



**BEDROCK
DATA SCIENCE**



BROS

GAME TIME



In [48]: `footballers = {'ronaldo':7,'messi':10,'gerrard':8}`



Out[104]: `[('messi', 10), ('gerrard', 8), ('ronaldo', 7)]`

Options

- A. `sorted(footballers.items(),key = lambda x: x[0])`
- B. `sorted(footballers.items(),key = lambda x:-x[0])`
- C. `sorted(footballers.items(),key = lambda x: x[1])`
- D. `sorted(footballers.items(),key = lambda x: -x[-1])`

In [48]: `footballers = {'ronaldo':7,'messi':10,'gerrard':8}`



Out[104]: `[('messi', 10), ('gerrard', 8), ('ronaldo', 7)]`

Options

- A. `sorted(footballers.items(),key = lambda x: x[0])`
- B. `sorted(footballers.items(),key = lambda x:-x[0])`
- C. `sorted(footballers.items(),key = lambda x: x[1])`
- D. `sorted(footballers.items(),key = lambda x: -x[-1])`



```
In [47]: def reverse_dictionary(**kwargs):
....:     return {v:k for k,v in kwargs.items()}
....:
In [48]: footballers = {'ronaldo':7,'messi':10,'gerrard':8}

Out[49]: {7: 'ronaldo', 10: 'messi', 8: 'gerrard'}
```

Options

- A. reverse_dictionary(footballers)
- B. reverse_dictionary(*footballers)
- C. reverse_dictionary(**footballers)
- D. reverse_dictionary(***footballers)

```
In [47]: def reverse_dictionary(**kwargs):  
....:     return {v:k for k,v in kwargs.items()}  
....:  
In [48]: footballers = {'ronaldo':7,'messi':10,'gerrard':8}  
  
Out[49]: {7: 'ronaldo', 10: 'messi', 8: 'gerrard'}
```

Options

- A. reverse_dictionary(footballers)
- B. reverse_dictionary(*footballers)
- C. reverse_dictionary(**footballers)
- D. reverse_dictionary(***footballers)



```
In [66]: def swiggy_orders(order,foodlist):  
....:     foodlist.append(order)  
....:     return foodlist
```

```
In [68]: cart = ['chicken wings','glass noodles']
```

```
In [69]: new_cart = swiggy_orders('butter chicken',cart)
```

Options

- A. cart = ['chicken wings', 'glass noodles', 'butter chicken']
new_cart = ['chicken wings', 'glass noodles', 'butter chicken']
- B. cart = ['chicken wings', 'glass noodles']
new_cart = ['chicken wings', 'glass noodles', 'butter chicken']
- C. cart = ['chicken wings', 'glass noodles']
new_cart = ['chicken wings', 'glass noodles']
- D. cart = ['chicken wings', 'glass noodles', 'butter chicken']
new_cart = ['chicken wings', 'glass noodles']



```
In [66]: def swiggy_orders(order,foodlist):  
....:     foodlist.append(order)  
....:     return foodlist
```

```
In [68]: cart = ['chicken wings','glass noodles']
```

```
In [69]: new_cart = swiggy_orders('butter chicken',cart)
```

Options

A. cart = ['chicken wings', 'glass noodles', 'butter chicken']
new_cart = ['chicken wings', 'glass noodles', 'butter chicken']

B. cart = ['chicken wings', 'glass noodles']
new_cart = ['chicken wings', 'glass noodles', 'butter chicken']

C. cart = ['chicken wings', 'glass noodles']
new_cart = ['chicken wings', 'glass noodles']

D. cart = ['chicken wings', 'glass noodles', 'butter chicken']
new_cart = ['chicken wings', 'glass noodles']

Lambda function

(SYNTAX) `square = lambda x : x**2`

- Lambda functions (also called **Anonymous functions**) are a concise way to declare functions.
- This pythonic way of defining functions avoids the keyword **RETURN**
- Very helpful when working with mathematical function implementation.

```
In [89]: square = lambda x: x**2
```

```
In [90]: square(2)  
Out[90]: 4
```

```
In [91]: square(2,2)
```

```
-----  
-----  
TypeError  
back (most recent call last)  
<ipython-input-91-fc957e4c04b0> in <module>  
----> 1 square(2,2)
```

```
TypeError: <lambda>() takes 1 positional argument but 2 were given
```

Use lambda functions

The `list.sort()` and `sorted()` methods can take lambda functions

```
#Example of Sort function(sort in age and in potential)
lst = [('De Ligt',19,92.1), ('Alexander-Arnold',20,90.5),
        ('Sancho',19,94.0),('Havertz',20,93.2),('Rodrygo',18,91.4)]

lst.sort(key=lambda x:x[1]) #Sort by age
print(lst)
lst.sort(key=lambda x:x[2]) #Sort by potential
print(lst)
>>>
[('Rodrygo', 18, 91.4), ('De Ligt', 19, 92.1), ('Sancho', 19, 94.0), ('Alexander-Arnold',
20, 90.5), ('Havertz', 20, 93.2)]

[('Alexander-Arnold', 20, 90.5), ('Rodrygo', 18, 91.4), ('De Ligt', 19, 92.1), ('Havertz',
20, 93.2), ('Sancho', 19, 94.0)]
```



It is NOT compulsory to use a lambda function. Any regular function would do as well

Use lambda functions

- Certain methods/functions exist which make it very easy to apply lambda functions on elements of iterable objects.
- map(), filter() both take a function as an input. (see e.g. below)

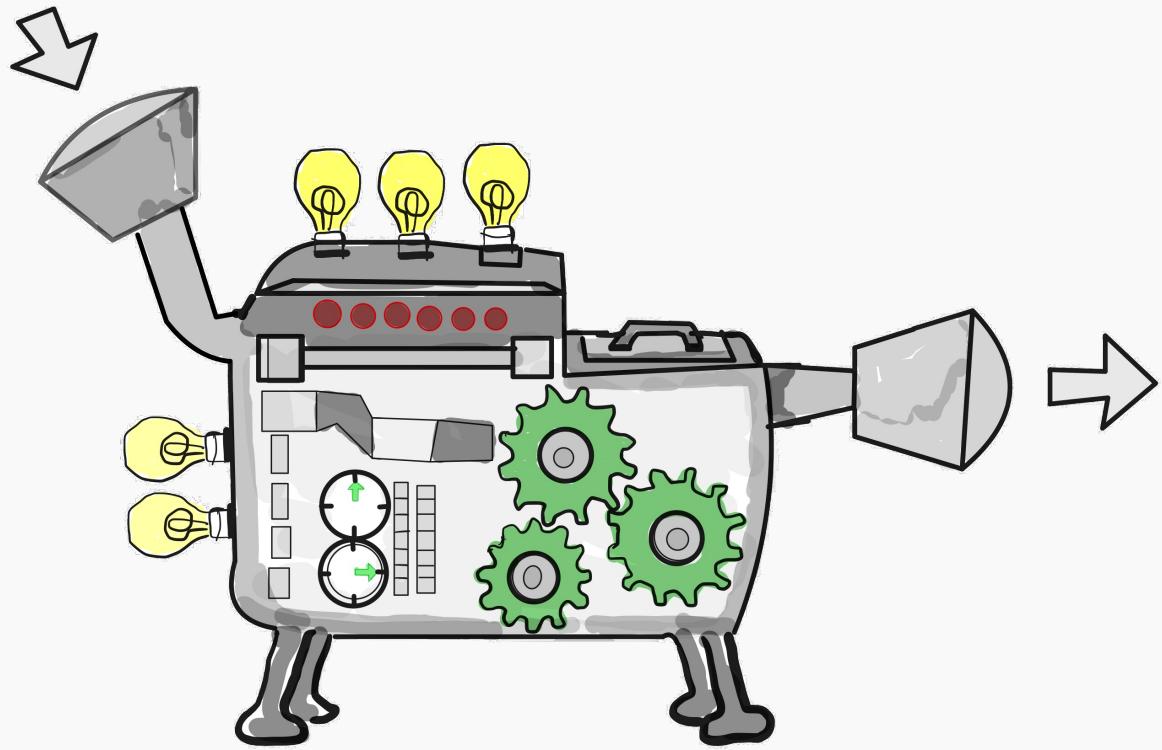
```
## Usage of lambda functions

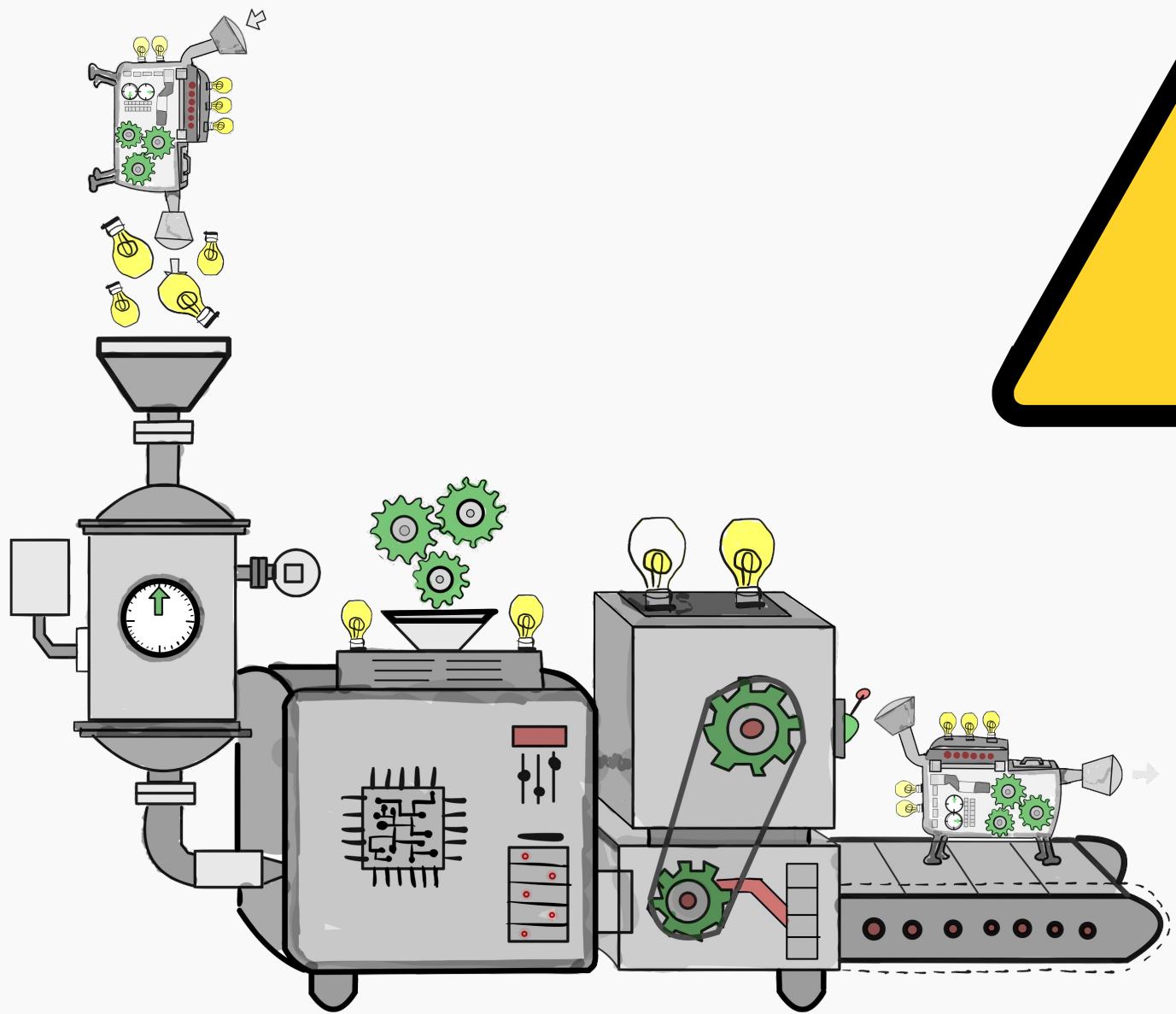
#Example of Map function(calculate discount)
a = list(map(lambda x: x - 0.2*x, [120,300,150,3000,42.55,67.11]))

#Example of Filter function(get all 'back' positions)
b = list(filter(lambda x: 'B' in x, ['LB', 'CB', 'ST','CAM','CMD','GK','RB']))

print(a)
print(b)
>>>
[96.0, 240.0, 120.0, 2400.0, 34.04, 53.688]
['LB', 'CB', 'RB']
```

