

# **Part I**

# **Python**

## **Part II**

# **Scikit-Learn**

# Chapter 1

## feature\_extraction

### 1.1 DictVectorizer

### 1.2 text

#### 1.2.1 CounterVector

#### 1.2.2 TfidfVectorizer

Table 1.1: feature\_extraction

CounterVector	DictVectorizer	TfidfVectorizer
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# Chapter 2

## preprocessing

### 2.1 PolynomialFeatures

Table 2.1: preprocessing

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PolynomialFeatures

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## Chapter 3

# impute

### 3.1 SimpleImputer

Table 3.1: impute

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[SimpleImputer](#)

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# Chapter 4

## pipeline

### 4.1 make\_pipeline

Table 4.1: pipeline

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[make\\_pipeline](#)

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# Chapter 5

## datasets

### 5.1 make\_blobs

### 5.2 make\_friedman1

```
make_friedman1(n_samples=100, n_features=10, *, noise=0.0, random_state=None)
```

Inputs  $X$  are independent features uniformly distributed on the interval  $[0, 1]$ . The output  $y$  is created according to the formula:

$$y = 10 \sin(\pi x_1 x_2) + 20(x_3 - \frac{1}{2})^2 + 10x_4 + 5x_5 + \text{Gaussian Noise}(0, \sigma) \quad (5.1)$$

A synthetic data set called *Friedman-1*, originally created by Jerome Friedman in 1991 to explore how well his new multivariate adaptive regression splines (MARS) algorithm was fitting high-dimensional data.

This data set was carefully generated to evaluate a regression method's ability to only pick up true feature dependencies in the data set and ignore others.

Table 5.2: make\_friedman1 主要参数

Properties	Names	Descriptions
Parameters	<code>n_samples: int, default=100</code>	The number of samples.
Parameters	<code>n_features: int, default=10</code>	The number of features. Should be at least 5.
Parameters	<code>noise: float, default=0.0</code>	The standard deviation of the gaussian noise applied to the output.
Returns	<code>X: ndarray of shape (n_samples, n_features)</code>	The input samples.
Returns	<code>y: ndarray of shape (n_samples,)</code>	The output values.

Table 5.1: datasets

<a href="#">fetch_20newsgroups</a>	<a href="#">fetch_lfw_people</a>	<a href="#">make_friedman1</a>	<a href="#">make_friedman2</a>	<a href="#">make_friedman3</a>	<a href="#">make_circles</a>
<a href="#">make_blobs</a>					

**5.3** `make_friedman2`

**5.4** `make_friedman3`

**5.5** `make_circles`

**5.6** `fetch_20newsgroups`

**5.7** `fetch_lfw_people`



# Chapter 6

## naive\_bayes

### 6.1 GaussianNB

### 6.2 MultinomialNB

Table 6.1: naive\_bayes

GaussianNB	MultinomialNB
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# Chapter 7

## metrics

**7.1** `confusion_matrix`

**7.2** `pairwise_distances`

**7.3** `pairwise_distances_argmin`

**7.4** `pairwise_distances_argmax`

Table 7.1: metrics

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<code>confusion_matrix</code>	<code>pairwise_distances</code>	<code>pairwise_distances_argmin</code>	<code>pairwise_distances_argmax</code>
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# Chapter 8

## linear\_model

### 8.1 LinearRegression

### 8.2 Ridge

### 8.3 Lasso

Table 8.1: linear\_model

LinearRegression	Ridge	Lasso
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# Chapter 9

## utils

### 9.1 resample

Table 9.1: utils

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[resample](#)

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# Chapter 10

## svm

### 10.1 svc

Table 10.1: svm

<a href="#">svc</a>
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# Chapter 11

## cluster

### 11.1 KMean

### 11.2 SpectralClustering

### 11.3 MiniBatchKMeans

Table 11.1: cluster

KMean	SpectralClustering	MiniBatchKMeans
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## **Part III**

# **NumPy**

# Chapter 12

## routines

### 12.1 Mathematical functions

#### 12.1.1 prod

### 12.2 Set routines

#### 12.2.1 setdiff1d

`numpy.setdiff1d(ar1, ar2, assume_unique=False)`

Find the set difference of two arrays.

Return the unique values in *ar1* that are not in *ar2*.

Table 12.1: routines: Mathematical functions

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<a href="#">prod</a>	<a href="#">setdiff1d</a>
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