```
In [27]:
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear model import LinearRegression
from sklearn.datasets import make regression
%matplotlib notebook
In [2]:
X, y = make_regression()
In [6]:
print(X.shape, y.shape)
(100, 100) (100,)
In [18]:
X, y = make_regression(n_samples=1000, n_features=2, n_informative=2, noise=10 ,random_s
tate=1)
In [19]:
print(X.shape, y.shape)
(1000, 2) (1000,)
In [20]:
dfX = pd.DataFrame(X, columns=['f1', 'f2'])
Out[20]:
          f1
                  f2
  0 -0.915424 -1.945047
  1 -2.304908 -0.592461
  2 -0.437509 -0.781912
  3 -1.061797 -0.569149
  4 0.008359 0.034066
995 -0.479185 -0.418938
996 -1.899463 -1.168961
997 -0.468674 0.021863
998 -0.442655 0.557533
999 -0.139712 0.853282
1000 rows × 2 columns
In [21]:
dfY = pd.DataFrame(y, columns=['y'])
dfY
Out[21]:
```

```
0 -198.367142
      -71.877706
      -86.970324
      -69.649240
  3
      18.274576
995
      -38.361458
996
     -146.401642
997
      -19.200111
998
      47.016140
      70.677107
999
```

1000 rows × 1 columns

```
In [22]:
```

```
print(type(X[:, 0]))
print(type(y))

print(X[:, 0].shapea)
print(y.shape)

<class 'numpy.ndarray'>
<class 'numpy.ndarray'>
(1000,)
(1000,)
```

In [23]:

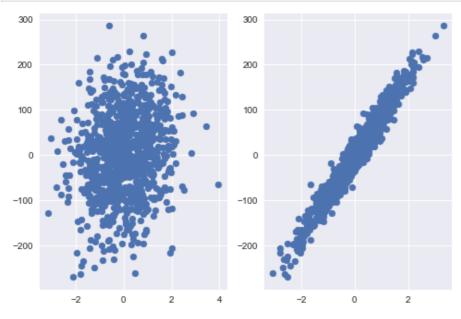
```
plt.style.use('seaborn')
```

In [24]:

```
plt.subplot(1, 2, 1)
plt.scatter(X[:, 0], a y)

plt.subplot(1, 2, 2)
plt.scatter(X[:, 1], y)

plt.show()
```

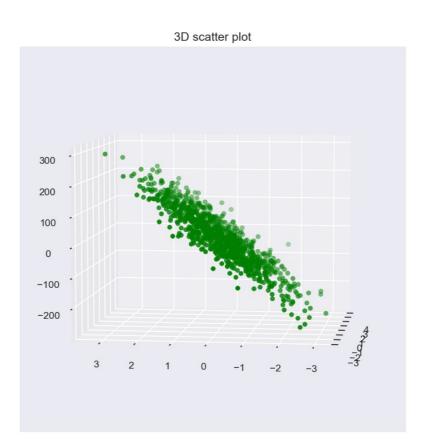


In [28]:

from mpl_toolkits import mplot3d

```
fig = plt.figure(figsize = (10, 7))
ax = plt.axes(projection='3d')

# Creating Plot
ax.scatter3D(X[:, 0], X[:, 1], y, color='green')
plt.title('3D scatter plot')
plt.show()
```



Linear Regression

```
In [29]:
model = LinearRegression()

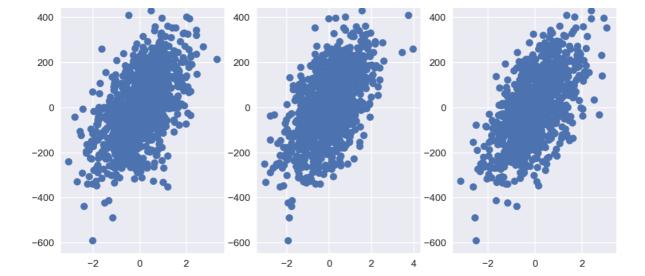
In [30]:
model.fit(X, y)

Out[30]:
LinearRegression()

In [31]:
model.coef_
Out[31]:
array([16.62437639, 87.99042847])

In [33]:
model.intercept_
Out[33]:
```

```
-0.21/66413/60018422
In [34]:
print(type(X[0]))
print(X[0].shape)
<class 'numpy.ndarray'>
(2,)
In [37]:
model.predict([X[0], X[1]])
Out[37]:
array([-186.58153868, -90.66624357])
In [38]:
y[0], y[1]
Out[38]:
(-198.36714169030483, -71.87770558832477)
In [41]:
# R2 score
model.score(X, y)
Out[41]:
0.988401838483157
In [42]:
# model.score?
Exploring
In [43]:
X, y = make regression(n samples=1000, n features=3, n informative=3, noise=10 ,random s
tate=1)
In [44]:
print(X.shape, y.shape)
(1000, 3) (1000,)
In [48]:
fig = plt.figure(figsize = (10, 5))
plt.subplot(1, 3, 1)
plt.scatter(X[:, 0], y)
plt.subplot(1, 3, 2)
plt.scatter(X[:, 1], y)
plt.subplot(1, 3, 3)
plt.scatter(X[:, 2], y)
plt.show()
```



In []:

```
from mpl_toolkits import mplot3d

fig = plt.figure(figsize = (10, 7))
ax = plt.axes(projection='3d')

# Creating Plot
ax.scatter3D(X[:, 0], X[:, 1], y, color='green')
plt.title('3D scatter plot')
plt.show()
```

In []:

In [49]:

```
model = LinearRegression()
model.fit(X, y)
print(model.coef_)
print(model.intercept_)

# R2 score
model.score(X, y)
```

[86.88487003 89.16247282 95.80681858] -0.02574715594392263

Out[49]:

0.9957566230931877

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