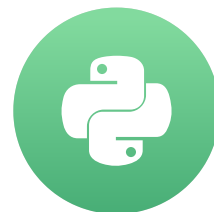


# What is the difference between a NumPy array and a list?

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON

**Kirill Smirnov**

Data Science Consultant, Altran



# 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('0'))
```

```
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]])
```

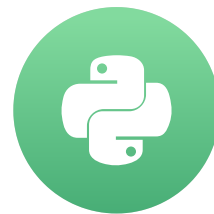
# Let's practice

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON

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

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON

**Kirill Smirnov**  
Data Science Consultant, Altran



# 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(scores_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	

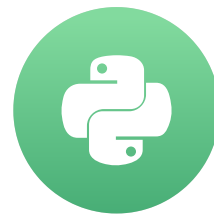
# Let's practice!

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON

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

PREPARING FOR 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].head(3))
    # 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].head(3))
    # 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!

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON

# How to visualize data in Python?

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON



**Kirill Smirnov**

Data Science Consultant, Altran

# matplotlib

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

- scatter plot
- histogram
- boxplot

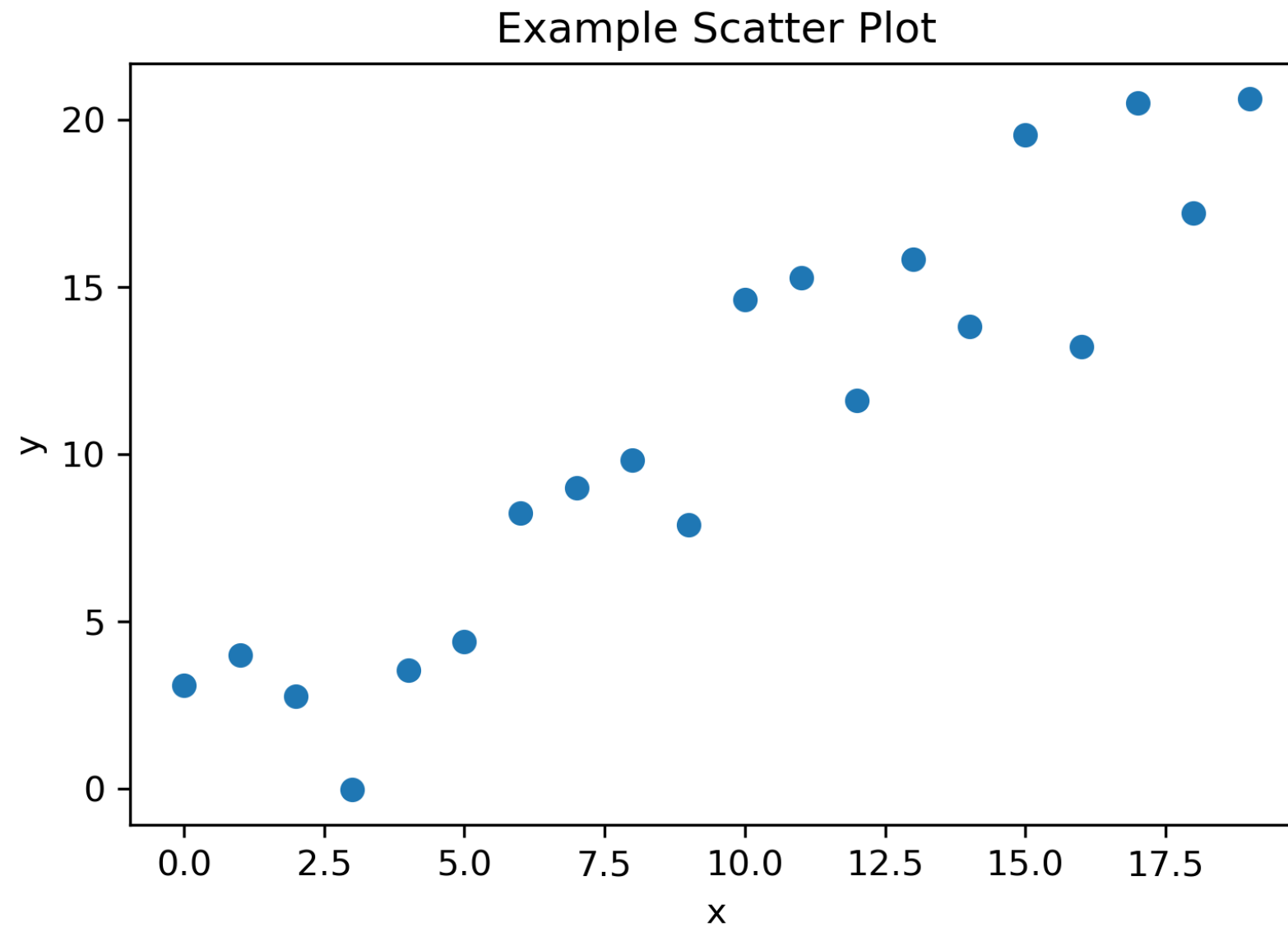
# Dataset