Advanced Functions

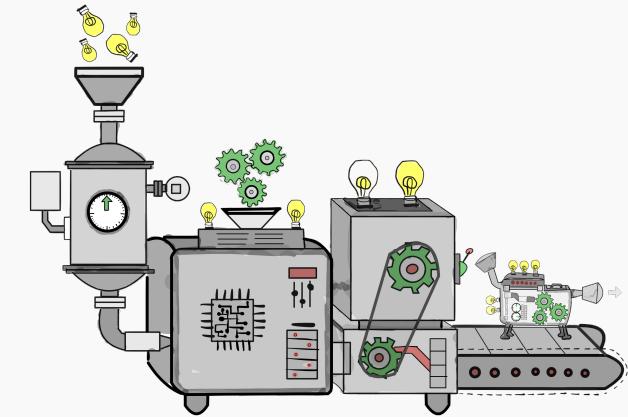
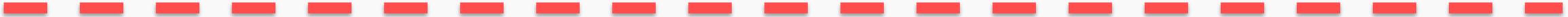




Advanced functions

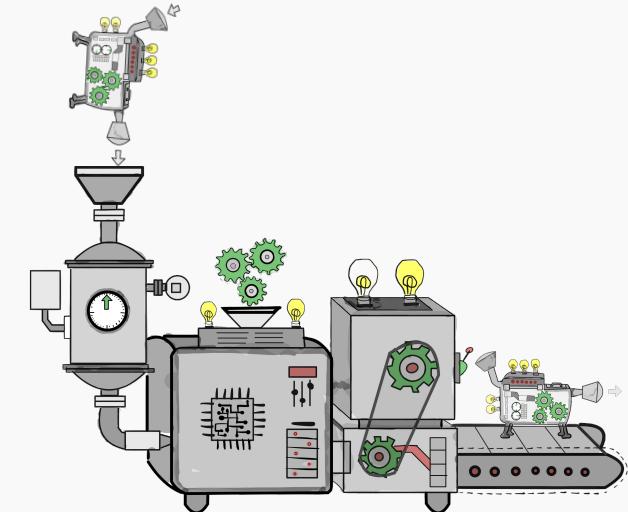
Level #1

Number in, Function out



Level #2

Function in, Function out



Level #1 - Number in, function out

- Combine a lambda function and use it inside a function to declare an even higher level of generalization.
- It means you are now returning a function, the specifics of which are decided by the input to higher function.

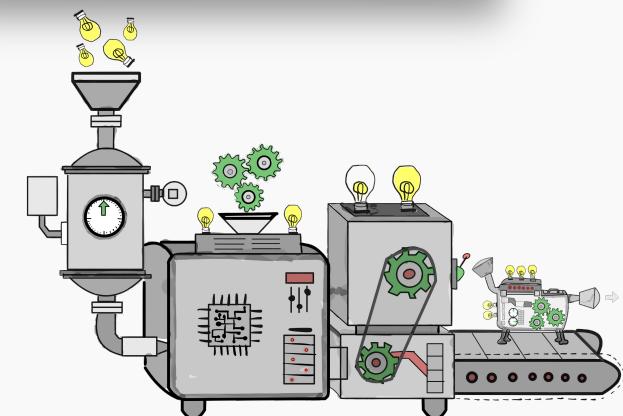
You are returning a function here

```
def multiplier(n):  
    func = lambda x:x*n  
    return func
```

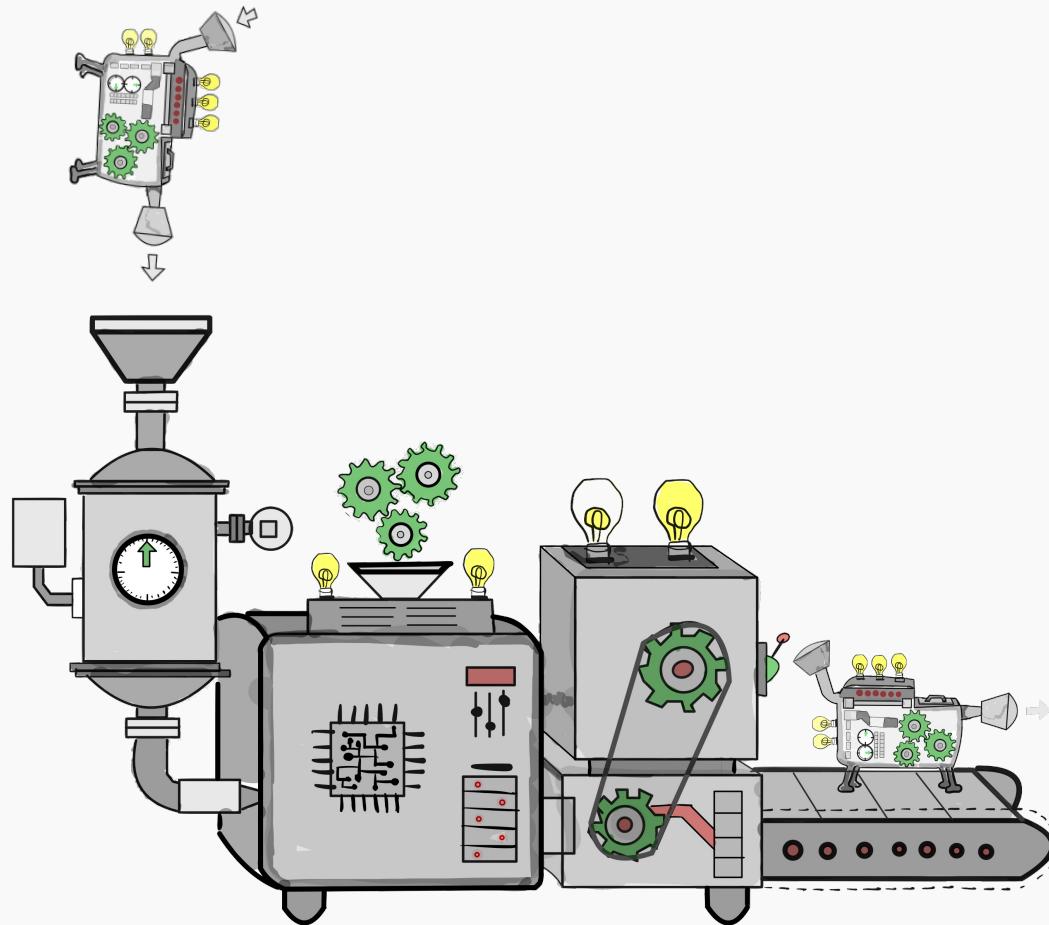
```
doubler = multiplier(2)  
tripler = multiplier(3)
```

```
print(doubler(100))  
print(tripler(100))  
=>  
200  
300
```

It is an n-multiplier function. Based on input n the output function will vary



Level #2: Function in, Function out



Level #2 – Function in, Function out

```
In [19]: def decorator_function(original_function):
....:     def packaging(x):
....:         print("Tao's function. All rights reserved")
....:         return original_function(x)
....:     return packaging
```

Taking a function as input

You are returning a
function here

Level #2 – Function in, Function out

```
In [19]: def decorator_function(original_function):
....:     def packaging(x):
....:         print("Tao's function. All rights reserved")
....:         return original_function(x)
....:     return packaging
```

```
In [21]: def square(x):
....:     return x**2
```

```
In [27]: square(5)
```

```
Out[27]: 25
```

unprocessed
function call

Level #2 – Function in, Function out

Remember!
decorator_function()
returns a function



```
In [19]: def decorator_function(original_function):
....:     def packaging(x):
....:         print("Tao's function. All rights reserved")
....:         return original_function(x)
....:     return packaging

In [21]: def square(x):
....:     return x**2

In [27]: square(5)
Out[27]: 25

In [22]: processed_square = decorator_function(square)
```

after processing using the
decorator function

Level #2 – Function in, Function out

```
In [19]: def decorator_function(original_function):
....:     def packaging(x):
....:         print("Tao's function. All rights reserved")
....:         return original_function(x)
....:     return packaging
```

```
In [21]: def square(x):
....:     return x**2
```

```
In [27]: square(5)
```

```
Out[27]: 25
```

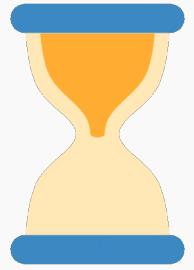
Function call includes the
copyright message

```
In [22]: processed_square = decorator_function(square)
```

```
In [23]: processed_square(5)
```

Tao's function. All rights reserved

```
Out[23]: 25
```



Digestion Time

Function Decorators

```
In [19]: def decorator_function(original_function):
....:     def packaging(x):
....:         print("Tao's function. All rights reserved")
....:         return original_function(x)
....:     return packaging
```

Without decorator

```
square = lambda x:x**2
```



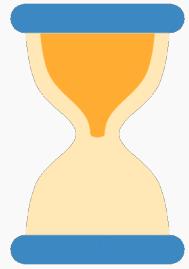
```
square_sum = decorator_function(square)
```

Using decorator

```
@decorator_function
def square(x):
    return x**2
```



The function you declared
below the decorator gets
written over. Now it no longer
takes only one input



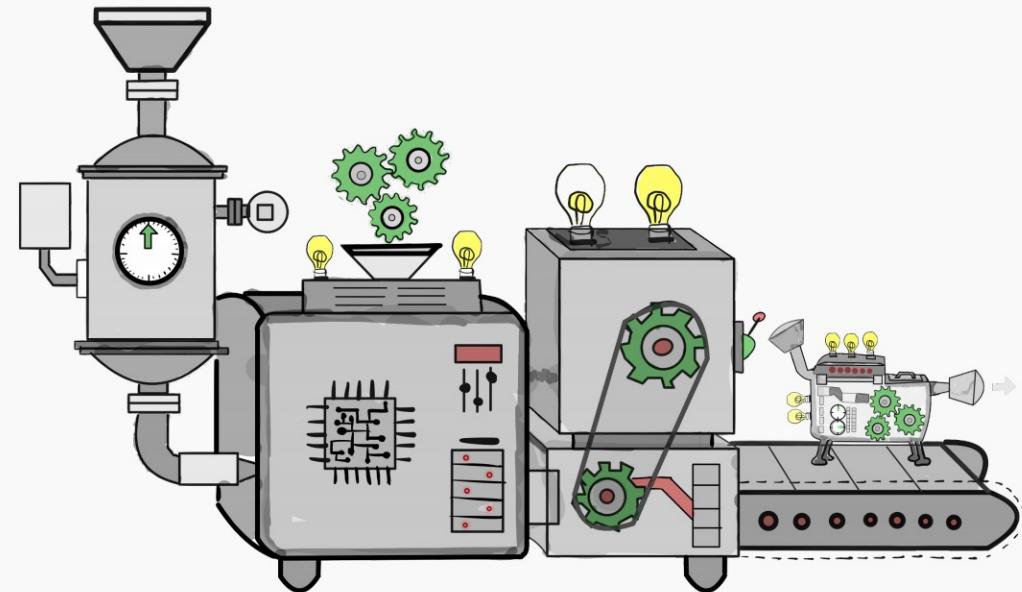
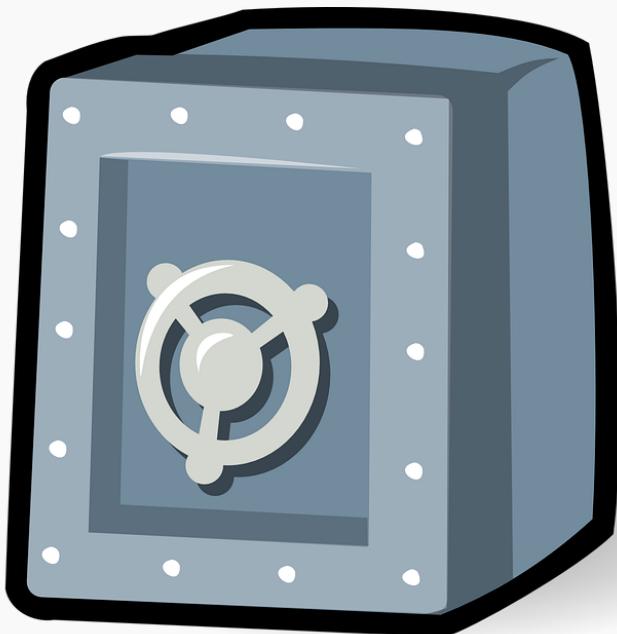
Digestion Time

