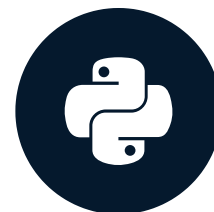


What is the difference between a NumPy array and a list?

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON

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NumPy array

```
import numpy as np
```

```
num_array = np.array([1, 2, 3, 4, 5])  
print(num_array)
```

```
[1 2 3 4 5]
```

```
num_list = [1, 2, 3, 4, 5]  
print(num_list)
```

```
[1, 2, 3, 4, 5]
```

Similarities between an array and a list

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_list = [1, 2, 3, 4, 5]
```

```
for item in num_array:  
    print(item)
```

```
for item in num_list:  
    print(item)
```

```
1  
2  
3  
4  
5
```

```
1  
2  
3  
4  
5
```

Similarities between an array and a list

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_list = [1, 2, 3, 4, 5]
```

```
num_array[1]
```

```
num_list[1]
```

```
2
```

```
2
```

```
num_array[1:4]
```

```
num_list[1:4]
```

```
array([2, 3, 4])
```

```
[2, 3, 4]
```

Similarities between an array and a list

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_list = [1, 2, 3, 4, 5]
```

```
num_array[3] = 40  
print(num_array)
```

```
num_list[3] = 40  
print(num_list)
```

```
[1 2 3 40 5]
```

```
[1, 2, 3, 40, 5]
```

```
num_array[0:3] = [10, 20, 30]  
print(num_array)
```

```
num_list[0:3] = [10, 20, 30]  
print(num_list)
```

```
[10 20 30 40 5]
```

```
[10, 20, 30, 40, 5]
```

Difference between an array and a list

NumPy arrays are designed for high efficiency computations

- NumPy arrays store values of the same type

.dtype property

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_array.dtype
```

```
dtype('int64')
```

Changing the data type of an element

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_array[2] = 'three'
```

ValueError

```
num_list = [1, 2, 3, 4, 5]
```

```
num_list[2] = 'three'  
print(num_list)
```

```
[1, 2, 'three', 4, 5]
```


Specifying the data type explicitly

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_array = np.array([1, 2, 3, 4, 5], dtype = np.dtype('int64'))  
print(num_array)
```

```
[1 2 3 4 5]
```

```
num_array.dtype
```

```
dtype('int64')
```

Specifying the data type explicitly

```
num_array = np.array([1, 2, 3, 4, 5])
```

```
num_array = np.array([1, 2, 3, 4, 5], dtype = np.dtype('str'))  
print(num_array)
```

```
['1' '2' '3' '4' '5']
```

```
num_array.dtype
```

```
dtype('<U1')
```

Object as a data type

```
num_array = np.array([1, 2, 3, 4, 5], dtype = np.dtype('O'))
```

```
num_array[2] = 'three'  
print(num_array)
```

```
[1 2 'three' 4 5]
```

Difference between an array and a list

NumPy arrays are designed for high efficiency computations

- NumPy arrays store values of a concrete data type
- NumPy arrays have a special way to access its elements

Accessing items

```
list2d = [  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
]
```

```
# Retrieve 8  
list2d[1][2]
```

8

```
array2d = np.array([  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
])
```

```
# Retrieve 8  
array2d[1][2]
```

8

Accessing items

```
list2d = [  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
]
```

```
# Retrieve 8  
list2d[1][2]
```

8

```
array2d = np.array([  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
])
```

```
# Retrieve 8  
array2d[1, 2]
```

8

Accessing items

```
list2d = [  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
]
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]
```

```
array2d = np.array([  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
])
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]
```

Accessing items

```
list2d = [  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
]
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]  
[  
    [list2d[j][1:4] for j in range(0, 2)]  
]
```

```
[[2, 3, 4], [7, 8, 9]]
```

```
array2d = np.array([  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
])
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]
```


Accessing items

```
list2d = [  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
]
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]  
[  
    [list2d[j][1:4] for j in range(0, 2)]  
]
```

```
[[2, 3, 4], [7, 8, 9]]
```

```
array2d = np.array([  
    [1, 2, 3, 4, 5],  
    [6, 7, 8, 9, 10],  
    [11, 12, 13, 14, 15]  
])
```

```
# Retrieve [[2, 3, 4], [7, 8, 9]]  
array2d[0:2, 1:4]
```

```
array([[2, 3, 4],  
       [7, 8, 9]])
```

Difference between an array and a list

NumPy arrays are designed for high efficiency computations

- NumPy arrays store values of a concrete data type
- NumPy arrays have a special way to access its elements
- NumPy arrays have efficient way to perform operations on them.

Operations +, -, *, / with lists

```
num_list1 = [1, 2, 3]  
num_list2 = [10, 20, 30]
```

```
num_list1 + num_list2
```

```
[1, 2, 3, 10, 20, 30]
```

```
num_list2 - num_list1
```

```
TypeError
```

```
num_list1 * num_list2
```

```
TypeError
```

```
num_list2 / num_list1
```

```
TypeError
```

Operations +, -, *, / with arrays

```
num_array1 = np.array([1, 2, 3])  
num_array2 = np.array([10, 20, 30])
```

```
num_array1 + num_array2
```

```
array([11, 22, 33])
```

```
num_array2 - num_array1
```

```
array([9, 18, 27])
```

```
num_array1 * num_array2
```

```
array([10, 40, 90])
```

```
num_array2 / num_array1
```

```
array([10, 10, 10])
```

Operations +, -, *, / with multidimensional arrays

```
num_array1 = np.array([
    [1, 2, 3, 4, 5],
    [6, 7, 8, 9, 10],
    [11, 12, 13, 14, 15]
])
num_array2 = np.array([
    [10, 20, 30, 40, 50],
    [60, 70, 80, 90, 100],
    [110, 120, 130, 140, 150]
])
```

```
num_array1 + num_array2
```

```
array([[ 11,  22,  33,  44,  55],
       [ 66,  77,  88,  99, 110],
       [121, 132, 143, 154, 165]])
```

```
num_array2 / num_array1
```

```
array([[10., 10., 10., 10., 10.],
       [10., 10., 10., 10., 10.],
       [10., 10., 10., 10., 10.]])
```

Conditional operations

`>` , `<` , `>=` , `<=` , `==` , `!=`

```
num_array = np.array([-5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5])
```

```
num_array < 0
```

```
array([True,  True,  True, False, False, False, False])
```

```
num_array[num_array < 0]
```

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

Broadcasting

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

```
num_array * 3
```

```
array([3, 6, 9])
```

```
num_array + 3
```

```
array([4, 5, 6])
```

```
num_list = [1, 2, 3]
```

```
num_list * 3
```

```
[1, 2, 3, 1, 2, 3, 1, 2, 3]
```

Broadcasting with multidimensional arrays

array2d (3 x 4)

```
array2d = np.array([  
    [1, 2, 3, 4],  
    [1, 2, 3, 4],  
    [1, 2, 3, 4]  
])
```

array1d (1 x 4)

```
array1d = np.array([1, 2, 3, 4])
```

array2d / array1d

```
array([[1., 1., 1., 1.],  
       [1., 1., 1., 1.],  
       [1., 1., 1., 1.]])
```


Broadcasting with multidimensional arrays

array2d (3 x 4)

```
array2d = np.array([  
    [1, 2, 3, 4],  
    [1, 2, 3, 4],  
    [1, 2, 3, 4]  
])
```

array1d (3 x 1)

```
array1d = np.array([[1], [2], [3]])
```

array2d / array1d

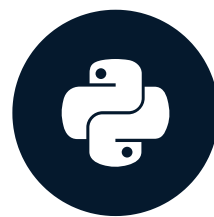
```
array([[1.    , 2.    , 3.    , 4.    ],  
       [0.5   , 1.    , 1.5   , 2.    ],  
       [0.333, 0.667, 1.    , 1.333]])
```

Let's practice

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON

How to use the .apply() method on a DataFrame?

