6.2) Functions and Tables

Vitor Kamada

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Reference

Tables, Graphics, and Figures from

Computational and Inferential Thinking: The Foundations of Data Science

Adhikari & DeNero (2019): Ch 8 Functions and Tables

https://www.inferentialthinking.com/

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Function

Signature

- Calls to the function will look like this (with the same name and number of arguments). Example: double(3).
- When you call double, the argument can be any expression.
 (The name x doesn't affect calls.)
- In the body of the function, x is the name of the argument, as if the body included the code x = <the first argument>.

Our first function definition

def double(x):

""" Double :

return 2*x

Documentation ("docstring")

- Text that describes what the function does.
 - Can be any string, traditionally triple-quoted so it can span several lines.

 Traditionally, the first line describes what the function does, briefly.
- Subsequent lines can give more detail and examples.
- Running double? will show this text, just like max? will show the
 documentation for the built-in function max

Body

- All the code in here runs each time you call the function.
- The special statement **return** tells Python what the value of each call to this function is: it's the value of the expression after **return**.
- For example, the value of double(3) is 6. (Remember, when the argument is 3, it's like the body starts with x = 3.)
- Often, the body will have multiple lines of code that build up to computing the returned value. You can write any Python code here that you could write anywhere else.

Indentation

- Each line of code in the body is indented (that is, it's preceded by spaces).
- Traditionally, we use 2 or 4 spaces. They only need to be consistent.
- This tells Python that those lines are part of the body.
 The function's body ends at any unindented line.

40 1 40 1 4 = 1 4 = 1 7 9 0

Double Function

```
def double(x):
        Double x """
    return 2*x
double(7)
                      14
from datascience import *
double(make array(3, 4, 5))
             array([ 6, 8, 10])
```

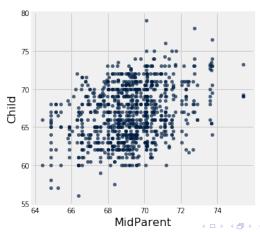
path_data = 'https://github.com/data-8/textbook/raw/gh-pages/data/
galton = Table.read_table(path_data + 'galton.csv')

family	father	mother	midparentHeight	children	childNum	gender	childHeight
1	78.5	67	75.43	4	1	male	73.2
1	78.5	67	75.43	4	2	female	69.2

MidParent	Child	
75.43	73.2	
75.43	69.2	

Midparent Height as Predictor Variable

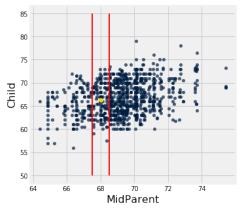
```
import matplotlib.pyplot as plt
plt.style.use('fivethirtyeight')
%matplotlib inline
heights.scatter(0) # column for x
```



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Predicting the Child's Height

```
heights.scatter('MidParent')
_ = plt.plot([67.5, 67.5], [50, 85], color='red', lw=2)
_ = plt.plot([68.5, 68.5], [50, 85], color='red', lw=2)
_ = plt.scatter(68, 66.24, color='gold', s=40)
```



close_to_68 = heights.where('MidParent', are.between(67.5, 68.5))

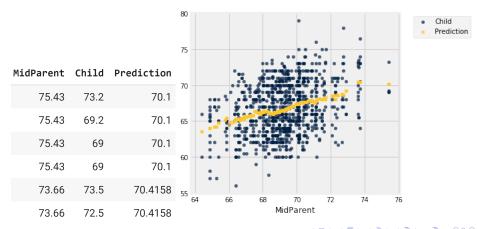
Child	MidParent	
62	68.44	
71.2	67.94	
67	67.94	

66.24045801526718

Function predict_child

```
def predict child(mpht):
    """Predict the height of a child whose
    parents have a midparent height of mpht.
    The prediction is the average height of
    the children whose midparent height is
    in the range mpht plus or minus 0.5.
    close points = heights.where('MidParent',
            are.between(mpht-0.5, mpht + 0.5))
    return close points.column('Child').mean()
predict child(68)
                              66.24045801526718
predict child(74)
                              70.41578947368421
```

heights_with_predictions.scatter('MidParent')



https://data.ca.gov/dataset/ca-educational-attainment-personal-income

Population Count	Personal Income	Educational Attainment	Gender	Age	Year
2058	H: 75,000 and over	No high school diploma	Female	18 to 64	1/1/14 0:00
2153	H: 75,000 and over	No high school diploma	Male	65 to 80+	1/1/14 0:00
4666	G: 50,000 to 74,999	No high school diploma	Female	65 to 80+	1/1/14 0:00

Sum Up the Population Count

```
educ inc = ca 2014.select('Educational Attainment',
            'Personal Income', 'Population Count')
education = educ inc.select('Educational Attainment',
                            'Population Count')
educ totals = education.group('Educational Attainment', sum)
```

Educational Attainment Population Count sum

Bachelor's degree or higher	8525698
College, less than 4-yr degree	7775497
High school or equivalent	6294141
No high school diploma	4258277

```
def percents(array_x):
    return np.round( (array_x/sum(array_x))*100, 2)
```

High school or equivalent

No high school diploma

```
import numpy as np
educ_distribution = educ_totals.with_column(
    'Population Percent', percents(educ_totals.column(1))
)
Educational Attainment Population Count sum Population Percent
Bachelor's degree or higher 8525698 31.75
College, less than 4-yr degree 7775497 28.96
```

23.44

15.86

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6294141

4258277

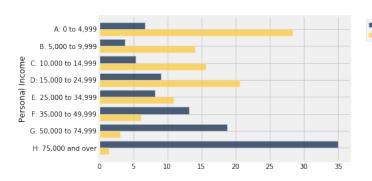
totals = educ_inc.pivot('Educational Attainment',
 'Personal Income', values='Population Count', collect=sum)

Personal Income	Bachelor's degree or higher	College, less than 4-yr degree
A: 0 to 4,999	575491	985011
B: 5,000 to 9,999	326020	810641
C: 10,000 to 14,999	452449	798596

```
distributions = totals.select(0).with_columns(
    "Bachelor's degree or higher", percents(totals.column(1)),
    'College, less than 4-yr degree', percents(totals.column(2)),
    'High school or equivalent', percents(totals.column(3)),
    'No high school diploma', percents(totals.column(4))
)
```

Personal Income	Bachelor's degree or higher	College, less than 4-yr degree
A: 0 to 4,999	6.75	12.67
B: 5,000 to 9,999	3.82	10.43
C: 10,000 to 14,999	5.31	10.27

distributions.select(0, 1, 4).barh(0)



Bachelor's degree or higher
 No high school diploma

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