Session 4: Python Classes

Session 3

```
# Variable assignments  
>>> x = 5  
>>> y = x **(2)  
>>> z = x+y
```

```
# functions  
>>> def myfunc():  
.....  
.....  
    return ...
```



Demo

What will happen if I run the following?

```
Hi = 5.2324
```

```
Hi.__floordiv__(2)
```

Everything in python is a class object instance

```
# Variables  
>>> x = 5  
>>> y = x **(2)
```

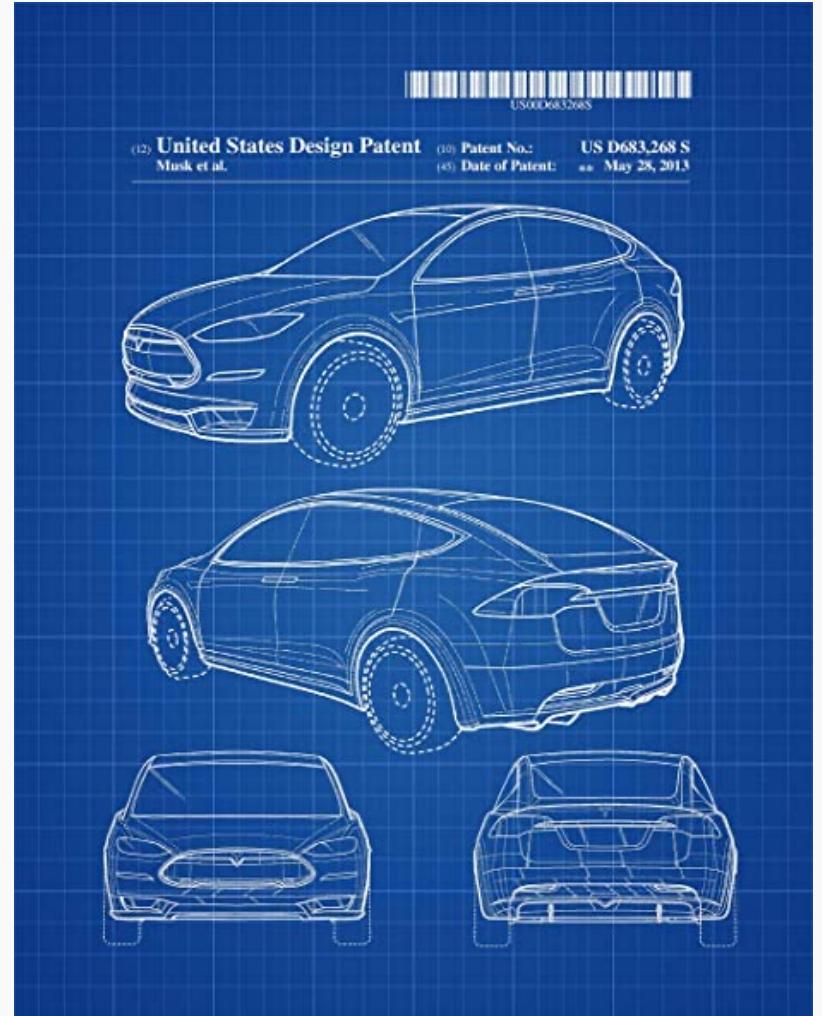
```
# functions  
>>> def myfunc():  
....  
....  
return ...
```

```
# classes  
In [18]: class myclass():  
....:  
....:     def  
....:         __init__(self):  
....:             self.x = 5  
....:         ....
```



Lets start with traditional analogies.

The class template is often called a "blueprint"



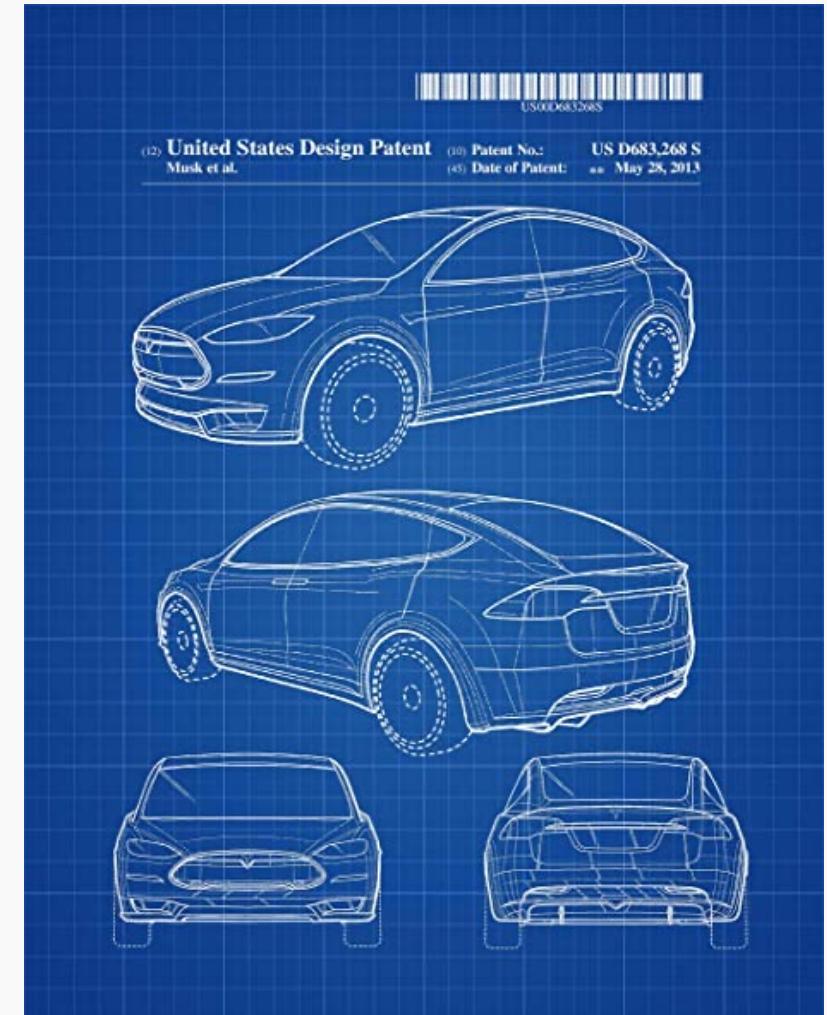
Lets start with traditional analogies.

The class template is often called a "blueprint"
or a **recipe**



Lets start with traditional analogies.

```
# Create a class template
class Model_S:
    year = 2013
    <statement 2>
    [...]
    <statement n>
```



Lets start with traditional analogies.

We **instantiate** the class object **instance** with function notation

```
# Create a class template
class Model_S:
    Year = 2013
    <statement 2>
    [...]
    <statement n>

# Create a class instance
my_tesla = Model_S()
```



Classes vs Instances

```
class Cookie:  
    """A generic cookie class.  
    cal_per_gram = {"sugar":  
                    "co":  
                    "cho":  
  
    def __init__(self, cookie_type, se:  
  
        @staticmethod  
        def calc_cal(ingredient_d:  
            total_calories = 0  
            for ingredient, amount:  
                total_calories +=  
            return total_calories  
  
class Chocolate_chip(Cookie):  
    """A chocolate chip cookie.  
    def __init__(self, chocola:  
        self.cookie_type = "choc":  
        self.num_choc_chips =  
  
        #ingredients are in grams  
        self.ingredients = {  
            "flour": 150,  
            "brown_sugar": 100,  
            "white_sugar": 50,  
            "baking_soda": 1,  
            "baking_powder": 1,  
            "salt": 1,  
            "chocolate_chips": 100,  
            "mms": 100,  
            "butter": 100,  
            "oil": 100,  
            "egg": 1,  
            "vanilla": 1,  
            "milk": 100  
        }  
  
        self.num_cookies = 16  
  
    def __str__(self):  
        total_calories = self.cal_per_cookie * self.ingredients["cal_per_gram"] * self.num_choc_chips  
        cal_per_cookie = round(total_calories / self.num_cookies)  
        cchips = self.ingredients["chocolate_chips"] / self.num_cookies  
        return 'I am a {} with {} chocolate chips and {} calories per cookie'.format(self.cookie_type, cchips, cal_per_cookie)
```



```
plain_cookie = Chocolate_chip(chocolate_chips = 0)  
normal_cookie = Chocolate_chip(chocolate_chips = 5)  
my_cookie = Chocolate_chip(chocolate_chips = 10)
```



Object Oriented Programming (OOP) languages

What is object oriented programming?

OOP "is a coding paradigm based on the concept of **objects**, which can contain **data** and **code**."

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OOP "is a coding paradigm based on the concept of **objects**, which can contain **data** and **code**."

Examples of OOP languages:

- Python, C++, R, Java, C#



Object Oriented Programming (OOP) languages

What is object oriented programming?

OOP "is a coding paradigm based on the concept of **objects**, which can contain **data** and **code**."

Examples of OOP languages:

- Python, C++, R, Java, C#



Examples of non-OOP languages:

- Assembler, C, Fortran

Object Oriented Programming (OOP) languages

What is object oriented programming?

OOP "is a coding paradigm based on the concept of **objects**, which can contain **data** and **code**."

Examples of OOP languages:

- Python, C++, R, Java, C#



Examples of non-OOP languages:

- Assembler, C, Fortran

*data fields == attributes
functions == methods*

Outline for Section 4: Python Classes and OOP

1. The anatomy of a class

1. Classes Syntax
2. What is 'self' ?
3. Methods & Attributes

2. Why Classes ?

1. A proof by contradiction
2. Looking at sklearn

3. Dunder methods:

1. `__init__`, `__call__`, `__str__`, `__repr__`
2. Iterable dunder methods

4. Bonus material

Class Basics

- We can define an object instance like so

```
# Create a class template
class Cookie:
    # a class method
    sound = "crunch"
    def make_sound(self):
        print(self.sound, !)
```

```
# Create a class instance
>>> my_cookie = Cookie()
```

Class Basics

- We can access attributes [instance].[attribute]

```
# Create a class template
class Cookie:
    # a class method
    sound = "crunch"
    def make_sound(self):
        print(self.sound, !)
```

```
# Create a class instance
>>> my_cookie = Cookie()

# explore a class
attribute
>>> my_cookie.sound
"crunch"
```

Class Basics

- we can call object **methods** with function notation

```
# Create a class template
class Cookie:
    # a class method
    sound = "crunch"
    def make_sound(self):
        print(self.sound, !)
```

```
# Create a class instance
>>> my_cookie = Cookie()

# explore a class
attribute
>>> my_cookie.sound
"crunch"

# call an instance method
>>> my_cookie.make_sound()
"crunch!"
```

Instance Methods

- Remember methods are functions

```
# Create a class template
class Model_S:
    def make(self):
        [...]
```

Instance Methods

- Remember methods are functions
- We can call them just like functions!

```
# Create a class template
class Model_S:
    def make(self):
        print(f"This car is a {self.year} Tesla Model S")

# Create a class instance
>>> my_tesla = Model_S()
>>> my_tesla.make()
"This car is a 2013 Tesla Model S"
```

Instance Methods

What is `self`?