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON



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Dataset

```
import pandas as pd

scores = pd.read_csv('exams.csv')
scores = scores[['math score', 'reading score', 'writing score']]
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

Default .apply()

```
df.apply(function)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.sqrt)  
print(score_new)
```

| | math score | reading score | writing score |
|-----|------------|---------------|---------------|
| 0 | 8.602325 | 9.273618 | 9.055385 |
| 1 | 6.633250 | 7.000000 | 7.280110 |
| 2 | 7.348469 | 6.782330 | 6.557439 |
| 3 | 9.380832 | 9.746794 | 9.591663 |
| 4 | 9.219544 | 9.000000 | 9.000000 |
| ... | | | |

Default .apply()

```
df.apply(function)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.mean)  
print(score_new.head())
```

```
math score      65.18  
reading score   69.28  
writing score   67.96  
dtype: float64
```

```
type(scores_new)
```

```
pandas.core.series.Series
```

Default .apply()

```
df.apply(function)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
function(pd.Series)
```

input size n

→ `np.sqrt(pd.Series)`

→ output size n

input size n

→ `np.mean(pd.Series)`

→ single value

Default .apply(): own functions

```
df.apply(function)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
def divide_scores(x):  
    return x / 2
```

```
scores_new = scores.apply(divide_scores)  
print(scores_new)
```

| | math score | reading score | writing score |
|-----|------------|---------------|---------------|
| 0 | 37.0 | 43.0 | 41.0 |
| 1 | 22.0 | 24.5 | 26.5 |
| 2 | 27.0 | 23.0 | 21.5 |
| 3 | 44.0 | 47.5 | 46.0 |
| 4 | 42.5 | 40.5 | 40.5 |
| ... | | | |

Default .apply(): own functions

```
df.apply(function)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
def perfect_score(x):  
    return 100
```

```
scores_new = scores.apply(perfect_score)  
print(scores_new)
```

```
math score      100  
reading score   100  
writing score   100  
dtype: int64
```

Lambda expressions

```
def divide_scores(x):  
    return x / 2
```

```
scores_new = scores.apply(divide_scores)  
print(scores_new)
```

```
   math score  reading score  writing score  
0      37.0      43.0      41.0  
1      22.0      24.5      26.5  
2      27.0      23.0      21.5  
3      44.0      47.5      46.0  
4      42.5      40.5      40.5  
...
```

```
def perfect_score(x):  
    return 100
```

```
scores_new = scores.apply(perfect_score)  
print(scores_new)
```

```
math score      100  
reading score   100  
writing score   100  
dtype: int64
```

Lambda expressions

```
scores_new = scores.apply(lambda x: x / 2)
print(scores_new)
```

```
   math score  reading score  writing score
0        37.0         43.0         41.0
1        22.0         24.5         26.5
2        27.0         23.0         21.5
3        44.0         47.5         46.0
4        42.5         40.5         40.5
...
```

```
scores_new = scores.apply(lambda x: 100)
print(scores_new)
```

```
math score      100
reading score   100
writing score   100
dtype: int64
```

Additional arguments: axis

```
df.apply(function, axis= )
```

Additional arguments: axis

```
df.apply(function, axis=0)
```

Additional arguments: axis

```
df.apply(function, axis=1)
```

Additional arguments: axis

```
df.apply(function, axis= )
```

`axis=0` - `function` is applied over columns

`axis=1` - `function` is applied over rows

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.mean)  
print(scores_new.head())
```

```
math score      65.18  
reading score   69.28  
writing score   67.96  
dtype: float64
```

Additional arguments: axis

```
df.apply(function, axis= )
```

`axis=0` - `function` is applied over columns

`axis=1` - `function` is applied over rows

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.mean, axis=0)  
print(scores_new.head())
```

```
math score      65.18  
reading score   69.28  
writing score   67.96  
dtype: float64
```


Additional arguments: axis

```
df.apply(function, axis= )
```

`axis=0` - `function` is applied over columns

`axis=1` - `function` is applied over rows

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.mean, axis=1)  
print(scores_new.head())
```

```
0    80.666667  
1    48.666667  
2    47.666667  
3    91.666667  
4    82.333333  
5    84.000000  
6    75.000000  
7    70.666667  
...
```

Additional arguments: result_type

```
df.apply(function, result_type= )
```

```
result_type='expand'
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy
```

```
def span(x):  
    return [np.min(x), np.max(x)]
```

```
scores_new = scores.apply(span)  
print(scores_new)
```

```
math score      [27, 100]  
reading score   [33, 100]  
writing score   [30, 100]  
dtype: object
```

Additional arguments: result_type

```
df.apply(function, result_type= )
```

```
result_type='expand'
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy
```

```
def span(x):  
    return [np.min(x), np.max(x)]
```

```
scores.apply(span, result_type='expand')
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 27 | 33 | 30 |
| 1 | 100 | 100 | 100 |

Additional arguments: result_type

```
df.apply(function, result_type= )
```

```
result_type='expand'
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy
```

```
def span(x):  
    return [np.min(x), np.max(x)]
```

```
scores.apply(span, result_type='expand', axis=1)
```

| | 0 | 1 |
|-----|----|----|
| 0 | 74 | 86 |
| 1 | 44 | 53 |
| 2 | 43 | 54 |
| 3 | 88 | 95 |
| 4 | 81 | 85 |
| ... | | |

Additional arguments: result_type

```
df.apply(function, result_type= )
```

```
result_type='broadcast'
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores_new = scores.apply(np.mean)
```

```
print(scores_new.head())
```

| | |
|---------------|---------|
| math score | 65.18 |
| reading score | 69.28 |
| writing score | 67.96 |
| dtype: | float64 |

Additional arguments: result_type

```
df.apply(function, result_type= )
```

```
result_type='broadcast'
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores.apply(np.mean, result_type='broadcast')
```

| | math score | reading score | writing score |
|-----|------------|---------------|---------------|
| 0 | 65 | 69 | 67 |
| 1 | 65 | 69 | 67 |
| 2 | 65 | 69 | 67 |
| 3 | 65 | 69 | 67 |
| 4 | 65 | 69 | 67 |
| 5 | 65 | 69 | 67 |
| 6 | 65 | 69 | 67 |
| 7 | 65 | 69 | 67 |
| ... | | | |

More than one argument in a function

```
function(pd.Series)
```

More than one argument in a function

```
function(pd.Series, arg1, arg2, ..., kwarg1=val1, kwarg2=val2, ...)
```

```
def check_mean(x, a, b, inside=True):  
    mean = np.mean(x)  
    if inside:  
        return mean > a and mean < b  
    else:  
        return mean < a or mean > b
```


Applying the function

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores.apply(check_mean)
```

```
TypeError
```

Additional arguments: args

```
df.apply(function, args= )
```

```
args - [arg1, arg2, ...]
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores.apply(check_mean, args=[67, 70])
```

```
math score      False
reading score    True
writing score    True
dtype: bool
```

Additional arguments: args

```
df.apply(function, args= )
```

```
args - (arg1, arg2, ...)
```

```
print(scores.head())
```

| | math score | reading score | writing score |
|---|------------|---------------|---------------|
| 0 | 74 | 86 | 82 |
| 1 | 44 | 49 | 53 |
| 2 | 54 | 46 | 43 |
| 3 | 88 | 95 | 92 |
| 4 | 85 | 81 | 81 |

```
import numpy as np
```

```
scores.apply(  
    check_mean, args=[67, 70], inside=False  
)
```

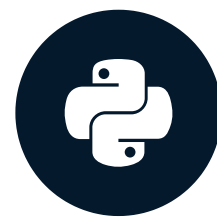
| | |
|---------------|-------|
| math score | True |
| reading score | False |
| writing score | False |
| dtype: bool | |

Let's practice!

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON

How to use the `.groupby()` method on a DataFrame?

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON



Kirill Smirnov

Data Science Consultant, Altran

Dataset

```
retinol = pd.read_csv('retinol.csv')
retinol = retinol[['age', 'gender', 'smoking', 'bmi', 'vitamin use', 'plasma B-carotene', 'plasma retinol']]
print(retinol.head())
```

| | age | gender | smoking | bmi | vitamin use | plasma B-carotene | plasma retinol |
|---|-----|--------|---------|----------|------------------|-------------------|----------------|
| 0 | 64 | Female | Former | 21.48380 | Yes_fairly_often | 200 | 915 |
| 1 | 76 | Female | Never | 23.87631 | Yes_fairly_often | 124 | 727 |
| 2 | 38 | Female | Former | 20.01080 | Yes_not_often | 328 | 721 |
| 3 | 40 | Female | Former | 25.14062 | No | 153 | 615 |
| 4 | 72 | Female | Never | 20.98504 | Yes_fairly_often | 92 | 799 |

background factors → plasma B-carotene , plasma retinol

.groupby()

groups the data according to some criteria allowing to perform an operation on each group.

```
df.groupby(column_name(s))
```

```
gens = retinol.groupby('gender')  
print(gens)
```

```
<pandas.core.groupby.groupby.DataFrameGroupBy object at 0x00000262DB5E2780>
```

```
gensmoks = retinol.groupby(['gender', 'smoking'])  
print(gensmoks)
```

```
<pandas.core.groupby.groupby.DataFrameGroupBy object at 0x00000262DB5F57B8>
```

Iterating through .groupby() output

```
gens = retinol.groupby('gender')

for group in gens:
    # Each group is a tuple
    # First element is a grouping factor
    print(group[0])
    # Second element is a DataFrame
    print(group[1].head(3))
```

```
len(gens)
```

```
2
```

```
Female
   age  gender  smoking      bmi  ...
0   64  Female  Former  21.48380  ...
1   76  Female   Never  23.87631  ...
2   38  Female  Former  20.01080  ...
Male
   age  gender  smoking      bmi  ...
12   57   Male   Never  31.73039  ...
14   66   Male   Never  27.31916  ...
15   64   Male  Former  31.44674  ...
```