```
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

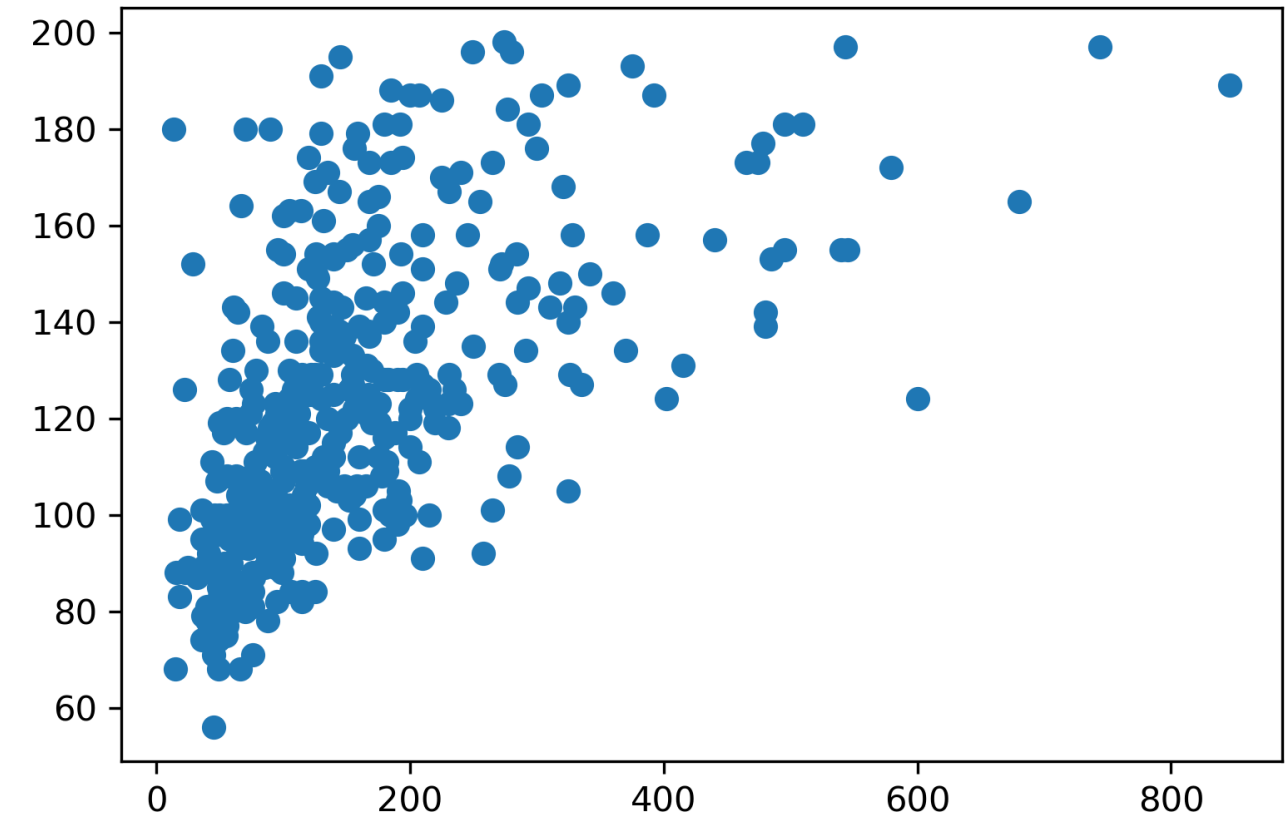
# Scatter plot



# Create a scatter plot

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

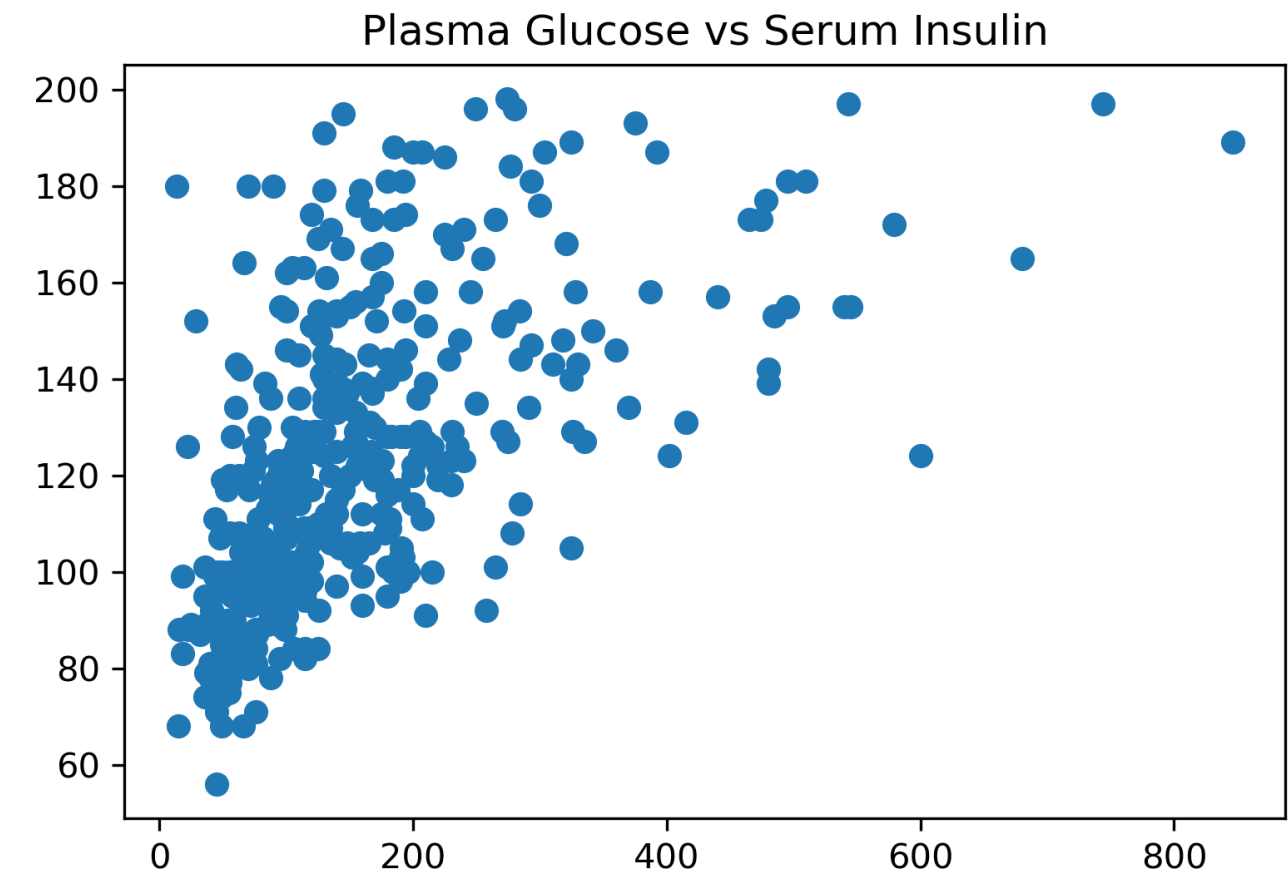
```
plt.scatter(  
    diabetes['serum insulin'],  
    diabetes['plasma glucose']  
)  
plt.show()
```