- The `instance` passes `itself` to its `instance methods` upon being called.

```
# Create a class template
class Model_S:
    year = 2013
    def make(self):
        print(f"This car is a {self.year} Tesla Model S")

# Create a class instance
>>> my_tesla = Model_S()
>>> my_tesla.make(2013)
"This car is a 2013 Tesla Model S"
```

Instance Methods

What is `self`?

- The `instance` passes `itself` to its `instance methods` upon being called.

```
# Create a class template
class Model_S:
    def make(self, year):
        print(f"This car is a {year} Tesla Model S")
```

```
# Create a class instance
>>> my_tesla = Model_S()
>>> my_tesla.make(2013)
"This car is a 2013 Tesla Model S"
```

An F string is a convenient way to format strings with variables,
more on this in section 5

Object self reference

What is *self*?

- The instance passes itself to its instance methods upon being called.
- with **self** we can assign instance attributes

```
# Create a class template
class Model_S:
    def make(self, year):
        self.model = str(year) + "model S with autopilot"
        [...]
```

Object self reference

What is *self*?

- The instance passes itself to its instance methods upon being called.
- with **self** we can assign instance attributes
- or call **instance methods**

```
# Create a class template
class Model_S:
    year = 2013
    def make(self):
        self.make = "model S with autopilot"
        self.vroom()
    def vroom(self):
        print("VROOM")
```

Instance method

```
In [36]: class MyClass:  
....:  
....:     def myfunc(self):  
....:         print('Object method')  
....:
```

```
In [37]: obj = MyClass()
```

```
In [38]: obj.myfunc()  
Object method
```

obj.myfunc()

Instance method

```
In [36]: class MyClass:
```

```
...:
```

```
...:     def myfunc(self):
```

```
...:         print('Object method')
```

```
...:
```

```
In [37]: obj = MyClass()
```

```
In [38]: obj.myfunc()
```

```
Object method
```

```
In [39]: MyClass.myfunc(obj)
```

```
Object method
```

obj.myfunc()

MyClass.myfunc(obj)

Exercise 1

Build your own Tesla Class!

Lessons Slides Prev Next

Session 4: Python Classes

- Storyboard: Python Classes ✓
- Pre-Class Reading Assignment ✓
- Pre-Class Quiz
- Python Classes ✓
- Exercise: Basic Class
- Exercise: Normal Scaler
- Exercise: MyDict
- Post-Class Quiz
- Post-Class Additional Reading - Pause & Think
- Tutorial: Class Basics
- Tutorial 2: Dunder methods and Inheritance

Description

Exercise: Basic Class

The goal of the exercise is to write your own python class.



Instructions:

1. Build a new class `Tesla` based on the instructions given in the docstring (text within the triple quotes).
2. Call an instance of your new class and access some of the instance methods.
3. Answer a couple of questions given in the scaffold.

Hints:

```
class
```

Exercise: Basic Class (Hidden)

Challenge Submissions

Code Text Run All Stop Save Commands

Your first class: Tesla

Read the docstring of the follow toy class and follow the instructions to complete it. Afterwards Remember: attributes and methods can be accessed with dot notation ie `my_tesla.vroom()`

```
[ ] #### edTest(test_class) ####
class Tesla:
    """
    This is the docstring section of a class.
    Adding this to this to the top of your python class or
    function is standard practice.
    See the solutions in one week for the proper form!

    Lets make a simple class!
    Define the following methods:

    methods
    -----
    vroom:
        Assign the string "vroom! vroom!" to the attribute 'sound'.
        Then print self.sound.

    make:
        Assign the car's 'color' to the string "red".
        Also assign the string "2020 model s" to 'model'.
```

/home/basic_class_questions.ipynb

However these analogies are problematic because they suggest that classes are immutable. **They are not. Instances and classes are mutable.**

Modifying Attributes

```
# object attributes are easily overwritten.  
>>> my_cookie.sound  
"crunch"  
  
>>> my_cookie.sound = "pop"  
>>> my_cookie.sound  
"pop"
```

Modifying Attributes

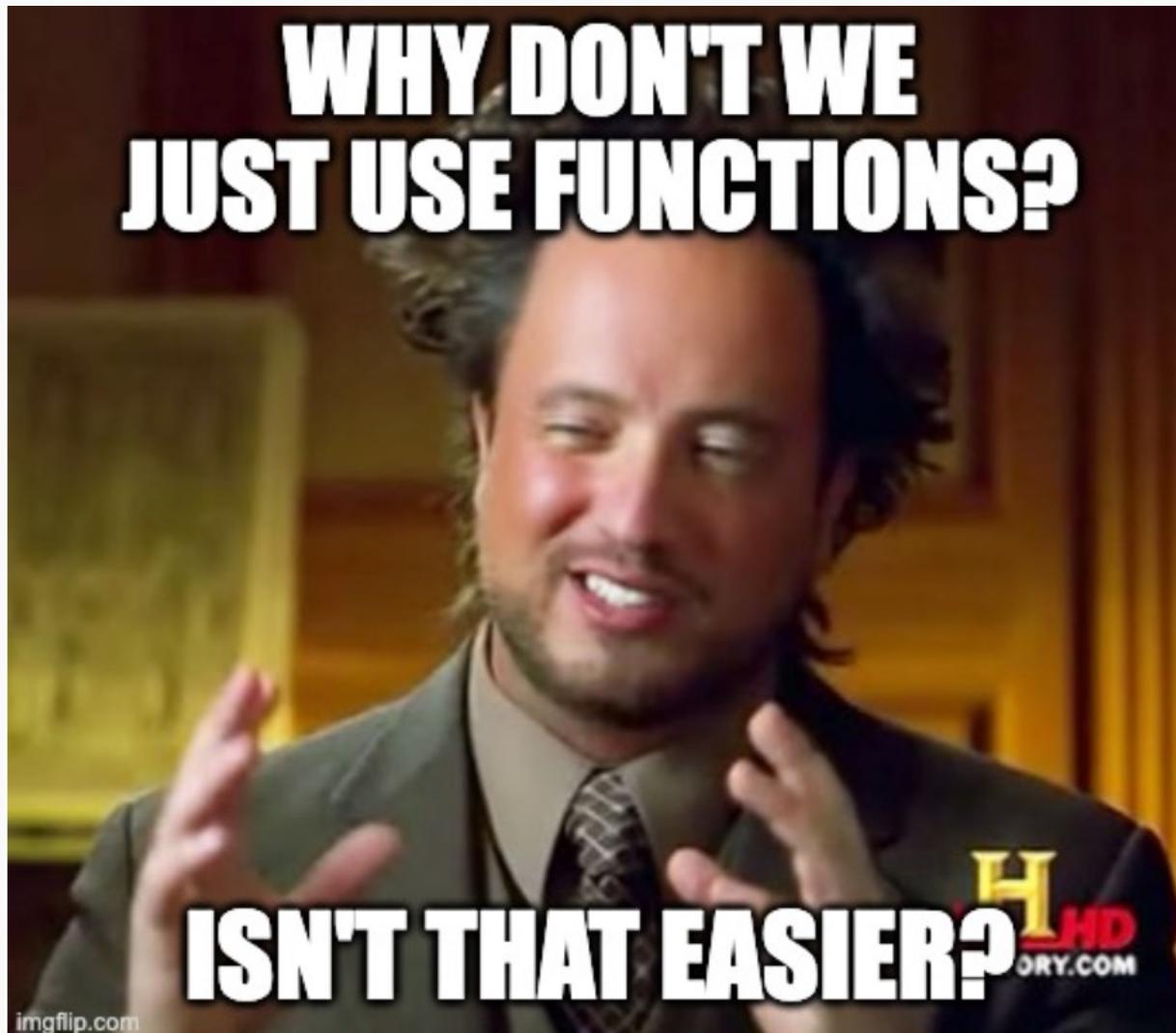
```
# object attributes are easily overwritten.  
>>> my_cookie.sound  
"crunch"  
  
>>> my_cookie.sound = "pop"  
>>> my_cookie.sound  
"pop"  
  
# We can also easily add methods.  
>>> def get_calories(self):  
        print(64)  
>>> my_cookie.get_cal = get_calories  
>>> my_cookie.get_cal()  
"64"
```

Act II Abstraction

Definition 1:

The main purpose of **abstraction** is hiding the unnecessary details from the users. ... It is one of the most important concepts of OOPs.

The problem



Lets do our own *Proof by contradiction*

Let's try a **proof by contradiction**. First assume something is false, then find logical inconsistency.

PROOF
BY
CONTRADICTION

Application: Data Standardization

Standardizing data is important for two reasons:

1. Bringing data to the same scale
2. Machine learning Models like NNs generally work better with small numbers because of **computer precision**

$$Z = \frac{x - \mu}{\sigma}$$

Application: Data Normalization

Standardizing data is important for two reasons:

1. Bringing data to the same scale
2. Machine learning Models like NNs generally work better with small numbers because of **computer precision**

The diagram illustrates the formula for standardizing a data point:

$$Z = \frac{x - \mu}{\sigma}$$

- A purple box labeled "A standardized data point" has a purple arrow pointing to the left side of the equation.
- A blue box labeled "A single data point" has a blue arrow pointing down to the variable x .
- A red box labeled "The mean" has a red arrow pointing left to the variable μ .
- A green box labeled "The standard deviation" has a green arrow pointing left to the variable σ .

Without functions:

```
# calculate standardized dataset 1
>>> data1 = [3.5, 4.2, 1.2, 7.5, 2.2]

# calculate the dataset mean and standard deviation
```

Without functions:

```
# calculate standardized dataset 1
>>> data1 = [3.5, 4.2, 1.2, 7.5, 2.2]

# calculate the dataset mean and standard deviation
>>> n = len(data)
>>> mean1 = sum(data)/n
>>> var1 = sum([(point - mean1)**2/(n - 1)) for point in data1])
>>> std_dev1 = math.sqrt(var1)
```

Without functions:

```
# calculate standardized dataset 1
>>> data1 = [3.5, 4.2, 1.2, 7.5, 2.2]

# calculate the dataset mean and standard deviation
>>> n = len(data)
>>> mean1 = sum(data)/n
>>> var1 = sum([(point - mean1)**2/(n - 1)) for point in data1])
>>> std_dev1 = math.sqrt(var1)

# calculated the standardized data
>>> norm_data = [(dp - mean1)/std_dev1 for dp in data1]
```

Without functions:

```
# calculate standardized dataset 1
>>> data1 = [3.5, 4.2, 1.2, 7.5, 2.2]

# calculate the dataset mean and standard deviation
>>> n = len(data)
>>> mean1 = sum(data)/n
>>> var1 = sum([(point - mean1)**2/(n - 1) for point in data1])
>>> std_dev1 = math.sqrt(var1)

# calculated the standardized data
>>> norm_data = [(dp - mean1)/std_dev1 for dp in data1]

# calculate standardized dataset 2
>>> data = [4.2, 0.1, 8.4, 7.5, 2.9]
>>> mean2 = sum(data)/len(data)
>>> var2 = sum([(point - mean2)**2 for point in data])
>>> std_dev2 = math.sqrt(data_variance)
>>> norm_data2 = [(dp - mean2)/std_dev2 for dp in std_dev2]

#we have to write the same five lines over and over again for every
new dataset, how annoying!
```

Without functions:

```
# calculate standardized dataset 1
>>> data1 = [3.5, 4.2, 1.2, 7.5, 2.2]

# calculate the dataset mean and standard deviation
>>> n = len(data)
>>> mean1 = sum(data)/n
>>> var1 = sum([(point - mean1)**2/(n - 1)]) for point in data1]
>>> std_dev1 = math.sqrt(var1)

# calculated the standardized data
>>> norm_data = [(dp - mean1)/std_dev1 for dp in data1]

# calculate standardized dataset 2
>>> data = [4.2, 0.1, 8.4, 7.5, 2.9]
>>> mean2 = sum(data)/len(data)
>>> var2 = sum([(point - mean2)**2 for point in data])
>>> std_dev2 = math.sqrt(data_variance)
>>> norm_data2 = [(dp - mean2)/std_dev2 for dp in std_dev2]

#we have to write the same five lines over and over again for every
new dataset, how annoying!
```

Patrick says:
“Functions bad!”



