(x, y) \geq 0$
2. Identity: $d(x, y) = 0$ if and only if $x == y$
3. Symmetry: $d(x, y) = d(y, x)$
4. Triangle Inequality: $d(x, y) + d(y, z) \geq d(x, z)$

Because of the Python object overhead involved in calling the python function, this will be fairly slow, but it will have the same scaling as other distances.

Methods

<code>dist_to_rdist</code>	Convert the true distance to the rank-preserving surrogate distance.
<code>get_metric</code>	Get the given distance metric from the string identifier.
<code>pairwise</code>	Compute the pairwise distances between X and Y
<code>rdist_to_dist</code>	Convert the rank-preserving surrogate distance to the distance.

`dist_to_rdist()`

Convert the true distance to the rank-preserving surrogate distance.

The surrogate distance is any measure that yields the same rank as the distance, but is more efficient to compute. For example, the rank-preserving surrogate distance of the Euclidean metric is the squared-euclidean distance.

Parameters:

dist : *double*
True distance.

Returns:

double
Surrogate distance.

`get_metric()`

Get the given distance metric from the string identifier.

See the docstring of `DistanceMetric` for a list of available metrics.

Parameters:

metric : *str or class name*
The distance metric to use

****kwargs**
additional arguments will be passed to the requested metric

`pairwise()`

Compute the pairwise distances between X and Y

[Toggle Menu](#)

venience routine for the sake of testing. For many metrics, the utilities in `scipy.spatial.distance.cdist` and

`scipy.spatial.distance.pdist` will be faster.

Parameters:**X : array-like**

Array of shape (Nx, D), representing Nx points in D dimensions.

Y : array-like (optional)

Array of shape (Ny, D), representing Ny points in D dimensions. If not specified, then Y=X.

Returns:**dist : ndarray**

The shape (Nx, Ny) array of pairwise distances between points in X and Y.

`rdist_to_dist()`

Convert the rank-preserving surrogate distance to the distance.

The surrogate distance is any measure that yields the same rank as the distance, but is more efficient to compute. For example, the rank-preserving surrogate distance of the Euclidean metric is the squared-euclidean distance.

Parameters:**rdist : double**

Surrogate distance.

Returns:**double**

True distance.

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