Iterating through .groupby() output

```
gensmoks = retinol.groupby(['gender', 'smoking'])

for group in gensmoks:
    # Each group is a tuple
    # First element is a grouping factor
    print(group[0])
    # Second element is a DataFrame
    print(group[1].head(3))
```

```
len(gensmoks)
```

```
6
```

```
('Female', 'Current_Smoker')
   age  gender  smoking      bmi ...
32   74  Female  Current_Smoker  16.33114 ...
35   44  Female  Current_Smoker  25.87867 ...
43   31  Female  Current_Smoker  23.34593 ...
('Female', 'Former')
   age  gender  smoking      bmi ...
0    64  Female  Former    21.48380 ...
2    38  Female  Former    20.01080 ...
3    40  Female  Former    25.14062 ...
('Female', 'Never')
   age  gender  smoking      bmi ...
1    76  Female  Never    23.87631 ...
4    72  Female  Never    20.98504 ...
...
```

Standard operations on groups

```
gens = retinol.groupby('gender')
```

```
retinol['plasma retinol'].mean()
```

```
602.790476
```

```
retinol['vitamin use'].count()
```

```
315
```

```
gens['plasma retinol'].mean()
```

| | plasma retinol |
|--------|----------------|
| gender | |
| Female | 587.721612 |
| Male | 700.738095 |

```
gens['vitamin use'].count()
```

| | vitamin use |
|--------|-------------|
| gender | |
| Female | 273 |
| Male | 42 |

The .agg() method

`.agg(function, axis= , args=)` - almost identical to the `.apply()` method

```
import numpy as np
```

```
retinol['plasma retinol'].agg(np.mean)
```

```
602.790476
```

The .agg() method

`.agg(function, axis= , args=)` - almost identical to the `.apply()` method

```
import numpy as np
```

```
retinol[['plasma B-carotene', 'plasma retinol']].agg(np.mean)
```

```
plasma B-carotene    189.892063
plasma retinol       602.790476
dtype: float64
```

The .agg() method

`.agg(function, axis= , args=)` - almost identical to the `.apply()` method

```
import numpy as np
```

```
retinol[['plasma B-carotene', 'plasma retinol']].agg([np.mean, np.std])
```

| | plasma B-carotene | plasma retinol |
|------|-------------------|----------------|
| mean | 189.892063 | 602.790476 |
| std | 183.000803 | 208.895474 |

.groupby() followed by .agg()

```
gens = retinol.groupby('gender')
```

```
gens['plasma retinol'].agg([np.mean, np.std])
```

| | plasma retinol | |
|--------|----------------|------------|
| | mean | std |
| gender | | |
| Female | 587.721612 | 185.430687 |
| Male | 700.738095 | 307.808783 |

```
gensmoks = retinol.groupby(['gender', 'smoking'])
```

```
gensmoks['plasma retinol'].agg([np.mean, np.std])
```

| | | plasma retinol | |
|--------|----------------|----------------|------------|
| | | mean | std |
| gender | smoking | | |
| Female | Current_Smoker | 556.111111 | 191.112649 |
| | Former | 607.752688 | 187.983733 |
| | Never | 582.687500 | 182.182398 |
| Male | Current_Smoker | 598.857143 | 289.618961 |
| | Former | 798.500000 | 323.196203 |
| | Never | 590.153846 | 249.307991 |

Own functions and lambda expressions

```
gens = retinol.groupby('gender')
```

```
def n_more_than_mean(series):  
    result = series[series > np.mean(series)]  
    return len(result)
```

```
gens[['plasma B-carotene', 'retinol']].agg(n_more_than_mean)
```

| | plasma B-carotene | plasma retinol |
|--------|-------------------|----------------|
| gender | | |
| Female | 87 | 119 |
| Male | 13 | 19 |

Own functions and lambda expressions

```
gens = retinol.groupby('gender')
```

```
def n_more_than_mean(series):  
    result = series[series > np.mean(series)]  
    return len(result)
```

```
gens[['plasma B-carotene', 'plasma retinol']].agg([n_more_than_mean, lambda x: len(x)])
```

| | plasma B-carotene | | plasma retinol | |
|--------|----------------------|----------|----------------------|----------|
| | count_more_than_mean | <lambda> | count_more_than_mean | <lambda> |
| gender | | | | |
| Female | 87 | 273 | 119 | 273 |
| Male | 13 | 42 | 19 | 42 |

Renaming the output

```
gens = retinol.groupby('gender')
```

```
def n_more_than_mean(series):  
    result = series[series > np.mean(series)]  
    return len(result)
```

```
gens[['plasma B-carotene', 'plasma retinol']].agg({'count': n_more_than_mean, 'len': lambda x: len(x)})
```

| | count | | len | |
|--------|-------------------|----------------|-------------------|----------------|
| | plasma B-carotene | plasma retinol | plasma B-carotene | plasma retinol |
| gender | | | | |
| Female | 87 | 119 | 273 | 273 |
| Male | 13 | 19 | 42 | 42 |

The `.transform()` method

`.transform(function, axis= , args=)` - almost identical to the `.apply()` method

- The input and output must have the same size

```
import numpy as np

def center_scale(series):
    return (series - np.mean(series))/np.std(series)
```

DataFrame and the .transform() method

```
compounds = ['plasma B-carotene', 'retinol']  
df = retinol[compounds].transform(center_scale)  
  
print(df)
```

| | plasma B-carotene | plasma retinol |
|-----|-------------------|----------------|
| 0 | 0.055322 | 1.496951 |
| 1 | -0.360637 | 0.595547 |
| 2 | 0.755886 | 0.566779 |
| 3 | -0.201916 | 0.058541 |
| 4 | -0.535778 | 0.940766 |
| 5 | -0.229282 | 0.245534 |
| 6 | 0.372765 | 1.108580 |
| ... | | |
| 309 | -0.251174 | 0.715415 |
| 310 | -0.141711 | -1.854544 |
| 311 | -0.601456 | -1.317538 |
| 312 | 0.602637 | -0.483260 |
| 313 | -0.377057 | 0.389375 |
| 314 | 0.235936 | 1.070223 |

.groupby() followed by .transform()

```
gensmoks = retinol.groupby(['gender', 'smoking'])
```

```
compounds = ['plasma B-carotene', 'retinol']  
df = gensmoks[compounds].transform(center_scale)  
  
print(df)
```

```
      plasma B-carotene  plasma retinol  
0          -0.018568         1.643294  
1          -0.436191         0.794897  
2           0.629616         0.605697  
3          -0.256573         0.038762  
4          -0.597427         1.191485  
5          -0.281892         0.247351  
6           0.238985         1.384270  
...  
309         -0.302148         0.771498  
310         -0.200869        -2.095267  
311         -0.657891        -1.402860  
312           0.450607        -0.444440  
313         -0.418619         0.407804  
314           0.113019         1.340205
```

.groupby() followed by .transform()

```
gensmoks = retinol.groupby(['gender', 'smoking'])
```

```
compounds = ['plasma B-carotene', 'retinol']  
df = gensmoks[compounds].transform(  
    lambda x: (x - np.mean(x))/np.std(x)  
)  
  
print(df)
```

```
      plasma B-carotene  plasma retinol  
0          -0.018568         1.643294  
1          -0.436191         0.794897  
2           0.629616         0.605697  
3          -0.256573         0.038762  
4          -0.597427         1.191485  
5          -0.281892         0.247351  
6           0.238985         1.384270  
...  
309         -0.302148         0.771498  
310         -0.200869        -2.095267  
311         -0.657891        -1.402860  
312           0.450607        -0.444440  
313         -0.418619         0.407804  
314           0.113019         1.340205
```

The `.filter()` method of `DataFrameGroupBy` object

`.filter(function)`

`function` → `True` - group stays

`function` → `False` - group leaves

`function(pd.DataFrame)` - the function acts on the whole `DataFrame` in each group.

.groupby() followed by .filter()

```
gensmoks = retinol.groupby(['gender', 'smoking'])  
len(gensmoks)
```

```
6
```

```
def check_bmi(dataframe):  
    return np.mean(dataframe['bmi']) > 26
```

```
retinol_filtered = gensmoks.filter(check_bmi)  
print(retinol_filtered)
```

| | age | gender | smoking | bmi | ... |
|-----|-----|--------|---------|----------|-----|
| 1 | 76 | Female | Never | 23.87631 | ... |
| 4 | 72 | Female | Never | 20.98504 | ... |
| 6 | 65 | Female | Never | 22.01154 | ... |
| 7 | 58 | Female | Never | 28.75702 | ... |
| 8 | 35 | Female | Never | 23.07662 | ... |
| 11 | 40 | Female | Never | 36.43161 | ... |
| 13 | 66 | Female | Never | 21.78854 | ... |
| ... | | | | | |
| 299 | 47 | Female | Never | 37.27761 | ... |
| 302 | 41 | Female | Never | 34.61493 | ... |
| 306 | 66 | Female | Never | 33.10759 | ... |
| 311 | 45 | Female | Never | 23.82703 | ... |
| 312 | 49 | Female | Never | 24.26126 | ... |
| 314 | 45 | Female | Never | 26.50808 | ... |

.groupby() followed by .filter()

```
gensmoks = retinol.groupby(['gender', 'smoking'])  
len(gensmoks)
```

6

```
def check_bmi(dataframe):  
    return np.mean(dataframe['bmi']) > 26
```

```
retinol_filtered = gensmoks.filter(check_bmi)  
len(retinol_filtered.groupby(['gender', 'smoking']))
```

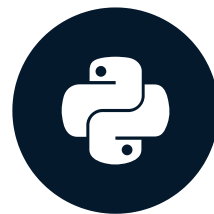
3

Let's practice!