I'M AN IDIOT

GET OVER IT

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
    pass
>>> def std(data):
    pass
>>> def standardize(data):
    pass
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        pass
>>> def standardize(data):
        pass
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        pass
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)
```

Application: Data Normalization

Standardizing data is reversible.

Standardize a data point

$$z = \frac{x - \mu}{\sigma}$$

Bring it back to the original scale

$$x = \sigma z + \mu$$

μ : the mean

σ : the standard deviation

Z : a standardized data point

x : a data point

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)

# But what if we want to go back to the original data?!
>>> mean_ = mean(data)
>>> std_ = std(data)
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean_ = mean(data)
>>> std_ = std(data)
>>> modified_data = [data_point*std_ + mean_ for data_point in norm_data_mod ]
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]
>>> def unstd(data):
        pass
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]
>>> def unstd(data, data_mean, data_std):
        return [(dp * data_std) + data_mean for dp in data]
```

Standardization II: With functions

```
# functions (psuedo-code)
>>> def mean(data):
        return sum(data) / len(data)
>>> def std(data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
>>> def standardize(data):
        return [(dp - mean(data))/std(data) for dp in data]
>>> def unstd(data, data_mean, data_std):
        return [(dp * data_std) + data_mean for dp in data]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean_ = mean(data)
>>> std_ = std(data)
>>> modified_data = unstd(data, mean_, std_)
```

Standardization II: With functions

```
>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean_, std_ = mean(data), std(data)
>>> modified_data = unstd(data, mean_, std_)

>>> data2 = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data2 = standardize(data2)
>>> norm_data_mod2 = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean2_ , std2_ = mean(data) , std(data)
>>> modified_data = unstd(data, mean_, std_)
```

The user has to save
assign and keep
track of a lot of
variables for every
dataset!

Standardization II: With functions

Squidward says:
“functions are perfect”

```
>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data = standardize(data)
>>> norm_data_mod = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean_, std_ = mean(data), std(data)
>>> modified_data = unstd(data, mean_, std_)

>>> data2 = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> norm_data2 = standardize(data2)
>>> norm_data_mod2 = f(norm_data)

# But what if we want to go back to the original data!?
>>> mean2_ , std2_ = mean(data) , std(data)
>>> modified_data = unstd(data, mean_, std_)
```



Professor Tentacles and his circle!

~~“functions are perfect”~~



Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
```

Standardization III: With class

```
# mean function  
>>> class Std_scaler:  
    """a standardization class."""
```

Docstrings are a great way to include documentation for your class

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class"""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
```

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class"""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
```

By convention methods internal to the class (not meant for the user to see) have a leading underscore.

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class"""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
    def fit_transform(self, data):
        self.mean_, self.std_ = self._mean(data), self._std(data)
        return [(dp - self.mean_)/self.std_ for dp in data]
```

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
    def fit_transform(self, data):
        self.mean_, self.std_ = self._mean(data), self._std(data)
        return [(dp - self.mean_)/self.std_ for dp in data]
```

By convention in sklearn attributes that have been estimated from the data must always have a name ending with trailing underscore _

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
    def fit_transform(self, data):
        self.mean_, self.std_ = self._mean(data), self._std(data)
        return [(dp - self.mean_)/self.std_ for dp in data]
    def inverse_transform(self, norm_data):
        return [dp * self.std_ + self.mean_ for dp in data]
```

Standardization III: With class

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    def _mean(self, data):
        return sum(data) / len(data)
    def _std(self, data):
        var = sum([(dp - mean(data))**2/(len(data) - 1) for dp in data])
        return math.sqrt(var)
    def fit_transform(self, data):
        self.mean_, self.std_ = self._mean(data), self._std(data)
        return [(dp - self.mean_)/self.std_ for dp in data]
    def inverse_transform(self, norm_data):
        return [dp * self.std_ + self.mean_ for dp in data]
```

Enter the python class, superior to functions!

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    [...]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> scaler = Std_scaler()
```

Enter the python class, superior to functions!

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    [...]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
```

Enter the python class, superior to functions!

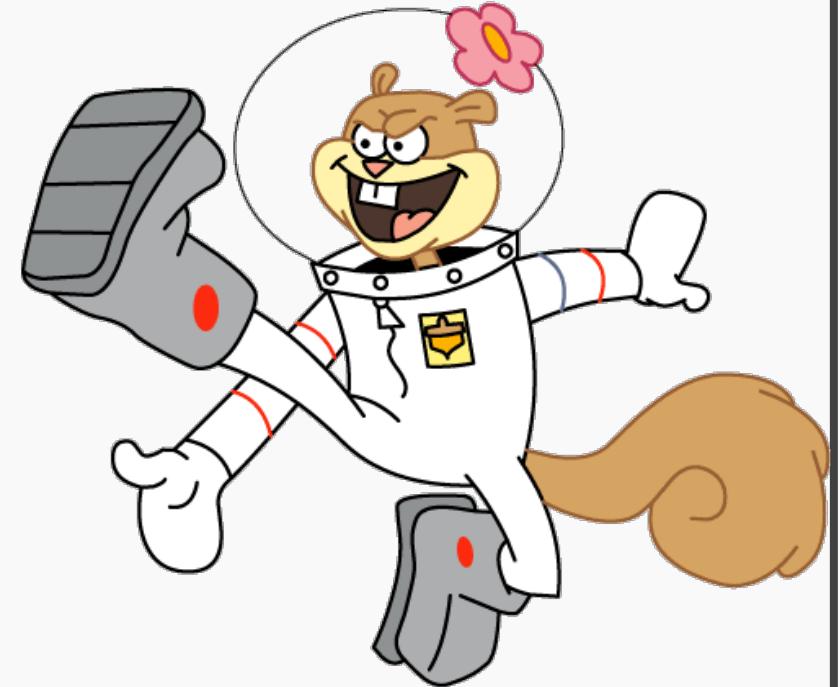
```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    [...]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
>>> altered_data = scaler.inverse_transform(f(norm_data))
```

Enter the python class, superior to functions!

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    [...]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
>>> altered_data = scaler.inverse_transform(f(norm_data))
```

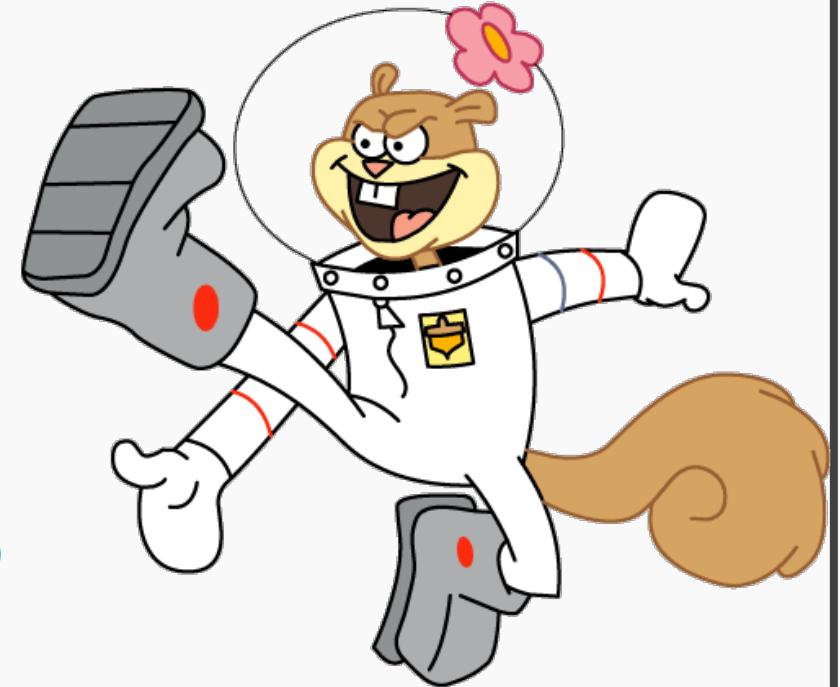


Enter the python class, superior to functions!

```
# mean function
>>> class Std_scaler:
    """a standardization class."""
    [...]

>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
>>> altered_data = scaler.inverse_transform(f(norm_data))

>>> data2 = [5.2, 0.2, 1.5, 6.3, 2.1]
>>> norm_data2 = scaler.fit_transform(data2)
>>> altered_data2 = scaler.inverse_transform(f(norm_data2))
```



Side by side: functions vs classes

```
>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> data2 = [0.3, 3.9, 1.1, 2.5, 6.1]
```

```
>>> mean_ = mean(data)
>>> std_dev_ = std_dev(data)
>>> norm_data = standardize(data)
>>> altered_data = unstd(f(norm_data), mean_, std_dev_)

>>> mean2_ = mean(data2)
>>> std_dev2_ = std_dev(data2)
>>> norm_data2 = standardize(data2)
>>> altered_data2 = unstd(f(norm_data2), mean2_, std_dev2_)
```

```
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
>>> altered_data = scaler.inverse_transform(f(norm_data))

>>> norm_data = scaler.fit_transform(data2)
>>> altered_data2 = scaler.inverse_transform(f(norm_data2))
```

Side by side: functions vs classes

```
>>> data = [3.5, 4.2, 1.2, 7.5, 2.2]
>>> data2 = [0.3, 3.9, 1.1, 2.5, 6.1]
```

```
>>> mean_ = mean(data)
>>> std_dev_ = std_dev(data)
>>> norm_data = standardize(data)
>>> altered_data = unstd(f(norm_data), mean_, std_dev_)
```

```
>>> mean2_ = mean(data2)
>>> std_dev2_ = std_dev(data2)
>>> norm_data2 = standardize(data2)
>>> altered_data2 = unstd(f(norm_data2), mean2_, std_dev2_)
```

```
>>> scaler = Std_scaler()
>>> norm_data = scaler.fit_transform(data)
>>> altered_data = scaler.inverse_transform(f(norm_data))

>>> norm_data2 = scaler.fit_transform(data2)
>>> altered_data2 = scaler.inverse_transform(f(norm_data2))
```