# Create a scatter plot

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

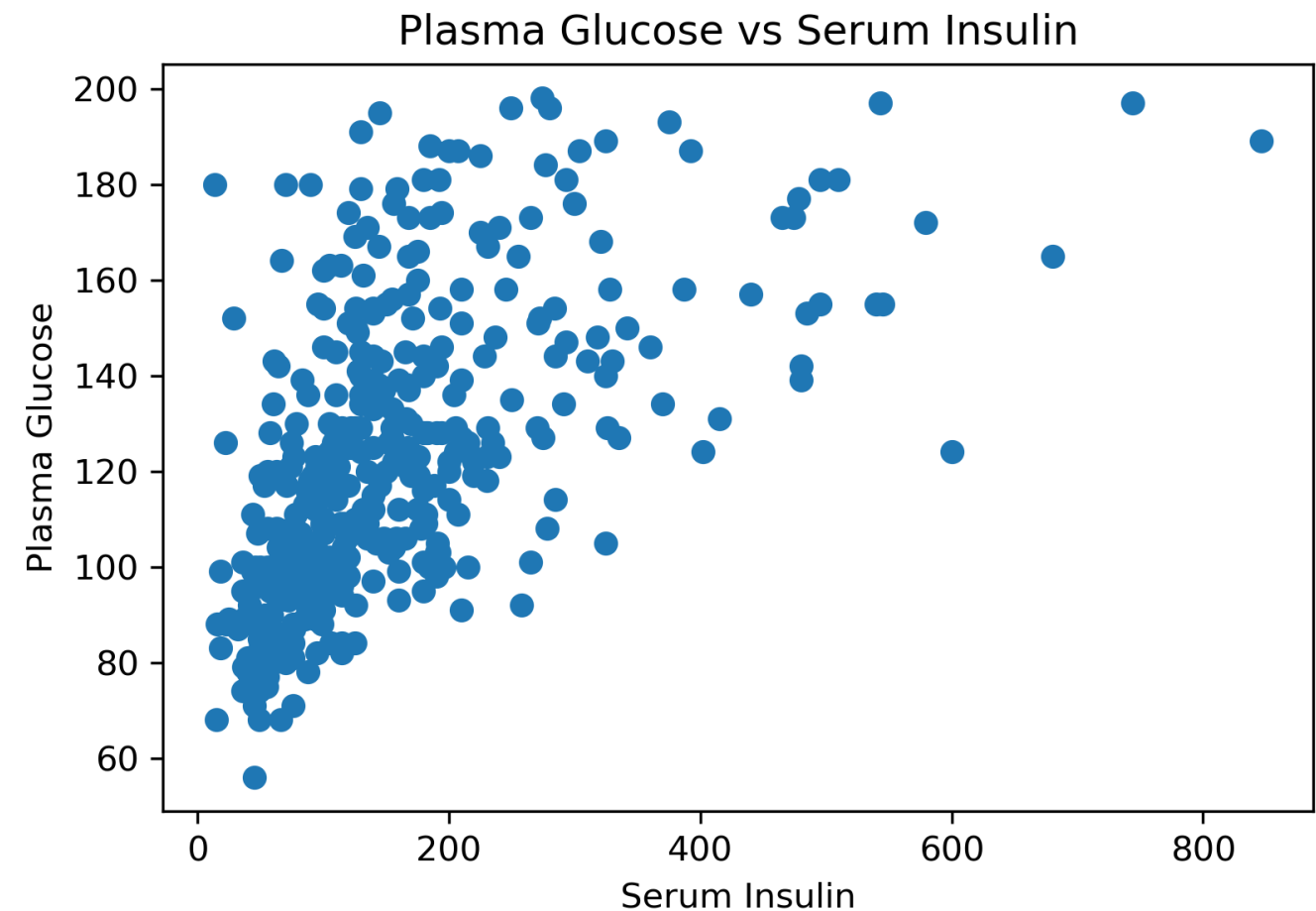
```
plt.scatter(  
    diabetes['serum insulin'],  
    diabetes['plasma glucose']  
)  
  
plt.title('Plasma Glucose vs Serum Insulin')  
  
plt.show()
```