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON

How to visualize data in Python?

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON



Kirill Smirnov

Data Science Consultant, Altran

matplotlib

```
import matplotlib.pyplot as plt
```

- scatter plot
- histogram
- boxplot

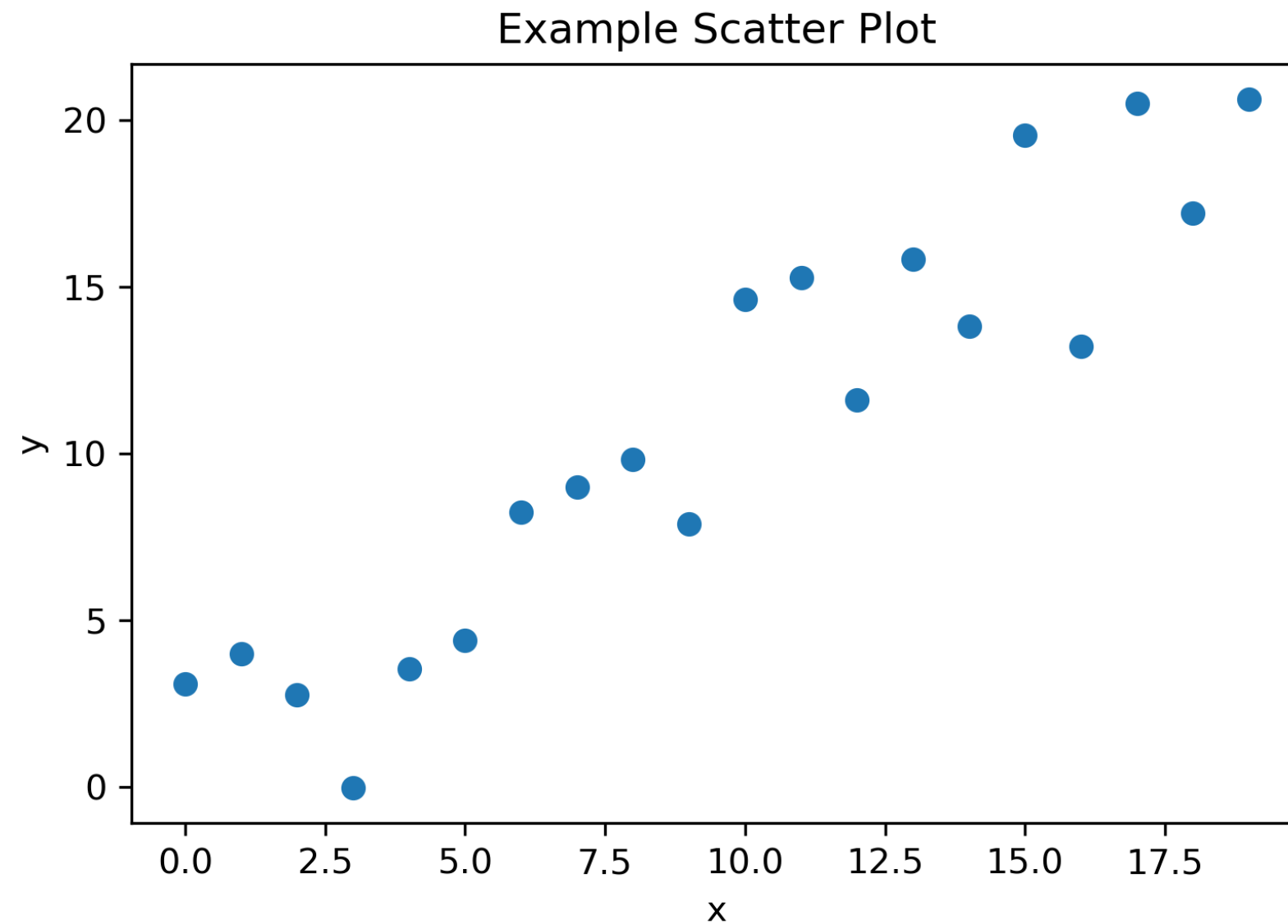
Dataset

```
import pandas as pd

diabetes = pd.read_csv('diabetes.csv')
diabetes = diabetes[[
    'n_pregnant', 'plasma glucose', 'blood pressure', 'skin thickness',
    'serum insulin', 'bmi', 'age', 'test result']]
print(diabetes.head())
```

| | n_pregnant | plasma glucose | blood pressure | skin thickness | serum insulin | bmi | age | test result |
|---|------------|----------------|----------------|----------------|---------------|------|-----|-------------|
| 0 | 6 | 148.0 | 72.0 | 35.0 | NaN | 33.6 | 50 | positive |
| 1 | 1 | 85.0 | 66.0 | 29.0 | NaN | 26.6 | 31 | negative |
| 2 | 8 | 183.0 | 64.0 | NaN | NaN | 23.3 | 32 | positive |
| 3 | 1 | 89.0 | 66.0 | 23.0 | 94.0 | 28.1 | 21 | negative |
| 4 | 0 | 137.0 | 40.0 | 35.0 | 168.0 | 43.1 | 33 | positive |

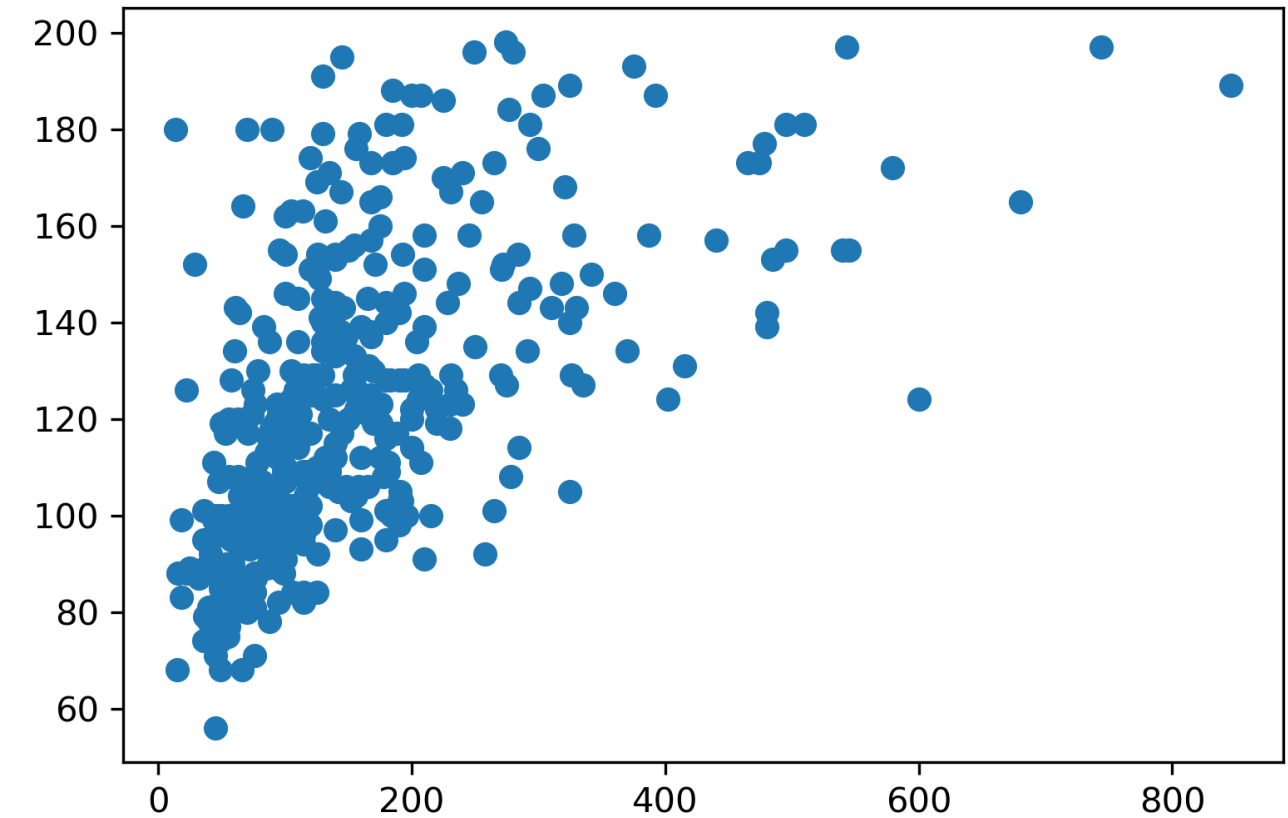
Scatter plot



Create a scatter plot

```
import matplotlib.pyplot as plt
```

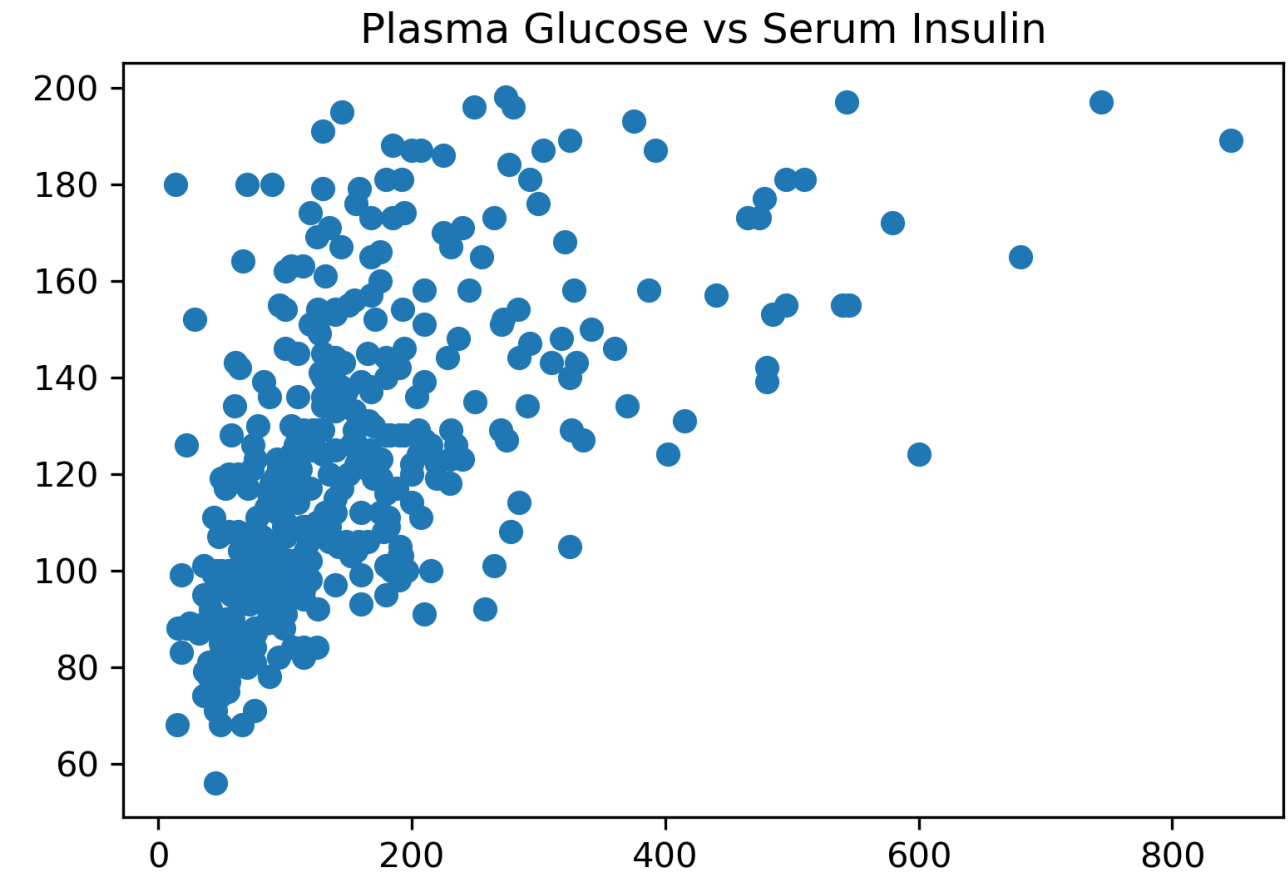
```
plt.scatter(  
    diabetes['serum insulin'],  
    diabetes['plasma glucose']  
)  
plt.show()
```



Create a scatter plot

```
import matplotlib.pyplot as plt
```

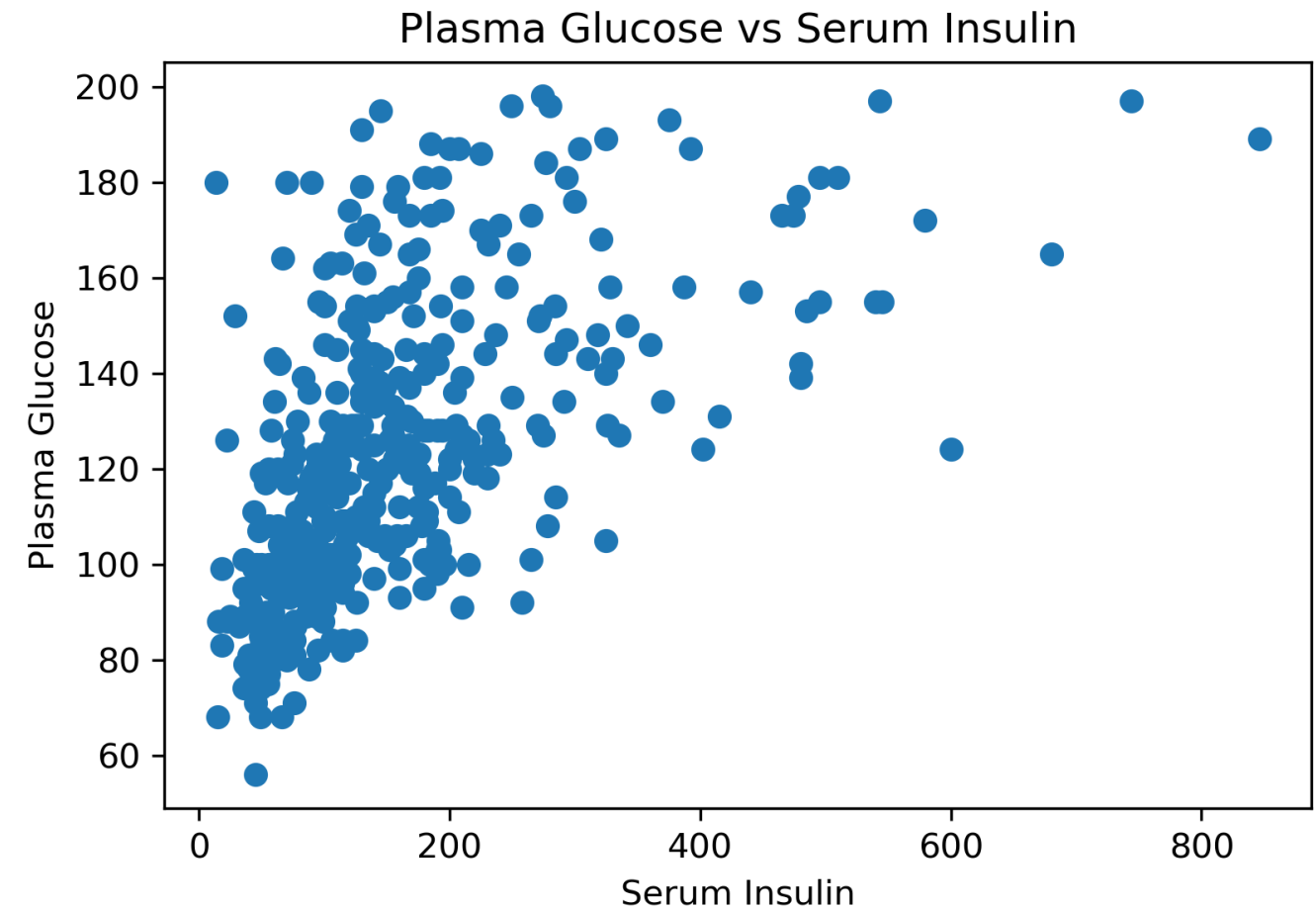
```
plt.scatter(  
    diabetes['serum insulin'],  
    diabetes['plasma glucose']  
)  
  
plt.title('Plasma Glucose vs Serum Insulin')  
  
plt.show()
```



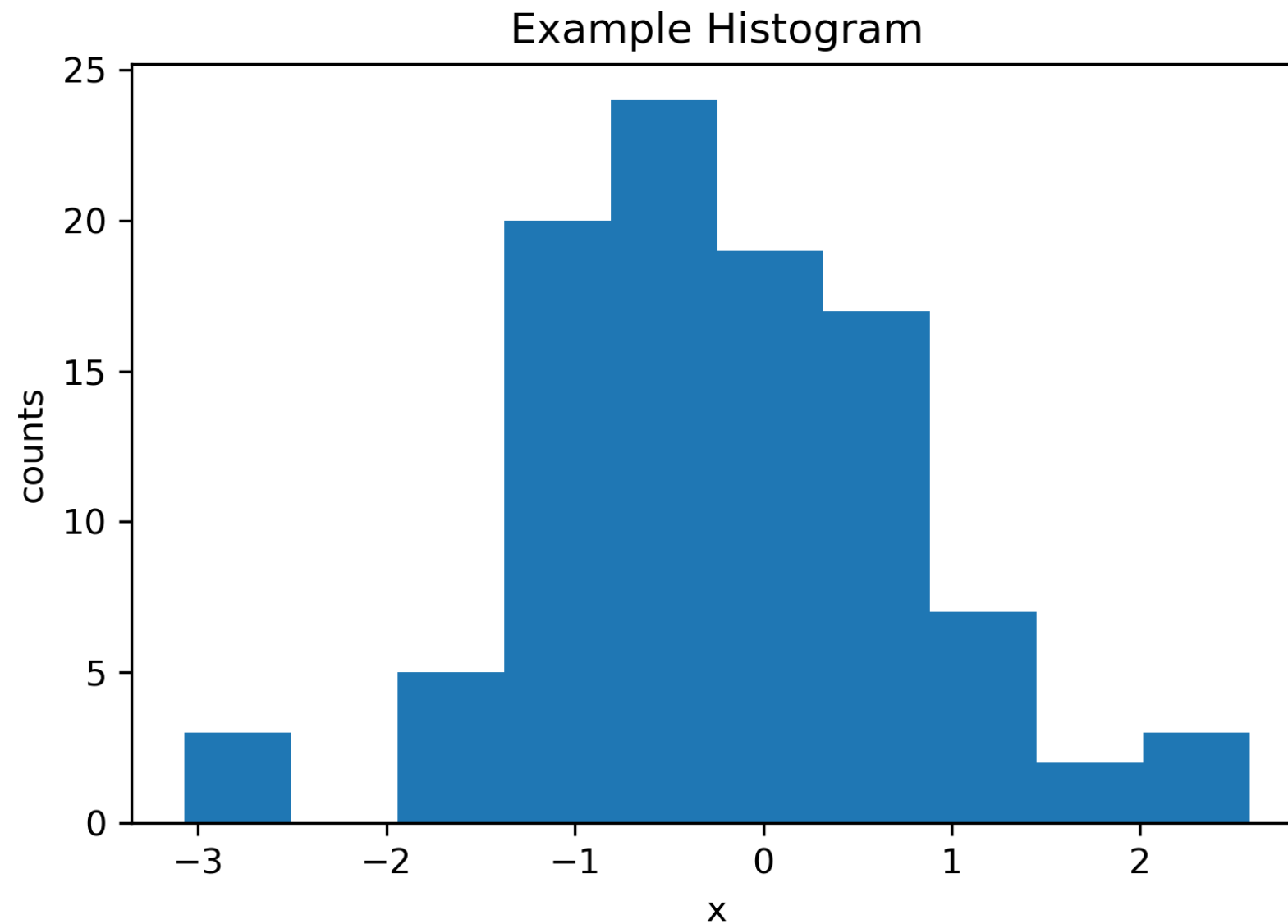
Create a scatter plot

```
import matplotlib.pyplot as plt
```

```
plt.scatter(  
    diabetes['serum insulin'],  
    diabetes['plasma glucose']  
)  
  
plt.title('Plasma Glucose vs Serum Insulin')  
  
plt.xlabel('Serum Insulin')  
plt.ylabel('Plasma Glucose')  
  
plt.show()
```



Histogram



Create a histogram

```
import matplotlib.pyplot as plt
```

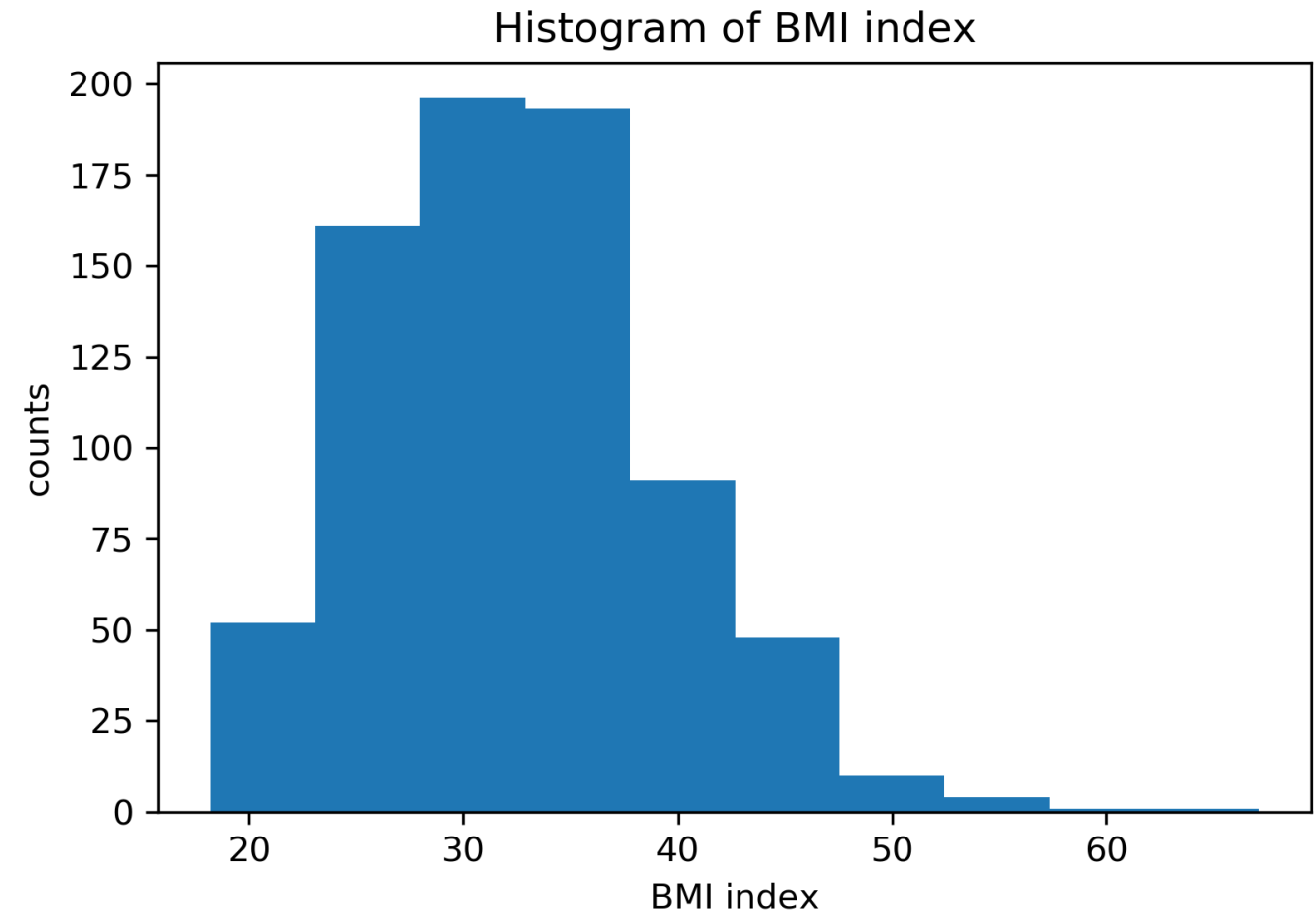
```
plt.hist(diabetes['bmi'])
```

```
plt.title('Histogram of BMI index')
```

```
plt.xlabel('BMI index')
```

```
plt.ylabel('counts')
```

```
plt.show()
```



Create a histogram

```
import matplotlib.pyplot as plt
```

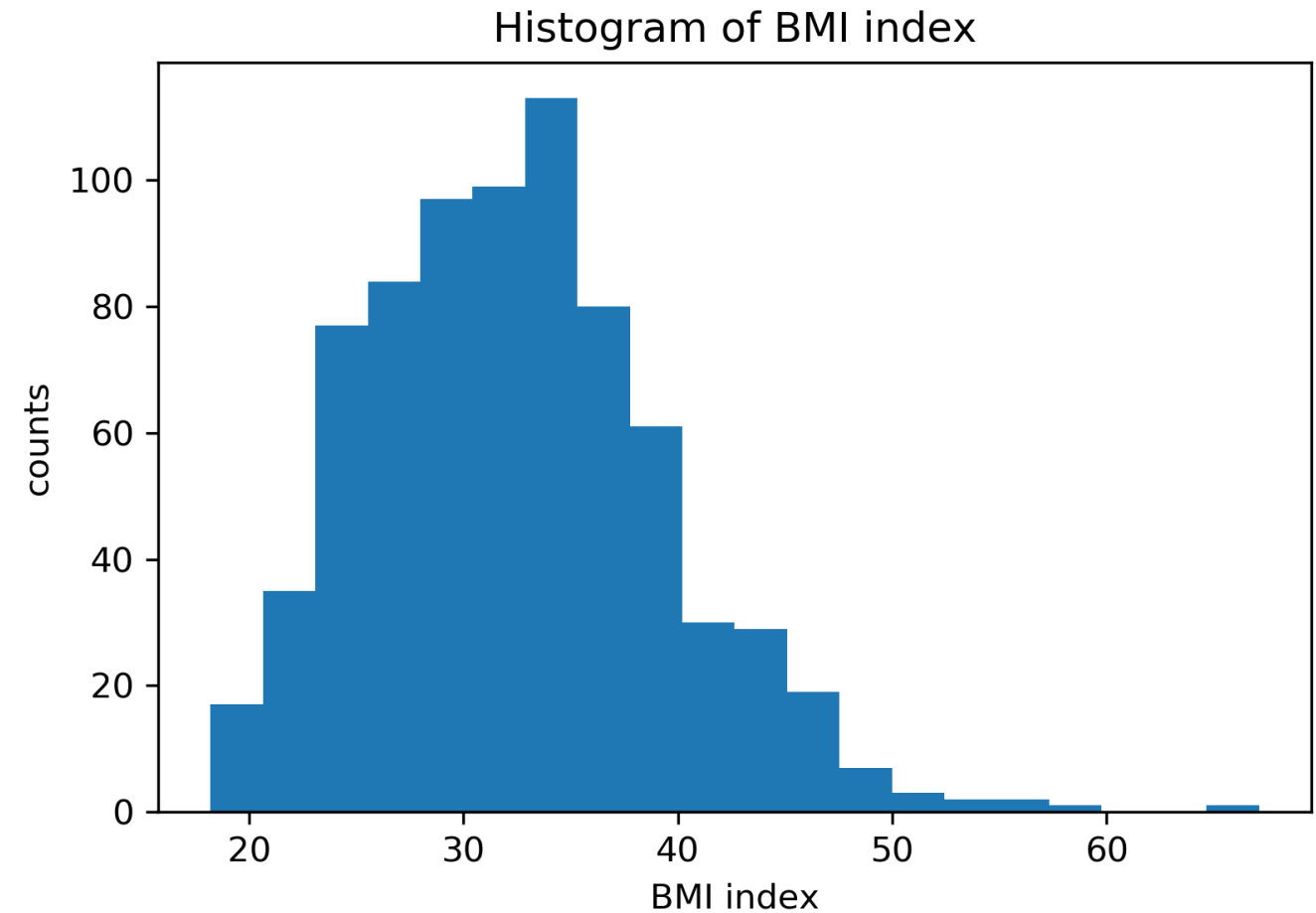
```
plt.hist(diabetes['bmi'], bins=20)
```

```
plt.title('Histogram of BMI index')
```

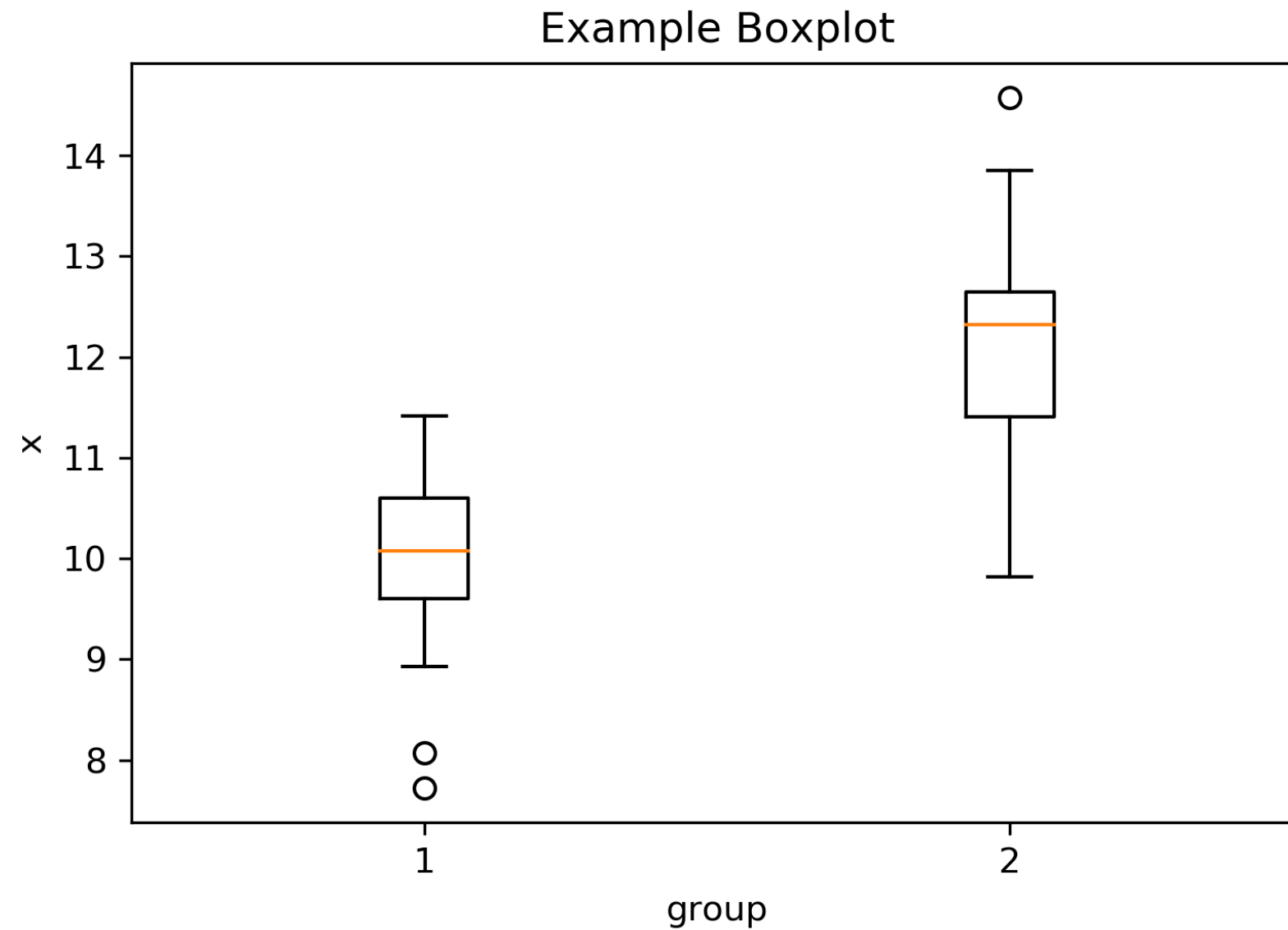
```
plt.xlabel('BMI index')
```

```
plt.ylabel('counts')
```

```
plt.show()
```



Boxplot

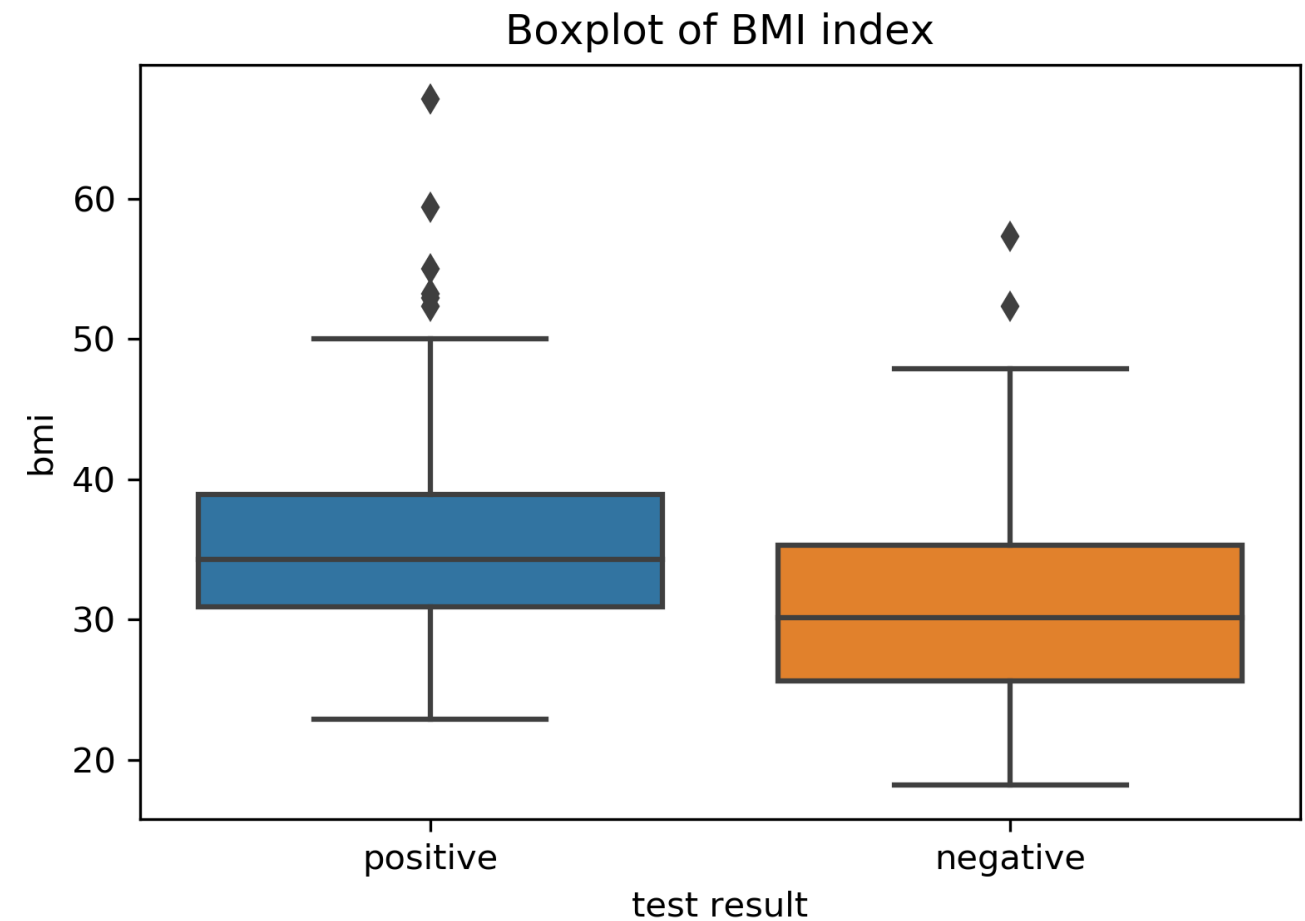


Create a boxplot

```
import seaborn as sns
```

```
sns.boxplot('test_result', 'bmi', data=diabetes)  
plt.title('Boxplot of BMI index')
```

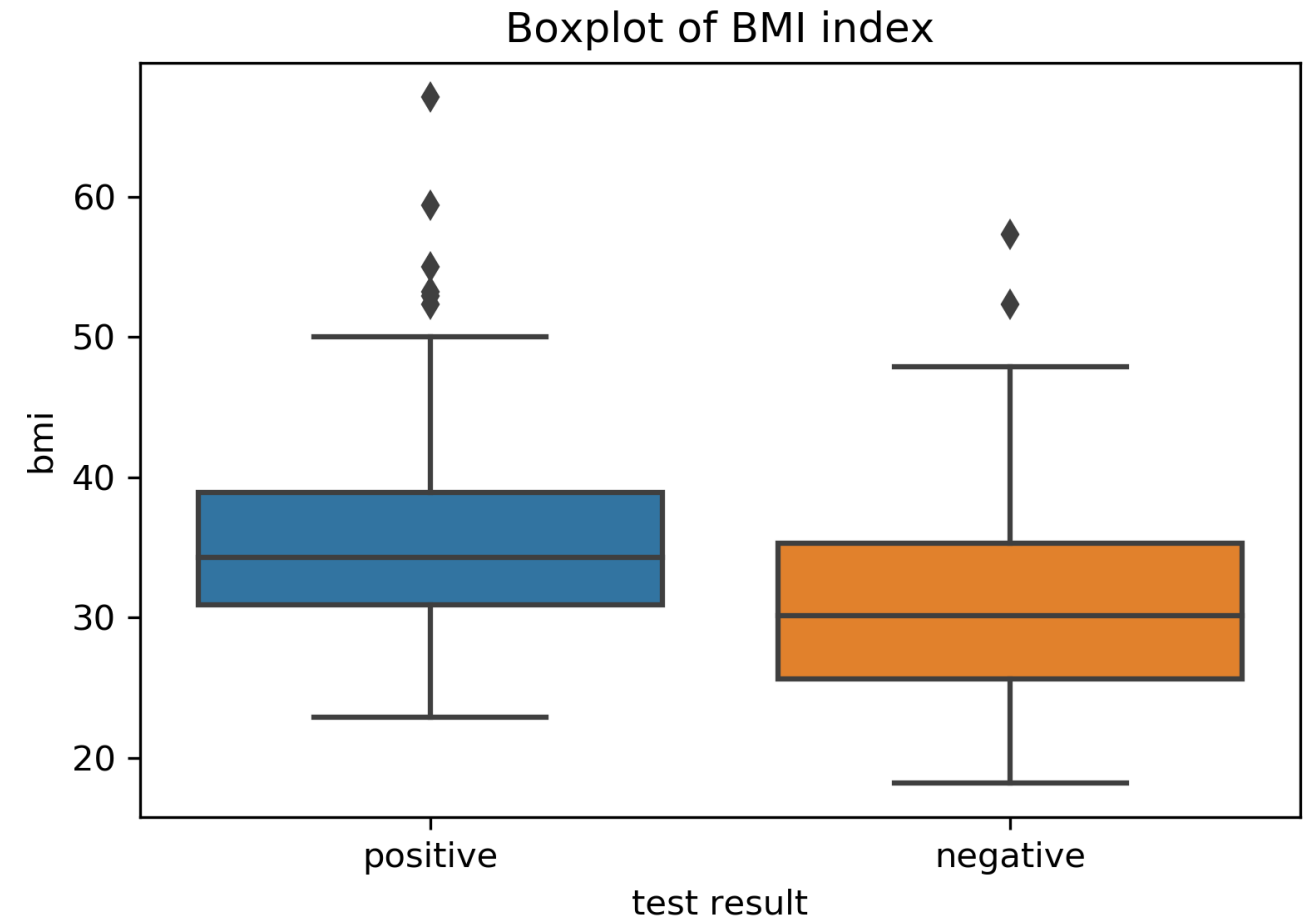
```
plt.show()
```



Create a boxplot

```
import seaborn as sns
```

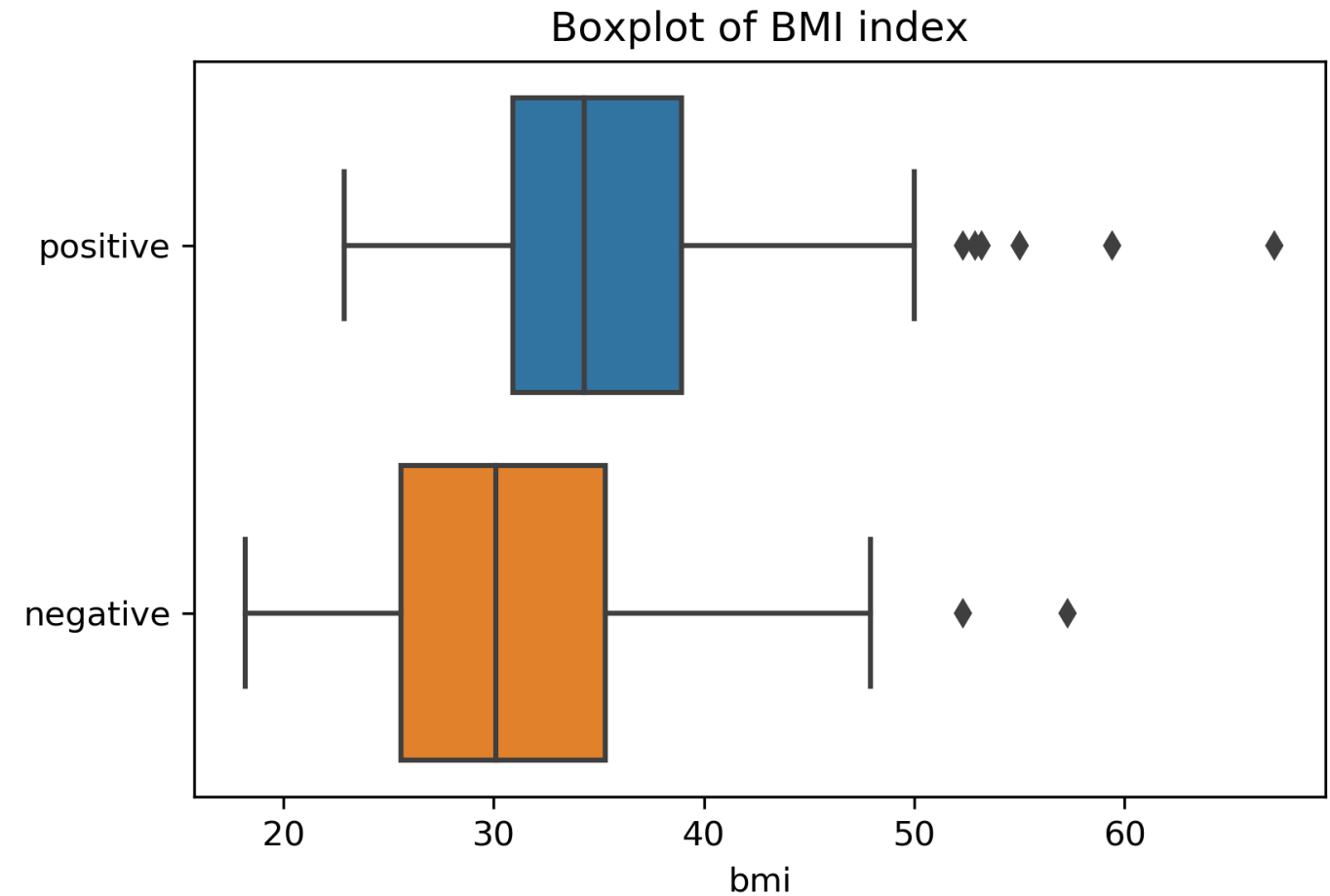
```
sns.boxplot(  
    x='test_result',  
    y='bmi',  
    data=diabetes  
)  
plt.title('Boxplot of BMI index')  
  
plt.show()
```



Create a boxplot

```
import seaborn as sns
```

```
sns.boxplot(  
    y='test_result',  
    x='bmi',  
    data=diabetes  
)  
plt.title('Boxplot of BMI index')  
  
plt.show()
```

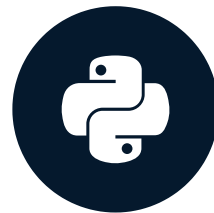


Let's practice!

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON

Final thoughts

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON



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Topics covered

- main data structures in Python
- string manipulation techniques
- iterable objects and their definition
- functions in Python
- NumPy arrays
- operations on DataFrames
- data visualization

Good luck!

PRACTICING CODING INTERVIEW QUESTIONS IN PYTHON