Characters: 288
Lines: 8
User defined objects: 4
User object calls: 8

Characters: 226
Lines: 5
User defined objects: 1
User object calls: 5

Side by side comparison

Characters: 576

Lines: 16

Userdefined objects: 8

User object calls: 16

Characters: 288

Lines: 8

User defined objects: 4

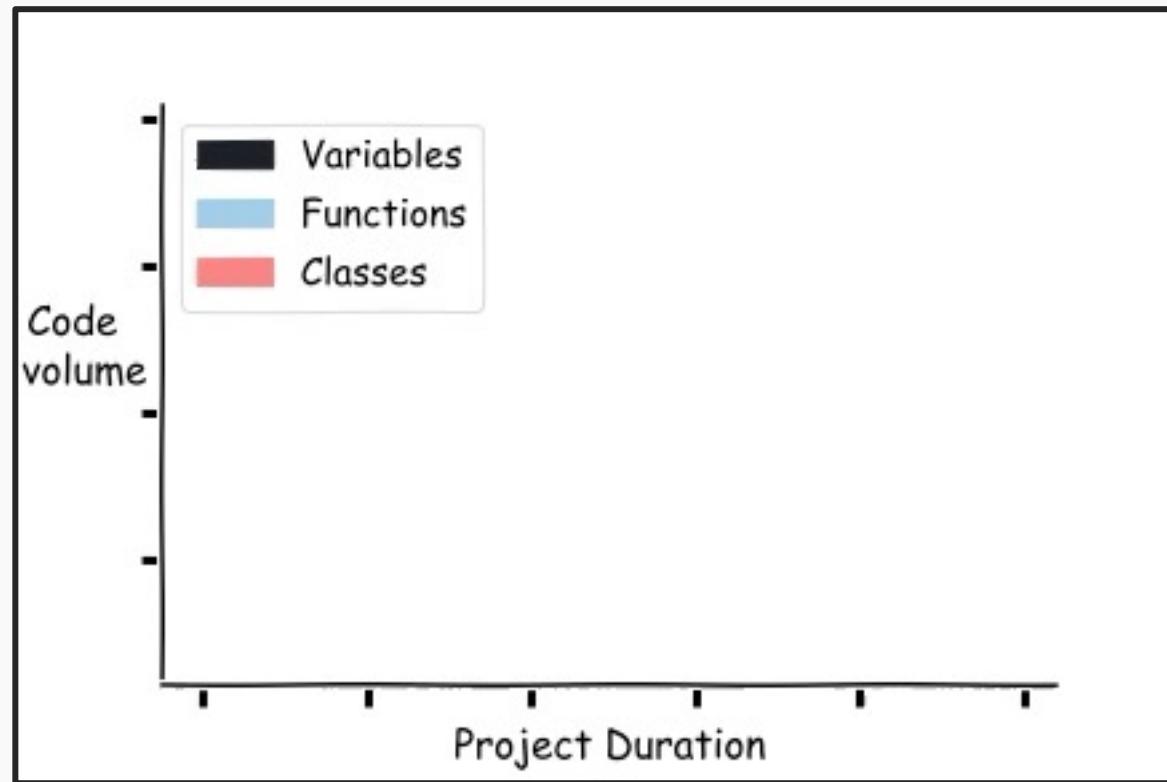
User object calls: 8

Characters: 226

Lines: 5

User defined objects: 1

User object calls: 5



Efficiency of Data Science code:

More abstraction: notebook -> functions -> classes

Notebook only



Co2 per km: ∞

...

Efficiency of Data Science code:

More abstraction: notebook -> functions -> classes

Notebook only



Co2 per km: ∞

...

Efficiency of Data Science code:

More abstraction: notebook -> functions -> classes

Notebook only



Co2 per km: ∞

...

With Functions



Co2 per km: 250 g

Efficiency of Data Science code:

More abstraction: notebook -> functions -> classes

Notebook only



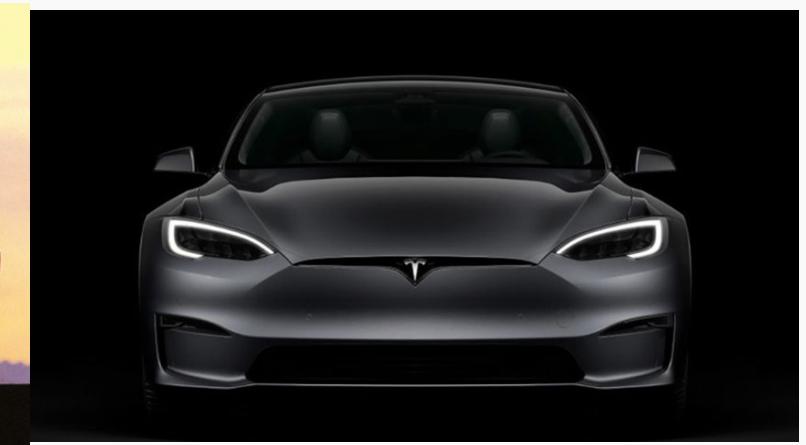
Co2 per km: ∞

With Functions



Co2 per km: 250 g

With Classes:



Co2 per km 0 g

Why should I care?

Data Science Applications

Examples

```
>>> from sklearn.preprocessing import StandardScaler
>>> data = [[0, 0], [0, 0], [1, 1], [1, 1]]
>>> scaler = StandardScaler()
>>> print(scaler.fit(data))
StandardScaler()
>>> print(scaler.mean_)
[0.5 0.5]
>>> print(scaler.transform(data))
[[-1. -1.]
 [-1. -1.]
 [ 1.  1.]
 [ 1.  1.]]
>>> print(scaler.transform([[2, 2]]))
[[3. 3.]]
```

Methods

<code>fit(X[, y, sample_weight])</code>	Compute the mean and std to be used for later scaling.
<code>fit_transform(X[, y])</code>	Fit to data, then transform it.
<code>get_params([deep])</code>	Get parameters for this estimator.
<code>inverse_transform(X[, copy])</code>	Scale back the data to the original representation
<code>partial_fit(X[, y, sample_weight])</code>	Online computation of mean and std on X for later scaling.
<code>set_params(**params)</code>	Set the parameters of this estimator.
<code>transform(X[, copy])</code>	Perform standardization by centering and scaling

- Python data science and Machine Learning (ML) packages are usually implemented as classes

Examples

```
>>> from sklearn.preprocessing import StandardScaler
>>> data = [[0, 0], [0, 0], [1, 1], [1, 1]]
>>> scaler = StandardScaler()
>>> print(scaler.fit(data))
StandardScaler()
>>> print(scaler.mean_)
[0.5 0.5]
>>> print(scaler.transform(data))
[[-1. -1.]
 [-1. -1.]
 [ 1.  1.]
 [ 1.  1.]]
>>> print(scaler.transform([[2, 2]]))
[[3. 3.]]
```

Methods

<code>fit(X[, y, sample_weight])</code>	Compute the mean and std to be used for later scaling.
<code>fit_transform(X[, y])</code>	Fit to data, then transform it.
<code>get_params([deep])</code>	Get parameters for this estimator.
<code>inverse_transform(X[, copy])</code>	Scale back the data to the original representation
<code>partial_fit(X[, y, sample_weight])</code>	Online computation of mean and std on X for later scaling.
<code>set_params(**params)</code>	Set the parameters of this estimator.
<code>transform(X[, copy])</code>	Perform standardization by centering and scaling

✗

~~Functions bad!~~



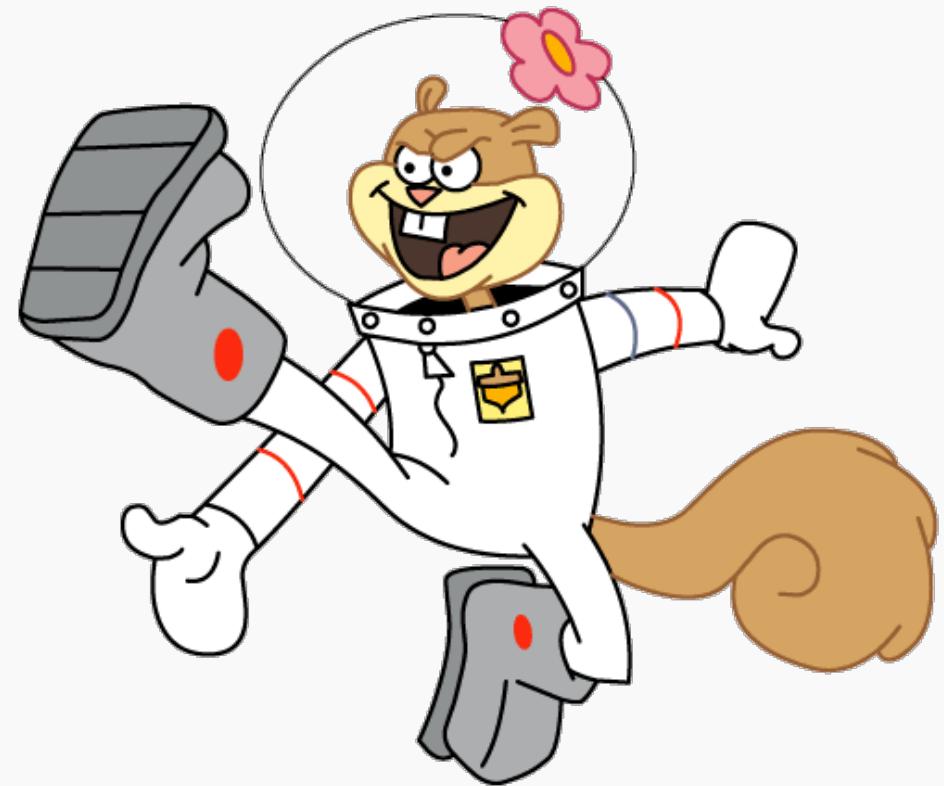
✗

~~I don't need classes~~



✓

Classes are the basis
of professional python
for data science







Dunder Methods

Dunder methods I: initialization and callability

`__init__`, `__call__`

Dunder method

What is a dunder method?

1. Dunder means **double underscore**
2. **Dunder methods ensure that objects follow a certain set of protocalls.**

Object Initialization

What is `__init__`?

When python objects are created

`self` is passed to the `__init__`

method and the object is initialized.

```
# Create a class template
class Model_S:
    def make(self):
        print(f"{self.year} Tesla Model S")
    def __init__(self, model_year):
```



Object Initialization

What is `__init__`?

When python objects are created
self is passed to the `__init__`
method and the object is initialized.

We can then assign
other attributes via other
arguments passed to `__init__`.

```
# Create a class template
class Model_S:
    def make(self):
        print(f"{self.year} Tesla Model S")
    def __init__(self, model_year):
        self.year = model_year
```



Object Initialization

What is `__init__`?

When python objects are created
self is passed to the `__init__`
method and the object is initialized.

We can then assign
other attributes via other
arguments passed to `__init__`. We
can even call other methods
from `__init__`!

```
# Create a class template
class Model_S:
    def make(self):
        print(f"{self.year} Tesla Model S")
    def __init__(self, model_year):
        self.year = model_year
        self.make()
```

```
# Create a class instance
>>> my_tesla = Model_S(model_year = 2017)
"2017 Tesla Model S"
```



Object Initialization

What is `__init__`?

When python objects are created
self is passed to the `__init__`
method and the object is initialized.

We can then assign
other attributes via other
arguments passed to `__init__`. We
can even call other methods
from `__init__`!

```
# Create a class template
class Model_S:
    def make(self):
        print(f"{self.year} Tesla Model S")
    def __init__(self, model_year):
        self.year = model_year
        self.make()
```

```
# Create a class instance
>>> my_tesla = Model_S(model_year = 2017)
"2017 Tesla Model S"

# Instantiate the the class object
>>> your_tesla = Model_S(model_year = 2021)
"2021 Tesla Model S"
```



Object Callability

What is callability?

Callability is the ability of a python object to be called like a function.

```
# Create a class template
class Model_S:
    def __init__(self, loc = "the factory"):
        self.loc = loc
        self.where_am_I()
    def where_am_I(self):
        print(f"My Tesla is in {self.loc}")
```



Object Callability

What is callability?

Callability is the ability of a python object to be called like a function.

What happens if we don't have a call method defined?

```
# Create a class template
class Model_S:
    def __init__(self, loc = "the factory"):
        self.loc = loc
        self.where_am_I()
    def where_am_I():
        print(f"My Tesla is in {self.loc}")

# Create a class instance
>>> my_tesla = Model_S()
My Tesla is in the factory
>>> my_tesla()
TypeError: 'Tesla' object is not callable
```



Object Callability

What is callability?

Callability is the ability of a python object to be called like a function.

What happens if we don't have a call method defined?

You might have seen this error message before.



```
# Create a class template
class Model_S:
    def __init__(self, loc = "the factory"):
        self.loc = loc
        self.where_am_I()
    def where_am_I():
        print(f"My Tesla is in {self.loc}")

# Create a class instance
>>> my_tesla = Model_S()
My Tesla is in the factory
>>> my_tesla()
TypeError: 'Tesla' object is not callable

# make a list of cars
>>> cars = ["volvo", "nissan", "hyundai"]
>>> cars()
TypeError: 'list' object is not callable
```

Object Callability

What is `__call__`?

We can make python instances callable like normal functions via `__call__`.

```
class Tesla:  
    def __init__(self, location = "the factory"):  
        self.loc = location  
        self.where_am_I()  
    def where_am_I(self):  
        print(f"My Tesla is in {self.loc}")
```

This can be very useful when an object needs to change state.



Object Callability

What is `__call__`?

We can make python instances
callable like normal functions via
`__call__`.

```
class Tesla:  
    def __init__(self, location = "the factory"):  
        self.loc = location  
        self.where_am_I()  
    def where_am_I():  
        print(f"My Tesla is in {self.loc}")  
    def __call__(self, location):  
        self.loc = location  
        self.where_am_I()
```



Object Callability

What is `__call__`?

We can make python instances
callable like normal functions via
`__call__`.

```
class Tesla:  
    def __init__(self, location = "the factory"):  
        self.loc = location  
        self.where_am_I()  
    def where_am_I():  
        print(f"My Tesla is in {self.loc}")  
    def __call__(self, location):  
        self.loc = location  
        self.where_am_I()  
  
# Create a class instance  
>>> my_tesla = Tesla()  
"My Tesla is in the factory"
```



Object Callability

What is `__call__`?

We can make python instances
callable like normal functions via
`__call__`.

```
class Tesla:  
    def __init__(self, location = "the factory"):  
        self.loc = location  
        self.where_am_I()  
    def where_am_I():  
        print(f"My Tesla is in {self.loc}")  
    def __call__(self, location):  
        self.loc = location  
        self.where_am_I()
```

```
# Create a class instance  
>>> my_tesla = Tesla()  
"My Tesla is in the factory"
```

```
# treat our class instance like a function  
>>> my_tesla("California")  
"My Tesla is in California"
```



Object Callability

What is `__call__`?

We can make python instances callable like normal functions via `__call__`.

```
class Tesla:  
    def __init__(self, location = "the factory"):  
        self.loc = location  
        self.where_am_I()  
    def where_am_I():  
        print(f"My Tesla is in {self.loc}")  
    def __call__(self, location):  
        self.loc = location  
        self.where_am_I()
```

We can even return a value using the `__call__()` magic method

```
# Create a class instance  
>>> my_tesla = Tesla()  
"My Tesla is in the factory"
```

```
# treat our class instance like a function  
>>> my_tesla("California")  
"My Tesla is in California"
```

```
>>> my_tesla.loc  
"California"
```



Exercise II, Build your own scaler!

`__init__`, `__call__`

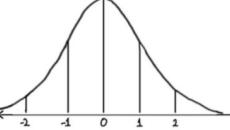
Lessons Slides Prev Next

Description

Exercise: Normal Scaler

The goal of the exercise is to write your own python scaler class.

In lecture we saw data standardization.



Normalization is different: It sets the minimum of the data to 0 and the maximum to 1.

Instructions:

1. Build a new class `Normal_scaler` based on the instructions given in the docstring (text within the triple quotes).
2. Build an instance of your class
3. Answer a multiple choice question
4. Compare two different scalers with the `'vars'` function

Lets Build our own scaler class!

In lecture we saw data standardization. Normalization is different: It sets the minimum of the data to 0 and the maximum to 1. This is especially useful in the context of data that is bounded (For example : Image pixel strength 255).

Specifically our task will be to normalize the following:

```
data = [11, 6, 3, 4.5, 18]
```

The instructions are given in the docstring.

```
### edTest(test_class) ###
class Normal_scaler:
    """
    Follow the instructions of this docstring to finish this class.
    Normal_scaler will normalize data and inverse transform it, keeping
    track of pesky details in the process!
    """

    The formula for normalization is  $(x - x_{\min}) / (x_{\max} - x_{\min})$ .
    Normalization ensures that all values will be between zero and one for this class.

    Make sure to save the following instance attributes:
    -----
    min_ : the dataset minimum
    max_ : the dataset maximum
```

/home/normalization.ipynb

Dunder methods II: (iterables)

`__getitem__`, `__setitem__`, `__len__`

__getitem__ and __setitem__

```
from random import randrange as rand

class Harlist:
    def __init__(self, num = 5):
        self.list = [rand(1, num, 1) for _ in
range(num)]
```

__getitem__ and __setitem__

```
from random import randrange as rand

class Harlist:
    def __init__(self, num = 5):
        self.list = [rand(1, num, 1) for _ in range(num)]

# Create a class instance
>>> random_ints = Harlist()
>>> random_ints.list
[5, 3, 4, 1, 4]
```

__getitem__ and __setitem__

```
from random import randrange as rand

class Harlist:
    def __init__(self, num = 5):
        self.list = [rand(1, num, 1) for _ in range(num)]

# Create a class instance
>>> random_ints = Harlist()
>>> random_ints.list
[5, 3, 4, 1, 4]
>>> random_ints[0]
TypeError: 'Rand_Int_List' object is not
subscriptable
```

__getitem__ and __setitem__

Unless we implement
__getitem__
and __setitem__ we
can't use the index
operator []

```
from random import randrange as rand

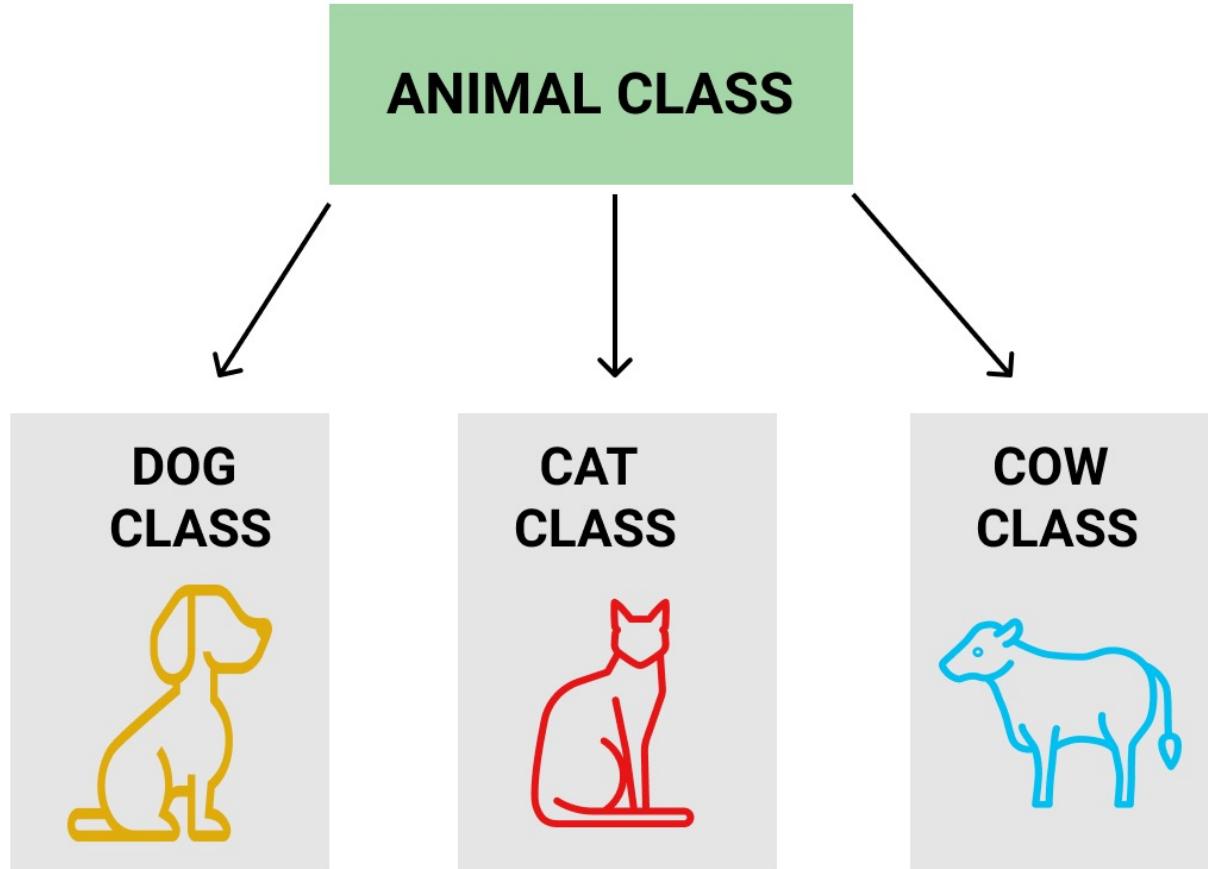
class Harlist:
    def __init__(self, num = 5):
        self.list = [rand(1, num, 1) for _ in range(num)]

# Create a class instance
>>> random_ints = Harlist()
>>> random_ints.list
[5, 3, 4, 1, 4]
>>> random_ints[0]
TypeError: 'Rand_Int_List' object is not
subscriptable
```

Act III Inheritance

Act III Inheritance

In **object-oriented** programming, **inheritance** is the mechanism of basing an object or class upon another object (prototype-based **inheritance**) or class (**class-based inheritance**), retaining similar implementation. ... An **inherited** class is called a subclass of its parent class or super class.



__getitem__, __setitem__ and Inheritance

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
```

__getitem__, __setitem__ and Inheritance

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
```

__getitem__, __setitem__ and Inheritance

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
```

Here we have snuck in the first example of inheritance. We don't need to re-write the `__init__` function from the previous class, we can inherit it!

```
class Harlist:
    def __init__(self, num):
        self.list = [rand(1, 6, 1) for _ in range(num)]
```

__getitem__, __setitem__ and Inheritance

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
    def __setitem__(self, index, value):
        self.list[index] = value
```

__getitem__, __setitem__

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
    def __setitem__(self, index, value):
        self.list[index] = value

# Create a class instance
>>> random_ints = Harlist_child()
>>> random_ints.list
[4, 3, 4, 2, 4]
>>> random_ints[0]
4
```

__getitem__, __setitem__

Nice! Now it works as expected.

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
    def __setitem__(self, index, value):
        self.list[index] = value

# Create a class instance
>>> random_ints = Harlist_child()
>>> random_ints.list
[4, 3, 4, 2, 4]
>>> random_ints[0]
4
>>> random_ints[0] = 100
>>> random_ints[0]
100
```

__getitem__, __setitem__

But we can't print
the list!

```
from random import randrange as rand

class Harlist_child(Harlist):
    def __getitem__(self, key):
        return self.list[key]
    def __setitem__(self, index, value):
        self.list[index] = value

# Create a class instance
>>> random_ints = Harlist_child()
>>> random_ints.list
[4, 3, 4, 2, 4]
>>> random_ints[0]
4
>>> random_ints[0] = 100
>>> random_ints[0]
100
>>> print(random_int)
<__main__.Rand_Int_List object at 0x7fb8d029b700>
```

Dunder methods III:
`__str__`, `__repr__`

Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str
```



Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

`__str__` is informal string representation.

```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
    def __str__(self):  
        print("str scope")  
        str = f"{self.year} Model {self.model}"  
        return str
```



Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

`__str__` is informal string representation.

```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
    def __str__(self):  
        print("str scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
  
# Create a class instance  
>>> my_tesla = Tesla("X")  
>>> my_tesla  
repr scope  
2021 Model X
```



Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

`__str__` is informal string representation.



```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
    def __str__(self):  
        print("str scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
  
# Create a class instance  
>>> my_tesla = Tesla("X")  
>>> my_tesla  
repr scope  
2021 Model X  
  
>>> print(my_tesla)  
str scope  
2021 Model X
```

Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

`__str__` is informal string representation.

When we lack `__str__` python falls back to `__repr__` (and not the other way around), so `__repr__` is more important.

```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
  
# Create a class instance  
>>> my_tesla = Tesla("X")  
>>> my_tesla  
repr scope  
2021 Model X  
  
>>> print(my_tesla)  
repr scope  
2021 Model X
```



Object Representation

What are `__str__` and `__repr__`?

`__repr__` is formal string representation.

`__str__` is informal string representation.

When we lack `__str__` python falls back to `__repr__` (and not the other way around), so `__repr__` is more important.

```
class Tesla:  
    def __init__(self, model, year = 2021):  
        self.year, self.model = year, model  
    def __repr__(self):  
        print("repr scope")  
        str = f"{self.year} Model {self.model}"  
        return str  
  
# Create a class instance  
>>> my_tesla = Tesla("X")  
>>> my_tesla  
repr scope  
2021 Model X  
  
>>> print(my_tesla)  
repr scope  
2021 Model X
```

We may want to customize the way we can interact with object representation. For example we may want more information from print and a simpler version from repr, allowing us to more easily interact with our machine learning models



Capping off Dunder Functions:

```
from random import randrange as rand

class Harlist_subsubclass(Harlist_subclass):
    [...]
```

Capping off Dunder Functions:

```
from random import randrange as rand

class Harlist_subsubclass(Harlist_subclass):
    def __len__(self):
        return len(self.list)
```

Capping off Dunder Functions:

```
from random import randrange as rand

class Harlist_subsubclass(Harlist_subclass):
    def __len__(self):
        return len(self.list)
    def __repr__(self):
        return str(self.list)
```

Capping off Dunder Functions:

```
from random import randrange as rand

class Harlist_subsubclass(Harlist_subclass):
    def __len__(self):
        return len(self.list)
    def __repr__(self):
        return str(self.list)

# Create a class instance
>>> random_ints = Harlist_subsubclass(3)
>>> print(random_ints)
[4, 0 ,5]
>>> random_ints, len(random_ints)
[4, 0 ,5], 3
```

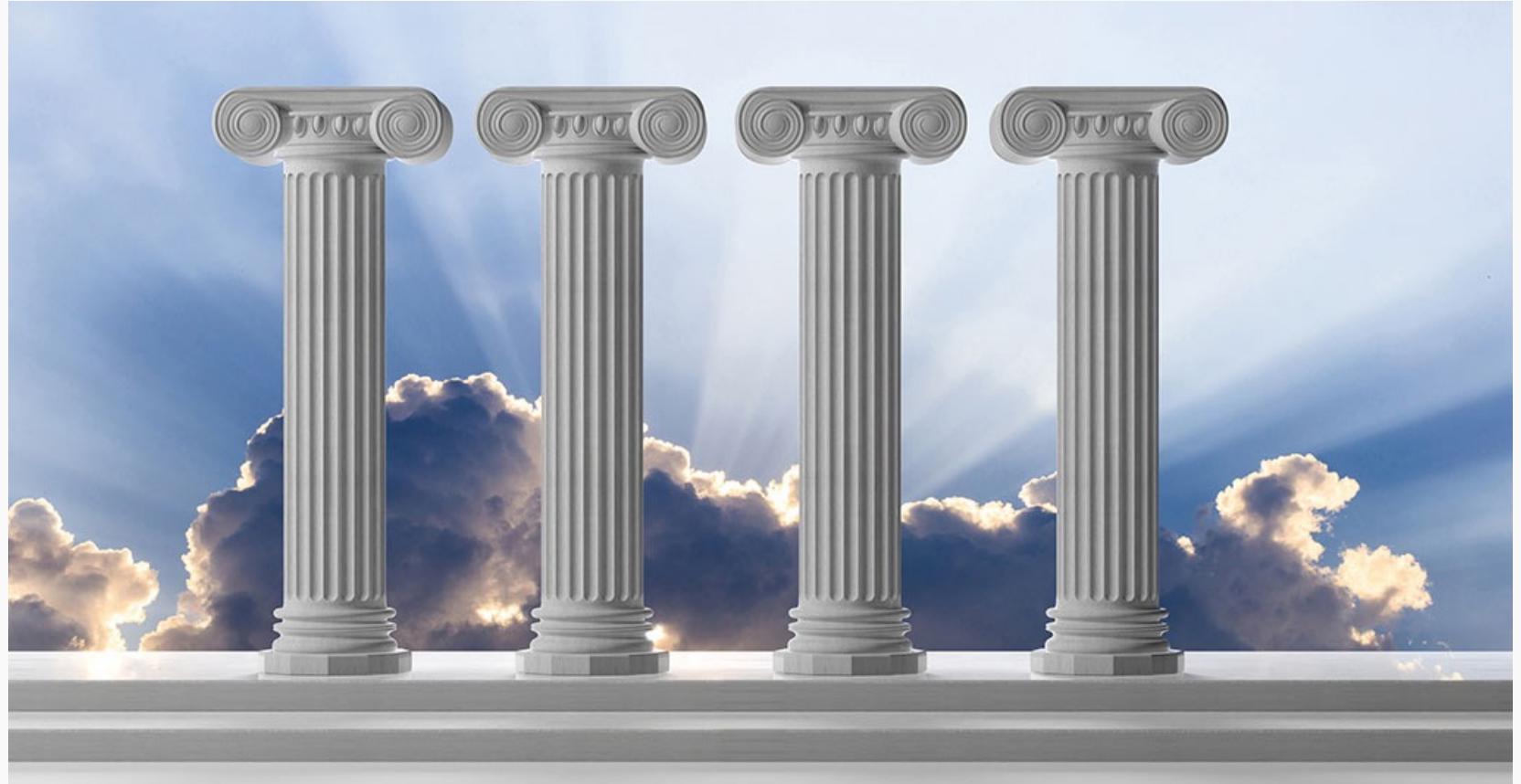
Object Oriented Programming (OOP): the four pillars

I. Encapsulation

II. Abstraction

III. Inheritance

IV. Polymorphism



Exercise to implement your own dictionary

Lets say we are trying to make the following assignment:

```
my_dict["bye"] = "hasta luego"
```

Exercise to implement your own dictionary

You might be surprised but a dictionary is just a list of lists!

We initialize the list of lists like so, up to n long.

```
storage = [[], [], [], [], ...]
```

Exercise to implement your own dictionary

You might be surprised but a dictionary is just a list of lists!

We initialize the list of lists like so, up to n long.

```
storage = [[], [], [], [], ...]
```

Then we use a **hash** function to convert the key to an integer.

For example:

hash(123) = 123

hash("bye") = -5844541644705434020

Exercise to implement your own dictionary

Next we take the modulo of these hashes by the length of the storage. `hash("bye") % 1000 = -5844541644705434020 = 980`

Exercise to implement your own dictionary

Next we take the modulo of these hashes by the length of the storage. For example my_dict[“bye”] = ”hasta luego”

`hash("bye") % 1000 = -5844541644705434020 = 980`

So now index 998 of the list will look like this:

`[..., [[“bye”, “hasta luego”]], ...]`

The final pillar: Polymorphism!

Polymorphism is **the ability of an object to take on many forms.**

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self, num):
        self.list = [rand(1, num, 1)**2 for _ in range(num)]
```

Polymorphism: overwriting parent methods

Here we
overwrote the
upstream classes
method.



```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self, num):
        self.list = [rand(1, num, 1)**2 for _ in range(num)]
```

Polymorphism: overwriting parent methods

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self):
        self.list = [rand(1, 6, 1)**2 for _ in range(num)]

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5
```

Polymorphism: overwriting parent methods

We can overwrite methods from the parent class **as shown**

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self):
        self.list = [rand(1, 6, 1)**2 for _ in range(num)]

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5
```

Polymorphism: `super()`

Alternatively we can use the `super()` function to call methods from the parent class.

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self, num):
        self.list = [rand(1, num, 1)**2 for _ in range(num)]
    def init_parent(self, num):
        super().__init__(num)
```

Polymorphism: `super()`

Alternatively we can use the `super()` function to call methods from the parent class.

Think of `super` as a place holder for the parent class.

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self, num):
        self.list = [rand(1, num, 1)**2 for _ in range(num)]
    def init_parent(self, num):
        super().__init__(num)

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5
```

Polymorphism: `super()`

Alternatively we can use the `super()` function to call methods from the parent class.

Think of `super` as a place holder for the parent class.

Here we called the parent classes `__init__` method!

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self):
        self.list = [rand(1, 6, 1)**2 for _ in range(num)]
    def init_parent(self, num):
        super().__init__(num)

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5

>>> square_ints.init_parent(5)
>>> square_ints
[3, 5, 2, 2, 4]
```

Polymorphism: `super()`

Think of it this way:

`super() ==`
`[Parent_class]`

```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self):
        self.list = [rand(1, 6, 1)**2 for _ in range(num)]
    def init_parent(self, num):
        Parent = super()
        Parent.__init__(num)

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5

>>> square_ints.init_parent(5)
>>> square_ints
[3, 5, 2, 2, 4]
```

Polymorphism: `super()`

Think of it this way:

`Super() ==`
[Parent_class]

Dunder methods
ensure that objects
follow a certain set of
protocalls. This is a
form of polymorphism.

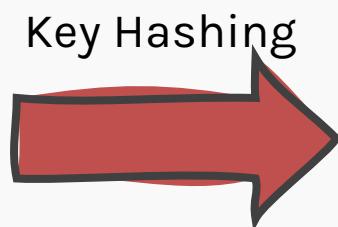
```
from random import randrange as rand

class Harlist_squared(Harlist_subclass):
    def __init__(self):
        self.list = [rand(1, 6, 1)**2 for _ in range(num)]
    def init_parent(self, num):
        Parent = super()
        Parent.__init__(num)

# Create a class instance
>>> square_ints = Harlist_squared(5)
>>> square_ints, len(square_ints)
[1, 16, 9, 4, 9], 5

>>> square_ints.init_parent(5)
>>> square_ints
[3, 5, 2, 2, 4]
```

Key	Value
	Brownie
	Simba
	Gillu
	Sandy



```
[  
  [['Dog','Brownie'],  
   [ ],  
   [ ],  
   [ ],  
   [ ],  
   [ ],  
   [['Cat','Simba'],['Camel','Sandy']],  
   [ ],  
   [ ],  
   [ ],  
   [ ],  
   ['Hamster','Gillu'],  
   [ ],  
   [ ],  
   [ ],  
   [ ],...  
 ]
```

MyDict()

What is a python **class**?

What is a docstring?

```
1 class Cookie:
2     """A generic cookie class"""
3     calories_per_gram = {"sugar": 4, "butter": 7,
4                           "cornstarch": 4, "egg": 2,
5                           "chocolate_chips": 5}
6
7     def __init__(self, cookie_type, sugar_grams):
8         self.cookie_type, self.sugar_grams = cookie_type, sugar_grams
9
10    @staticmethod
11    def calc_cal(ingredient_dict, calories_per_gram):
12        total_calories = 0
13        for ingredient, amount in ingredient_dict.items():
14            total_calories += amount * calories_per_gram[ingredient]
15        return total_calories
16
17    class Chocolate_chip(Cookie):
18        """A chocolate chip cookie recipe!"""
19        def __init__(self, num_chocolate_chips):
20            self.cookie_type = "chocolate_chip"
21            self.num_choc_chips = num_chocolate_chips
22            #grams
23            self.ingredient_dict = { "sugar": 32, "butter": 170,
24                                    "chocolate_chips": num_chocolate_chips,
25                                    "cornstarch": 12, "egg": 50
26                                    }
27            self.num_cookies = 16
28
29        def __str__(self):
30            total_calories = self.calc_cal(self.ingredient_dict, self.calories_per_gram)
31            cal_per_cookie = round(total_calories / self.num_cookies)
32            cchips = self.ingredient_dict
33            return 'I am a {} with {} calories and {} chocolate chips'.format(
34                self.cookie_type, cal_per_cookie, self.num_choc_chips)
```

What is **staticmethod**?

What is
__init__?

What is
Inheritance?

What is **self**?

What is
__str__?

What is **__call__**?

What is a dunder Method?

What is the difference
between classes and
methods?