# Create a scatter plot

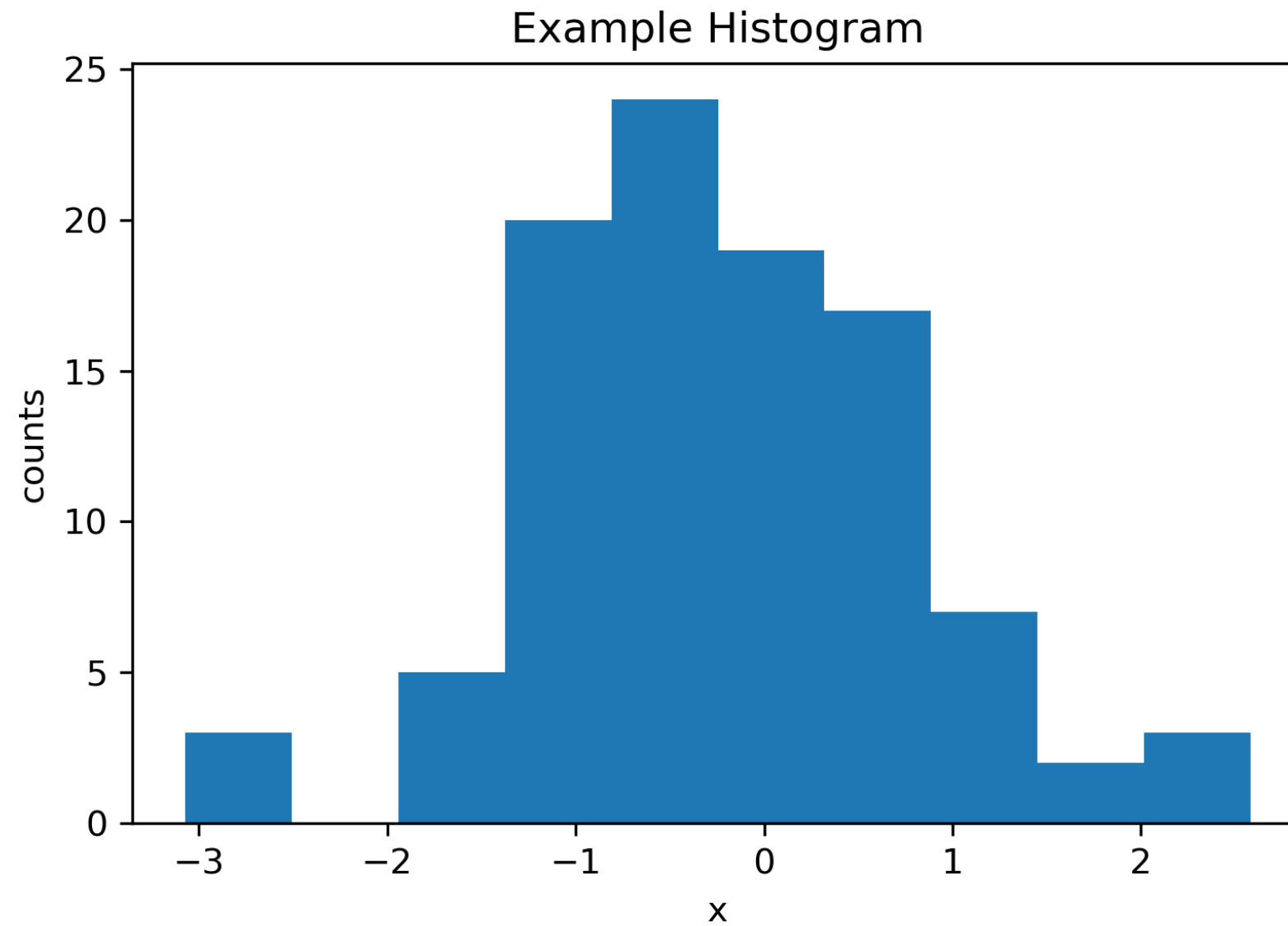
```
import matplotlib.pyplot as plt
```

```
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
```

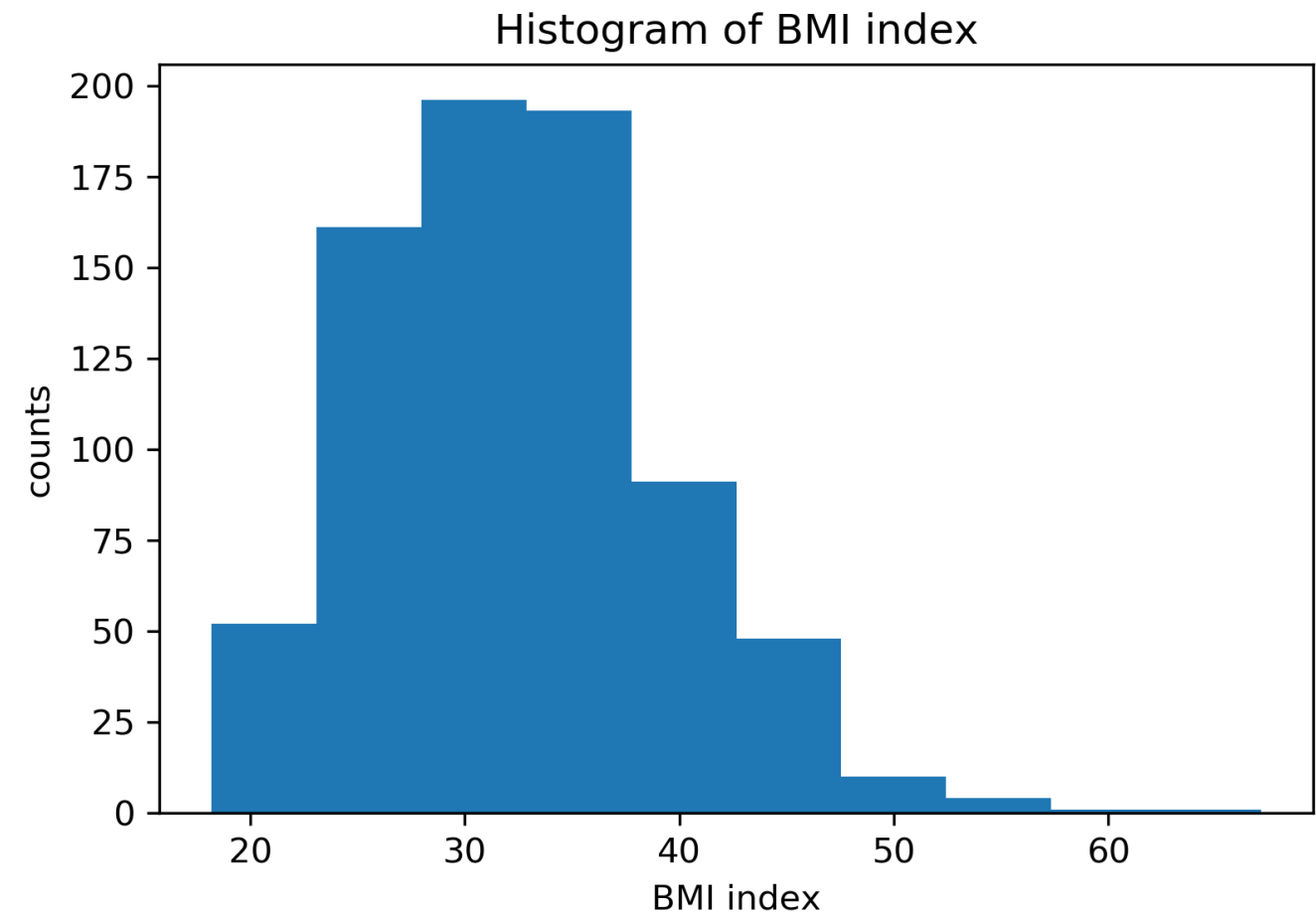
```
plt.hist(diabetes['bmi'])
```

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

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

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

```
plt.show()
```



# Create a histogram

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

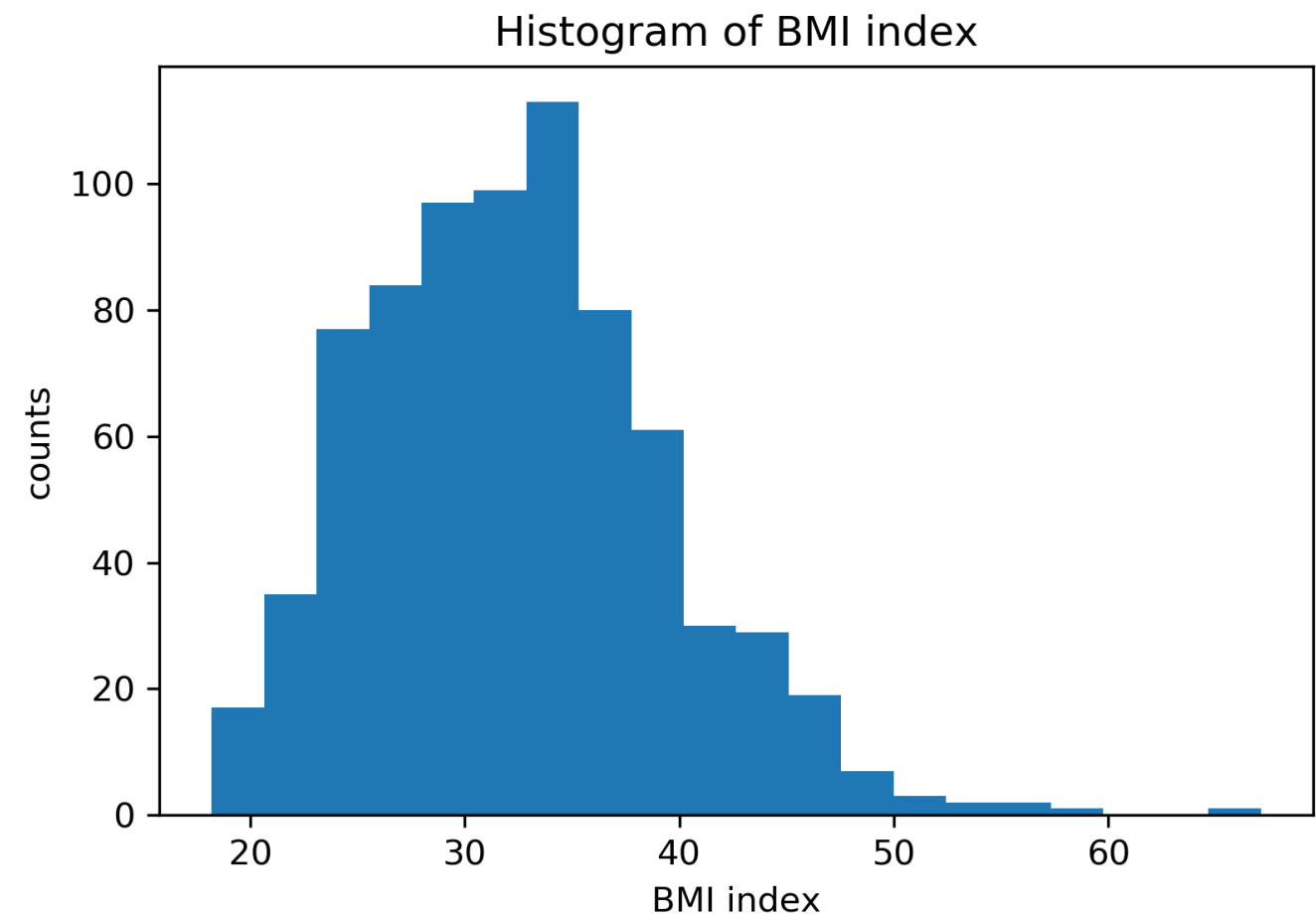
```
plt.hist(diabetes['bmi'], bins=20)
```

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

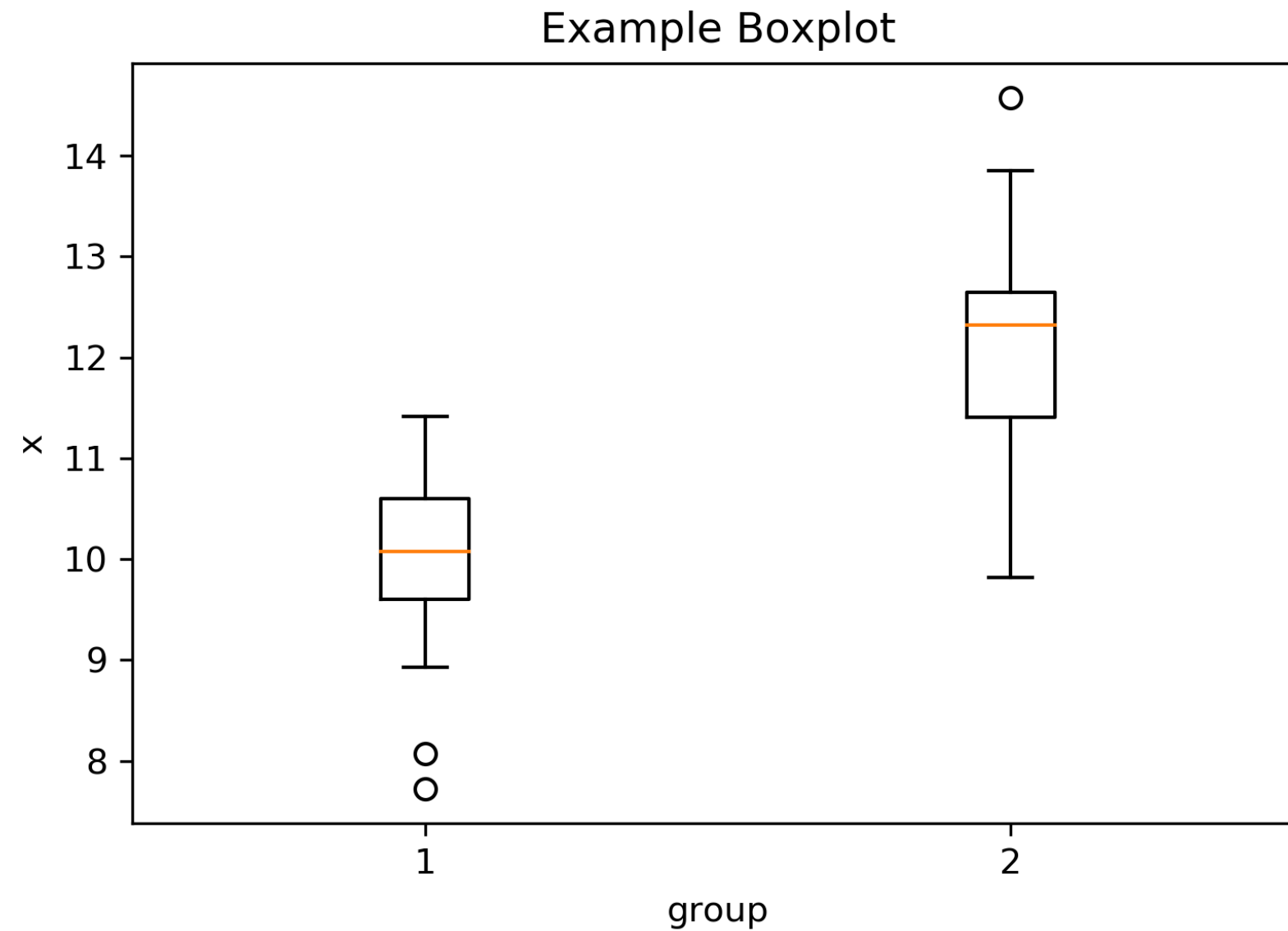
```
plt.xlabel('BMI index')
```

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

```
plt.show()
```



# Boxplot

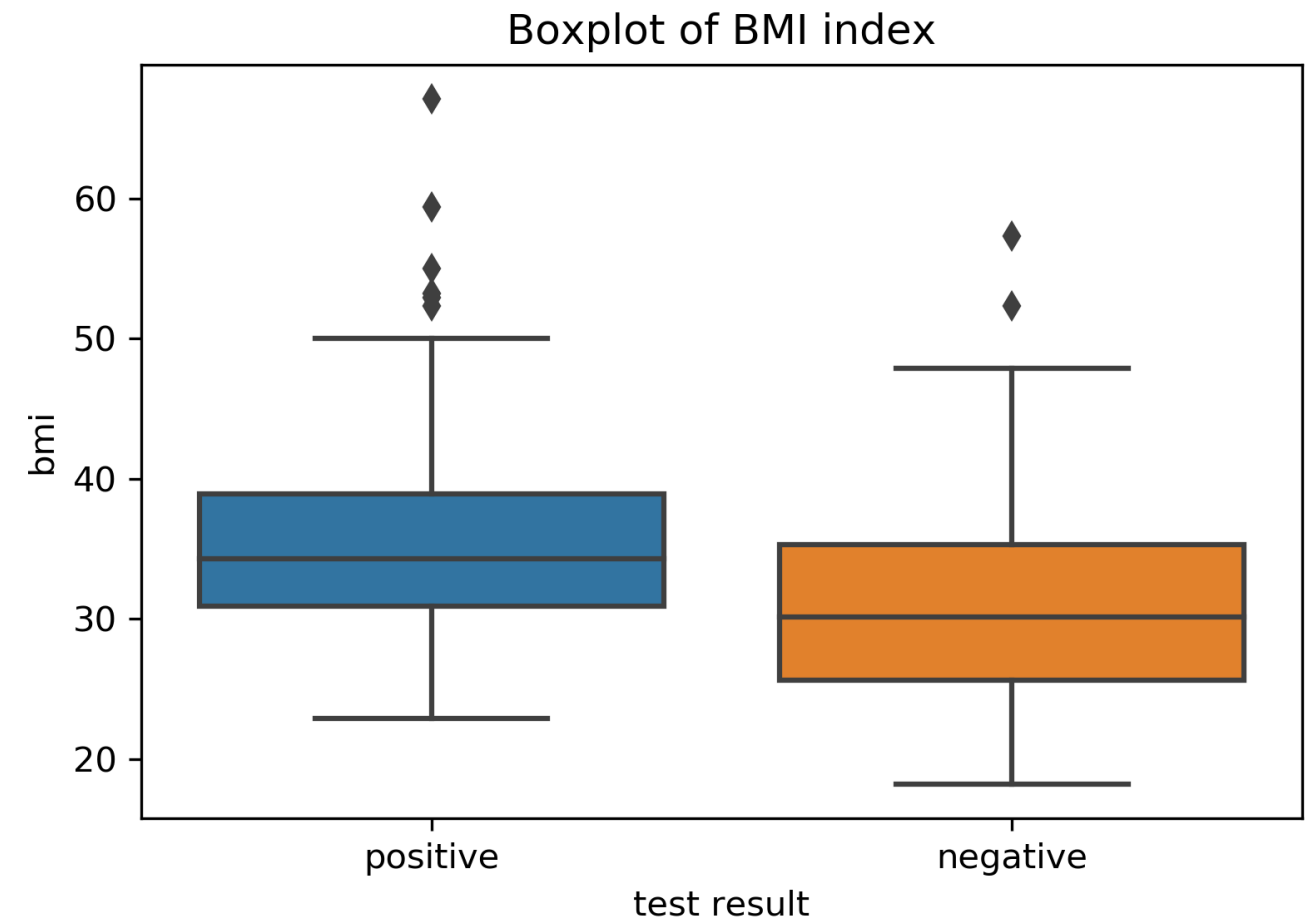


# Create a boxplot

```
import seaborn as sns
```

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

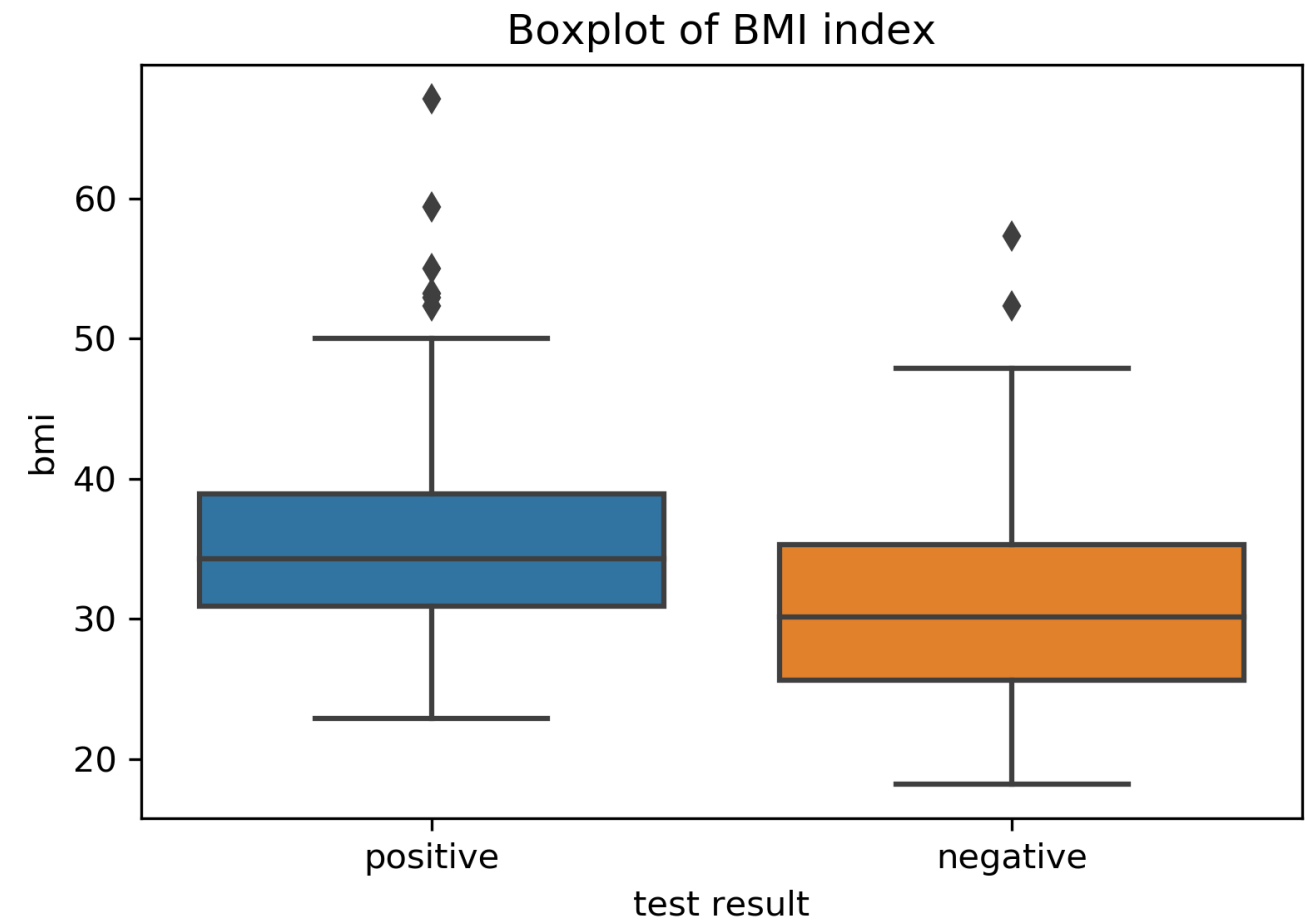
```
plt.show()
```



# Create a boxplot

```
import seaborn as sns
```

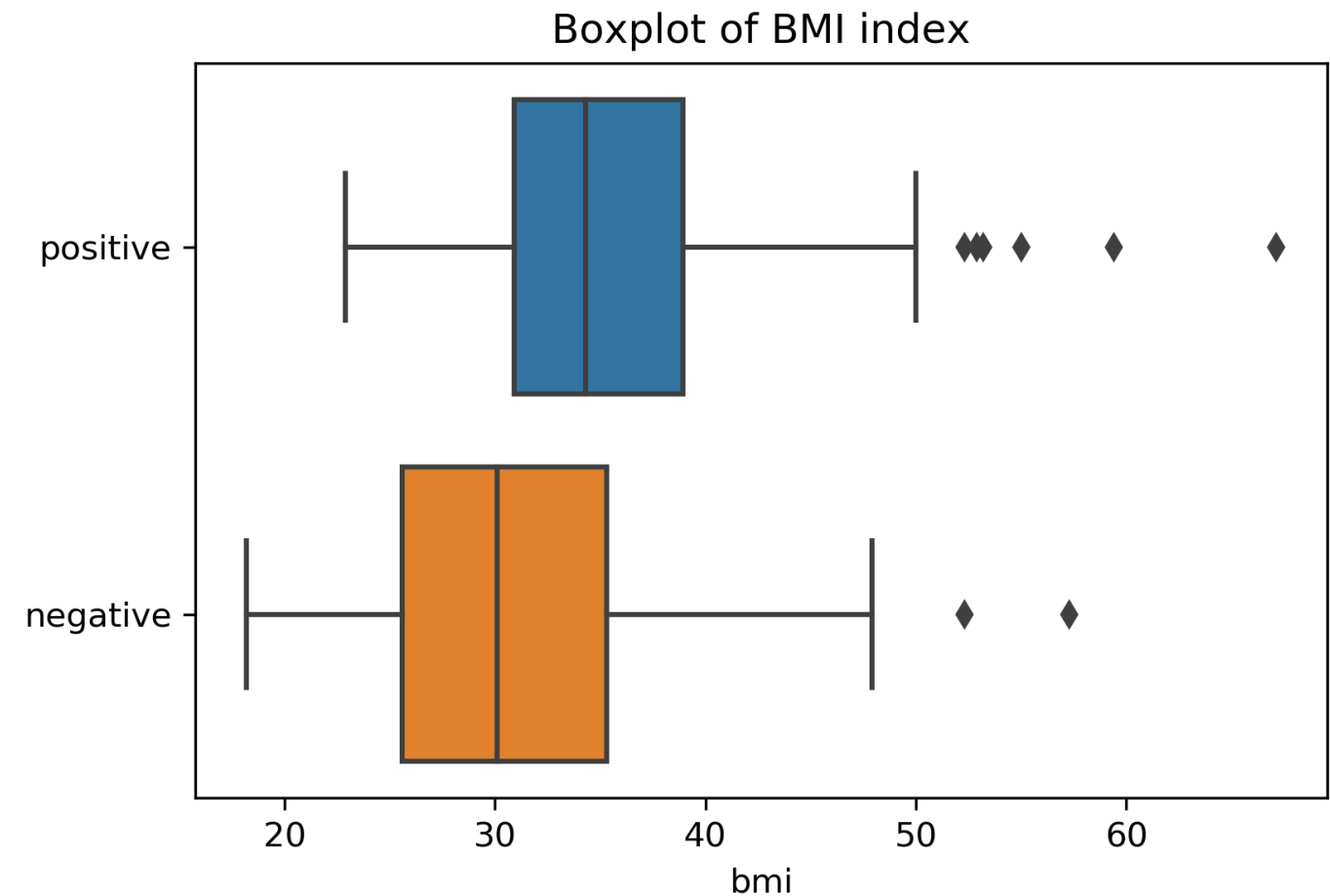
```
sns.boxplot(  
    x='test_result',  
    y='bmi',  
    data=diabetes  
)  
plt.title('Boxplot of BMI index')  
  
plt.show()
```



# 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!

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON



# Final thoughts

PREPARING FOR 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!

PREPARING FOR CODING INTERVIEW QUESTIONS IN PYTHON