# SGD example

```
import numpy as np
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
% matplotlib inline
# 生成随机数据集
np.random.seed(42)
X = 2 * np.random.rand(100, 1)
y = 4 + 3 * X + np.random.randn(100, 1)
X_b = np.c_[np.ones((100, 1)), X]
X \text{ new = np.array}([[0], [2]])
X_{new_b} = np.c_{np.ones((2, 1)), X_{new}}
plt.plot(X, y, 'b.')
     [<matplotlib.lines.Line2D at 0x7fdad071f250>]
      11
      10
       9
       8
       7
       6
       5
       4
       3
          0.00
               0.25
                     0.50
                           0.75
                                 100
                                       1.25
                                            150
                                                  1.75
```

```
# Batch Gradient Descent
eta = 0.1
n iterations = 1000
m = 100
theta = np.random.randn(2, 1)
for iteration in range(n_iterations):
    gradients = 2.0 / m * X_b.T.dot(X_b.dot(theta) - y)
    theta = theta - eta * gradients
    y_predict = X_new_b.dot(theta)
    plt.plot(X_new, y_predict, 'r-')
print("BGD:", theta)
# Stochastic Gradient Descent
n = 50
t0, t1 = 5, 50
theta = np.random.randn(2, 1)
def learning schedule(t):
    return float(t0) / (t + t1)
```

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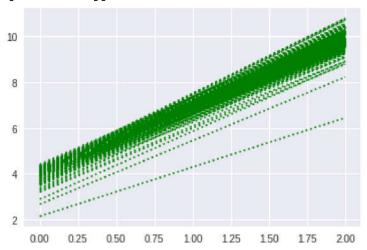
```
for epoch in range(n epochs):
   for i in range(m):
       random_index = np.random.randint(m)
       xi = X_b[random_index:random_index + 1]
       yi = y[random_index:random_index + 1]
       gradients = 2.0 * xi.T.dot(xi.dot(theta) - yi)
       eta = learning_schedule(epoch * m + i)
       theta = theta - eta * gradients
       y predict = X new b.dot(theta)
       plt.plot(X_new, y_predict, 'g:')
print("SGD:", theta)
     ('BGD:', array([[4.21509616],
₽
            [2.77011339]]))
     ('SGD:', array([[4.22352952],
            [2.77561526]]))
      10
             8
      6
      4
      2
```

### # Scikit-Learn SGD

from sklearn.linear\_model import SGDRegressor
sgd\_reg = SGDRegressor(n\_iter=50, penalty=None, eta0=0.1)
sgd\_reg.fit(X, y.ravel())
print(np.c\_[sgd\_reg.intercept\_, sgd\_reg.coef\_].T)

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## [4.20741477] [2.81806411]]



#### # Scikit-Learn SGD

from sklearn.linear\_model import SGDRegressor

```
sgd_reg = SGDRegressor(n_iter=50, penalty=None, eta0=0.1)
sgd_reg.fit(X, y.ravel())
print(np.c_[sgd_reg.intercept_, sgd_reg.coef_].T)

D [[4.2046679]
[2.77424383]]
```

## - OLS

simple linear regression implementation with example.

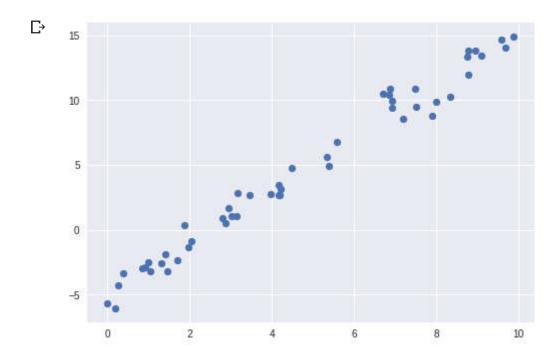
We will start with the most familiar linear regression, a straight-line fit to data. A straight-line fit is a model of the form

$$y = ax + b$$

where a is commonly known as the *slope*, and b is commonly known as the *intercept*.

Consider the following data, which is scattered about a line with a slope of 2 and an intercept of -5:

```
import seaborn as sns; sns.set()
rng = np.random.RandomState(1)
x = 10 * rng.rand(50)
y = 2 * x - 5 + rng.randn(50)
plt.scatter(x, y);
```



We can use Scikit-Learn's LinearRegression estimator to fit this data and construct the best-fit line:

```
from sklearn.linear_model import LinearRegression
model = LinearRegression(fit_intercept=True)

model.fit(x[:, np.newaxis], y)

xfit = np.linspace(0, 10, 1000)
yfit = model.predict(xfit[:, np.newaxis])
```

```
print("Model slope: ", model.coef_[0])
print("Model intercept:", model.intercept_)

plt.scatter(x, y)
plt.plot(xfit, yfit);
```

('Model slope: ', 2.0272088103606953) ('Model intercept:', -4.998577085553204)

