```
In [1]:
```

```
import numpy as np
In [5]:
X = np. array([2, 5, 1])
X. shape
Out[5]:
(3,)
In [6]:
np.linalg.norm(X, ord=1)
Out[6]:
8.0
In [8]:
a = np. array([
    [1, 2],
    [2, 3],
    [3, 4],
    [4, 5]
])
b = np.array([
    [2, 3],
    [3, 4],
    [4, 5],
    [5, 6]
])
a + b
Out[8]:
array([[ 3, 5],
```

```
[ 5, 7],
[ 7, 9],
[ 9, 11]])
```

```
In [10]:
```

Out[10]:

In [11]:

```
B= np.array([[2,1,-2],[3,-2,1],[2,-3,2]])
#逆行列の計算
inv_B = np.linalg.inv(B)
```

In [12]:

inv_B

Out[12]:

In [13]:

```
def sigmoid(a):
    return 1 / (1 + np. exp(-a))
```

In [20]:

```
sigmoid(100)
```

Out[20]:

1.0

In [22]:

```
def RSS(y, t):

return 0.5 * (np. sum((y - t)**2))
```

```
In [25]:
```

```
w = np. array([0. 2])
x = np. array([25])
t = np. array([1])
a = w. dot(x)
y = sigmoid(a)
E = RSS(y, t)
E
```

Out [25]:

2. 2397126747349496e-05

In [27]:

```
from keras.datasets import mnist (x_train, y_train), (x_test, y_test) = mnist.load_data() x_train.shape
```

Out [27]:

(60000, 28, 28)

In [28]:

```
x_train.ndim
```

Out[28]:

3

In [29]:

```
import pandas as pd
s = pd. Series([3, 2, 3, 2, 2])
s2 = pd. Series([2, 4, 6, 8, 10])
s
```

Out[29]:

```
0 3
1 2
2 3
3 2
4 2
dtype: int64
```

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```
Untitled
In [30]:
s2
Out[30]:
0
      2
      4
1
2
      6
3
      8
     10
dtype: int64
In [37]:
df = pd. concat([s, s2], axis=1)
\# df = pd.merge([s, s2], axis=1)
df
Out[37]:
   0
       1
   3
       2
 1 2
       4
 2 3
       6
 3 2
       8
 4 2 10
In [38]:
type(df)
Out[38]:
pandas. core. frame. DataFrame
In [41]:
```

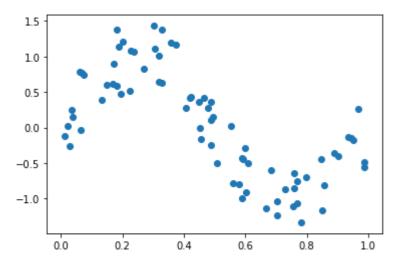
```
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split as tts
def f(x):
   return np. sin(2*np. pi*x)
```

In [46]:

```
sin_x = np.linspace(0, 1, 100)
x = np.random.rand(100)[:, np.newaxis]
y = f(x) + np.random.rand(100)[:, np.newaxis] - 0.5
x_train, x_test, y_train, y_test = tts(x, y)
# plt.plot(sin_x, f(sin_x), ':')
plt.scatter(x_train, y_train)
```

Out[46]:

<matplotlib.collections.PathCollection at 0x7f9aa05207d0>

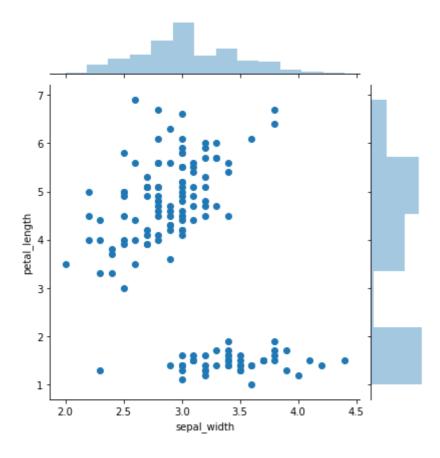


In [56]:

```
import seaborn as sns
iris = sns.load_dataset("iris")
type(iris)
sns.jointplot(x='sepal_width', y='petal_length', data=iris, kind='scatter')
```

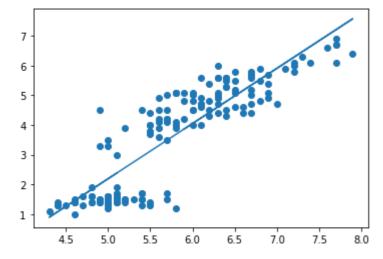
Out[56]:

<seaborn.axisgrid.JointGrid at 0x7f9aa366bd90>



In [66]:

```
from sklearn import datasets, linear_model
iris = datasets.load_iris()
x = iris.data[:, 0]
y = iris.data[:, 2]
model = linear_model.LinearRegression()
model.fit(x.reshape(-1, 1), y)
plt.scatter(x, y)
plt.plot(x, model.predict(x.reshape(-1, 1)))
# plt.plot(x, model.predict(y))
plt.show()
```



Out[66]:

```
array([5.1, 4.9, 4.7, 4.6, 5. , 5.4, 4.6, 5. , 4.4, 4.9, 5.4, 4.8, 4.8, 4.3, 5.8, 5.7, 5.4, 5.1, 5.7, 5.1, 5.4, 5.1, 4.6, 5.1, 4.8, 5. , 5. , 5.2, 5.2, 4.7, 4.8, 5.4, 5.2, 5.5, 4.9, 5. , 5.5, 4.9, 4.4, 5.1, 5. , 4.5, 4.4, 5. , 5.1, 4.8, 5.1, 4.6, 5.3, 5. , 7. , 6.4, 6.9, 5.5, 6.5, 5.7, 6.3, 4.9, 6.6, 5.2, 5. , 5.9, 6. , 6.1, 5.6, 6.7, 5.6, 5.8, 6.2, 5.6, 5.9, 6.1, 6.3, 6.1, 6.4, 6.6, 6.8, 6.7, 6. , 5.7, 5.5, 5.5, 5.8, 6. , 5.4, 6. , 6.7, 6.3, 5.6, 5.5, 5.5, 6.1, 5.8, 5. , 5.6, 5.7, 5.7, 6.2, 5.1, 5.7, 6.3, 5.8, 7.1, 6.3, 6.5, 7.6, 4.9, 7.3, 6.7, 7.2, 6.5, 6.4, 6.8, 5.7, 5.8, 6.4, 6.5, 7.7, 7.7, 6. , 6.9, 5.6, 7.7, 6.3, 6.4, 6.8, 5.7, 5.8, 6.4, 6.5, 7.4, 7.9, 6.4, 6.3, 6.1, 7.7, 6.3, 6.4, 6. , 6.9, 6.7, 6.9, 5.8, 6.8, 6.7, 6.7, 6.7, 6.3, 6.5, 6.2, 5.9])
```

In [70]:

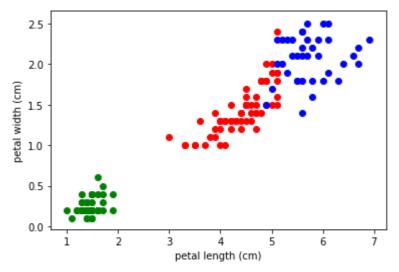
```
a = np. array([[1, 2, 3], [4, 5, 6]])
a. reshape(3, -1)
```

Out[70]:

```
array([[1, 2],
[3, 4],
[5, 6]])
```

In [84]:

```
from sklearn import cluster, datasets
iris = datasets.load_iris()
data = iris.data
model = cluster.KMeans(n_clusters=3)
model.fit(data)
labels = model.labels_
ldata = data[labels == 0]
plt.scatter(ldata[:,2], ldata[:,3], color='green')
ldata = data[labels == 1]
plt.scatter(ldata[:,2], ldata[:,3], color='red')
ldata = data[labels == 2]
plt.scatter(ldata[:,2], ldata[:,3], color='blue')
plt.scatter(ldata[:,2], ldata[:,3], color='blue')
plt.ylabel(iris['feature_names'][2])
plt.ylabel(iris['feature_names'][3])
plt.show()
```



In [85]:

```
a = np. array([[1, 2], [3, 4]])
b = np. array([[8, 7], [6, 5]])
a. dot(b)
```

Out[85]:

```
array([[20, 17],
[48, 41]])
```

In [108]:

```
x = np. array([
     [6],
     [9]
])
w = np. array([
     [3, 5],
     [4, 8]
])
b = np. array([
     [4],
     [6]
])
# w. shape
a1 = w. dot(x) + b
a1
```

Out[108]:

```
array([[ 67], [102]])
```

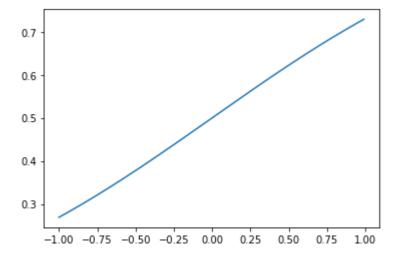
In [112]:

```
import matplotlib
import matplotlib.pyplot as plt
import numpy as np

t = np.arange(-1, 1, 0.01)
s = sigmoid(t)
plt.plot(t, s)
t
sigmoid(0.7), sigmoid(-0.3)
```

Out[112]:

(0.6681877721681662, 0.425557483188341)



In [114]:

```
a = np. array([
      [1, 2],
      [3, 4]
])

b = np. array([
      [1, 2],
      [2, 1]
])
a * b
```

Out[114]:

```
array([[1, 4],
[6, 4]])
```

In [115]:

```
y = np. array([
      [0.6],
      [0.2]
])

t = np. array([
      [0],
      [1]
])

(y - t) * ((1 - y) * y)
```

Out[115]:

```
array([[ 0.144],
[-0.128]])
```

微分と積分の具体例

 x^n の微分

$$nx^{n-1}$$
 $egin{array}{c} \int nx^{n-1}dx = x^n \
ightharpoonup \ rac{d}{dx}x^n = nx^{n-1} \end{array} x^n$

指数関数の微分

$$e^x egin{array}{ccc} \int e^x dx &= e^x \ &
ightharpoonup &
ightharpoonup & e^x \ rac{d}{dx} e^x &= e^x \end{array}$$

対数関数の微分

$$egin{array}{cccc} rac{1}{x} & \int rac{1}{x} dx = log_e x \
ightharpoons &
ightharpoons & log_e x \ rac{d}{dx} log_e x = rac{1}{x} \end{array}$$

三角関数の微分

$$-\sin x$$

$$\int (-\sin x) dx = \cos x$$
 $\rightleftharpoons \cos x$

$$\frac{d}{dx}\cos x = -\sin x$$

In []: