Metadata * Title: **Final Project Report** * Class: DS 5100 * Date: * Student Name: Sivaranjani Kandasami * Student Net ID: nyc2xu * This URL: https://github.com/ranjani-joseph/final_project.git * GitHub Repo URL:

The Monte Carlo Module

```
In [ ]: # A code block with your classes.
               import pandas as pd
               import numpy as np
              class Die:
                      \ensuremath{^{'\,'}}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{\,'}\xspace^{
                          W defaults to 1.0 for each face but can be changed after the object is created.
                          Note that the weights are just numbers, not a normalized probability distribution.
                          The die has one behavior, which is to be rolled one or more times.
                          Note that what we are calling a "die" here can represent a variety of random variables associated with stochastic
                           We can create these models by increasing the number of sides and defining the values of their faces.
                          Our probability models for such variables are, however, very simple - since our weights apply to only to single ev
                     def __init__(self,faces):
    '''Initializer takes an array of faces as an argument.
                            The array's data type (dtype) may be strings or numbers.
                            The faces must be unique; no duplicates.
                            Internally iInitializes the weights to 1.0 for each face.
                            Saves faces and weights in a private dataframe that is to be shared by the other methods.^{\circ\circ}
                            self.faces = list(set(faces)) # array of strings or numbers
                            self.weights = [1.0 for self.face in range(len(self.faces))] # array of integers
                            self.die = pd.DataFrame({'faces':self.faces, 'weights':self.weights}) #dataframe with faces and weights
                     def change weight(self, face value, new weight):
                                'This method is to change the weight of a single side.
                            Takes two arguments: the face value to be changed and the new weight.
                            Checks to see if the face passed is valid; is it in the array of weights?
                            Checks to see if the weight is valid; is it a float? Can it be converted to one?'''
                            self.checks = []
                            if (type(face_value)==str or type(face_value)==int) and face_value in self.faces:
                                   self.checks.append(True)
                                     print("Checks passed, valid face value and weight")
                            else:
                                  self.checks.append(False)
                                  print("Face value should be array of strings or integer. ")
                                  self.weight=float(new weight)
                                  self.checks.append(True)
                            except:
                                  print("Weight cannot be converted to float")
                                  self.checks.append(False)
                            if self.checks:
                                  self.i = self.faces.index(face value)
                                  self.weights[self.i] = new_weight
                            else:
                                  print("Checks failed. Pass valid face value and weight")
                            self.die = pd.DataFrame({'faces':self.faces, 'weights':self.weights})
                            return self.die
                     def roll(self,num_roll=1):
                              ''This method is to roll the die one or more times.
                            Takes a parameter of how many times the die is to be rolled; defaults to 1.
                            This is essentially a random sample from the vector of faces according to the weights.
                            Returns a list of outcomes.
                            Does not store internally these results.'''
                            return list(self.die.sample(n=num roll,weights='weights',replace=True)['faces'])
                     def show(self):
                             '''This method is to show the user the die's current set of faces and weights (since the latter can be changed).
                            Returns the dataframe created in the initializer but possibly updated by the weight changing method.'''
                            return self.die
               class Game:
                         'Game class consists of rolling of one or more dice of the same kind one or more times.
                          Each game is initialized with a list of one or more of similarly defined dice (Die objects).
                          By "same kind" and "similarly defined" we mean that each die in a given game has the same number of sides and set
                          The class has a behavior to play a game, i.e. to roll all of the dice a given number of times.
                          The class keeps the results of its most recent play.
                     def init (self,dice):
```

```
'''Takes a single parameter, a list of already instantiated similar Die objects.'''
       self.dice = dice
   def play(self,num_roll):
         'Takes a parameter to specify how many times the dice should be rolled. Saves the result of the play to a priva
       That is, each role is an observation and each column is a feature. Each cell should show the resulting face for t
       The private dataframe should have the roll number is a named index.''
       self.num roll = num roll
       self.play_result = pd.DataFrame()
       n=1
       for die in self.dice:
           self.result = die.roll(num_roll)
            self.play_result['Die'+str(n)] = self.result
       self.play_result['Roll']=list(np.arange(1,num_roll+1))
       self.result_df = self.play_result.set_index('Roll')
       return self.result_df
   def show(self,df form='wide'):
        '''This method just passes the private dataframe to the user.
       Takes a parameter to return the dataframe in narrow or wide form.
       This parameter defaults to wide form, which is what the previously described method produces.
       This parameter should raise an exception if the user passes an invalid option.
       The narrow form of the dataframe will have a two-column index with the roll number and the die number, and a sing
       self.df form = df form
       if self.df_form == 'wide':
             print("wide")
            self.df_to_return = self.result_df
       elif df form == 'narrow':
           self.df_to_return = self.result_df.stack().to_frame('Face')
       else:
           raise Exception("Dataframe form should be wide or narrow")
       return self.df to return
class Analyzer:
   def __init__(self,game):
          'An analyzer takes the results of a single game and computes various descriptive statistical properties about i
       A face counts per roll, i.e. the number of times a given face appeared in each roll.
       For example, if a roll of five dice has all sixes, then the counts for this roll would be 6 for the face value '6
       A jackpot count, i.e. how many times a roll resulted in all faces being the same, e.g. six ones for a six-sided d
       A combo count, i.e. how many combination types of faces were rolled and their counts.
       A permutation count, i.e. how may sequence types were rolled and their counts.'
       self.game = game
   def face counts per roll(self):
          'A face counts per roll method to compute how many times a given face is rolled in each event.
       Stores the results as a dataframe in a public attribute.
       The dataframe has an index of the roll number and face values as columns (i.e. it is in wide format).'''
       face_counts_per_roll = pd.DataFrame()
       for i in range(len(self.game)):
           each_roll = pd.DataFrame(self.game.iloc[i].value_counts()).transpose()
           face_counts_per_roll = face_counts_per_roll.append(each_roll)
       face_counts_per_roll.index.name = 'Game
       return face counts per roll
        '''A combo method to compute the distinct combinations of faces rolled, along with their counts.
       Combinations should be sorted and saved as a multi-columned index.
       Stores the results as a dataframe in a public attribute.'''
       self.combo_results = pd.DataFrame()
       for i in range(len(self.game)):
           \verb|self.combo_results = \verb|self.combo_results.append(pd.DataFrame(self.game.iloc[i].value\_counts()).transpose())|
           self.combo results.index.name = "Roll"
       return self.combo_results
   def jackpot(self):
         ''A jackpot method to compute how many times the game resulted in all faces being identical.
       Returns an integer for the number times to the user.
       Stores the results as a dataframe of jackpot results in a public attribute.
       The dataframe should have the roll number as a named index.'
       jackpot_results = pd.DataFrame()
       analyzer = Analyzer(self.game)
       self.combo_results = analyzer.combo()
       for i in range(len(self.combo_results)):
           if self.combo_results.iloc[i].count()==1:
               jackpot_results = jackpot_results.append(self.combo_results.iloc[i])
       all_faces_identical = len(jackpot_results)
       return all_faces_identical
```

Test Module

```
In [2]: # A code block with your test code.
         import unittest
         import numpy as np
         import pkg_mc
         from imp import reload
         reload(pkg_mc)
         from pkg mc.montecarlo import *
         class MonteCarloTestSuite(unittest.TestCase):
             def test_test(self):
                self.assertTrue(True)
             # 'test_1_change_weight()': Change weight of a face in list of faces.
             # 'test_2_roll_die()': Roll die 2 times and store results in dataframe
             # 'test_3_die_show()': Show faces of die
             # 'test 4 game play()': Play game twice and see results
             # 'test_5_face_counts()': Count faces of game
             # 'test_6_combo()': Verify if the combos are summed
             # 'test 7 jackpot()': Verify jackpot results are correct
             def test_1_die_change_weight(self):
                 # Change weight of a face
                 faces_list = [2,3,4,5]
                 mc change weight = Die(faces list)
                 df = mc_change_weight.change_weight(2,6)
                 assert (df['weights'][0] == 6)
             def test_2_die_roll(self):
                 # Roll die 2 times and store results in dataframe
                 faces_list = [2,3,4,5]
                 die = Die(faces_list)
                 roll list = die.roll(2)
                 assert (len(roll_list) == 2)
             def test_3_die_show(self):
                 # Roll die 2 times and store results in dataframe
                 faces list = [2,3,4,5]
                 die = Die(faces_list)
                 df = die.show()
                 assert (len(df) == 4)
             def test_4_game_play(self):
                 # Play game and store the results in a wide dataframe
                 faces_list = [1,2,3,4,5]
                 die1 = Die(faces_list)
                 die2 = Die(faces_list)
                 dice list = [die1,die2]
                 game = Game(dice_list)
                 game play = game.play(4)
                 game.show("wide")
                 assert len(game_play)==4
             def test_5_face_counts(self):
                 # Count faces of game
                 df = pd.DataFrame({'Game':[1,2,3,4], 'die1':[2,6,4,7], 'die2':[4,8,4,2], 'die3':[1,5,4,6]})
                 df = df.set_index('Game')
                 analyze = Analyzer(df)
                 face_counts = analyze.face_counts_per_roll()
                 assert (face_counts.iloc[3][6] ==1)
             def test_6_combo(self):
                 # Verify if the combos are summed
                 df = pd.DataFrame({'Game':[1,2,3,4], 'die1':[2,6,4,7], 'die2':[4,8,4,2], 'die3':[1,5,4,6]})
                 df = df.set_index('Game')
                 analyze = Analyzer(df)
                 game combo = analyze.combo()
                 assert (game_combo.iloc[2][4] == 3)
             def test_7_jackpot(self):
                 #Verify jackpot results are correct
                 df = pd.DataFrame({'Game':[1,2,3,4], 'diel':[2,6,4,7], 'die2':[4,8,4,2], 'die3':[1,5,4,6]})
                 df = df.set_index('Game')
                 analyze = Analyzer(df)
                 game_combo = analyze.jackpot()
                 assert (game combo == 1)
```

Test Results

bash-4.2\$python montecarlo_tests.py test_1_die_change_weight (__main__.MonteCarloTestSuite) ... ok test_2_die_roll (__main__.MonteCarloTestSuite) ... ok test_3_die_show (__main__.MonteCarloTestSuite) ... ok test_4_game_play (__main__.MonteCarloTestSuite) ... ok test_5_face_counts (__main__.MonteCarloTestSuite) ... ok test_6_combo (__main__.MonteCarloTestSuite) ... ok test_7_jackpot (__main__.MonteCarloTestSuite) ... ok test_test (__main__.MonteCarloTestSuite) ... ok test_0.080s OK bash-4.2\$

Scenarios

Code blocks with your scenarios and their outputs.

These should have appropriate import statements even though the code is now in the same notebook as the classes it calls.

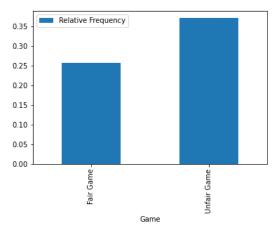
Scenario 1

Scenario 1: 2-headed coin

Create one fair coin (with faces H and T) and one unfair coin. For the unfair coin, give one of the faces a weight of 5 and the others 1. Play a game of 1000 flips of three coins with all fair dice. Play a game of 1000 flips with two unfair dice and one fair die. For each game, use an Analyzer object to determine the relative frequency of jackpots – getting either all Hs or all Ts. Compute relative frequency as the number of jackpots over the total number of rolls. Show your results, comparing the two relative frequencies, in a simple bar chart.

```
In [8]: # Code blocks with output
         # Create one fair coin (with faces H and T) and one unfair coin. For the unfair coin, give one of the faces a weight of 5
         faces = ['H', 'T']
         num_roll = 1000
         #Fair Coin
         fair_coin = Die(faces)
         fair_coin_face_list = fair_coin.faces
         fair_coin.show()
         #Unfair Coin
         unfair_coin = Die(faces)
         unfair_coin_face_list = unfair_coin.faces
         unfair_coin.change_weight('T',5)
         # Play a game of 1000 flips of three coins with all fair dice
         fair_coin_list = []
         for i in range(3):
             fair coin list.append(fair coin)
         fair_game = Game(fair_coin_list)
         fair_game.play(num_roll)
         fair game.show()
         analyze_fair_game = Analyzer(fair_game.show())
         fair_game_jackpot = analyze_fair_game.jackpot()
         print("Number of jackpots for 3 fair dice : "+str(fair_game_jackpot))
         #Relative frequency
         fair_game_rf = fair_game_jackpot/num_roll
         # Play a game of 1000 flips with two unfair dice and one fair die - referred as unfair_game
         unfair_coin_list = [fair_coin]
         for i in range(2):
            unfair_coin_list.append(unfair_coin)
         unfair game = Game(unfair coin list)
         unfair_game.play(num_roll)
         unfair_game.show()
         analyze_unfair_game = Analyzer(unfair_game.show())
         unfair_game_jackpot = analyze_unfair_game.jackpot()
         print("Number of jackpots for 2 unfair and 1 fair dice: "+str(unfair game jackpot))
         #Relative frequency
         unfair_game_rf = unfair_game_jackpot/num_roll
         #Show your results, comparing the two relative frequencies, in a simple bar chart.
         rf df = pd.DataFrame()
         df = pd.DataFrame({'Game':['Fair Game','Unfair Game'],'Relative Frequency':[fair_game_rf,unfair_game_rf]})
         df = df.set_index('Game')
         df.sort index().plot.bar();
        Number of jackpots for 3 fair dice: 256
```

Number of jackpots for 2 unfair and 1 fair dice : 370



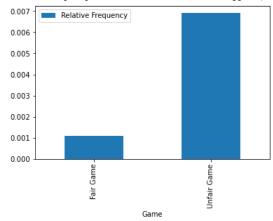
Scenario 2

Create a fair die and two unfair dice, all of six sides with the faces 1 through 6. One unfair die (Type 1) will weight 6 five times more than the others (i.e. it has weight of 5 and the others a weight of 1 each). The other unfair die (Type 2) will weight 1 five times more than the others. Play a game of 10000 rolls with 5 fair dice. Play a game of 10000 rolls with 2 unfair dice of type 1, 1 unfair die of type 2, and the rest fair dice. For each game, use an Analyzer object to determine the relative frequency of jackpots and show your results, comparing the two relative frequencies, in a simple bar chart. Also compute 10 most frequent combinations of faces for each game. Plot each of these as bar charts.

```
In [9]:
         # Code blocks with output
         faces = [1,2,3,4,5,6]
         num_roll = 10000
         #Fair Coin
         fair_die = Die(faces)
         fair_die_face_list = fair_die.faces
         fair_die.show()
         #Unfair Coin Type1
         unfair_coin_type1 = Die(faces)
         unfair_coin_face_list = unfair_coin_type1.faces
         unfair_coin_type1.change_weight(6,5)
         #Unfair Coin Type2
         unfair_coin_type2 = Die(faces)
         unfair coin face list = unfair coin type2.faces
         unfair_coin_type2.change_weight(1,5)
         # Play a game of 10000 rolls with 5 fair dice.
         fair_die_list = []
         for i in range(5):
             fair_die_list.append(fair_die)
         fair_game = Game(fair_die_list)
         fair game.play(num roll)
         fair_game.show()
         analyze_fair_game = Analyzer(fair_game.show())
         fair_game_jackpot = analyze_fair_game.jackpot()
         print("Number of jackpots for 5 fair dice : "+str(fair_game_jackpot))
         #Relative frequency
         fair_game_rf = fair_game_jackpot/num_roll
         #----#
         #Play a game of 10000 rolls with 2 unfair dice of type 1, 1 unfair die of type 2, and the rest fair dice.
         unfair_coin_list = [unfair_coin_type1,unfair_coin_type1,unfair_coin_type2,fair_die]
         unfair_game = Game(unfair_coin_list)
         unfair_game.play(num_roll)
         unfair_game.show()
         analyze_unfair_game = Analyzer(unfair_game.show())
unfair_game_jackpot = analyze_unfair_game.jackpot()
         print("Number of jackpots for 2 unfair dice of type 1, 1 unfair die of type 2 and 1 fair dice: "+str(unfair_game_jackpot
         #Relative frequency
         unfair_game_rf = unfair_game_jackpot/num_roll
         #For each game, use an Analyzer object to determine the relative frequency of jackpots and show your results, comparing \mathfrak t
         rf_df = pd.DataFrame()
```

```
df = pd.DataFrame({'Game':['Fair Game','Unfair Game'],'Relative Frequency':[fair_game_rf,unfair_game_rf]})
df = df.set_index('Game')
df.sort_index().plot.bar();
```

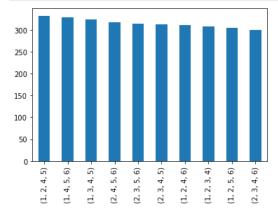
```
Number of jackpots for 5 fair dice : 11 Number of jackpots for 2 unfair dice of type 1, 1 unfair die of type 2 and 1 fair dice : 69
```



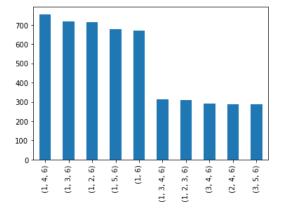
```
In [12]: #Also compute 10 most frequent combinations of faces for each game. Plot each of these as bar charts.

def get10_freq_combo(df):
    face_combos = []
    for i in range(len(df)):
        face_combo = df.iloc[i]
            face_combos.append(tuple(face_combo.drop_duplicates().sort_values()))
    pd.Series(face_combos).value_counts()[:10].plot.bar()

fair_game_frequest_combos = get10_freq_combo(fair_game.show())
```



```
In [11]: unfair_game_frequest_combos = get10_freq_combo(unfair_game.show())
```



Scenario 3

Create a "die" of letters from a to z with weights based on their frequency of usage. See Appendix for these weights. Play a game involving 5 of these dice with 1000 rolls. Generate 10 random samples of 10 from your data and count the number of times you see a word that looks like an English word in each sample. Keep a running count; this will result in an estimate of the percent of English words in the data.

```
In [13]: # Code blocks with output import enchant
```

```
english dict = enchant.Dict("en US")
 #Create a "die" of letters from a to z with weights based on their frequency of usage. weights from Appendix are stored i
 weights_df = pd.read_csv("weights.csv")
 faces = list(weights_df['Alphabet'])
 die = Die(faces)
die face list = die.faces
 for i in range(len(weights_df)):
    die.change_weight(weights_df.iloc[i].Alphabet,weights_df.iloc[i].Weight)
 die.show()
 #Play a game involving 5 of these dice with 1000 rolls.
 num roll = 1000
 die_list = []
 for i in range(5):
     die_list.append(die)
 game = Game(die list)
 game.play(num_roll)
 # Generate 10 random samples of 10 from your data and count the number of times you see a word that looks like an English
 check_word = lambda word :english_dict.check(word)
 word count =0
 words_list = []
 for i in range(1,11):
     samples = game.show().sample(10)
words = samples.apply(''.join ,axis=1)
     for word in words:
         if check word(word):
             words_list.append(word)
             word_count+=1
 print("English words in data : "+str(word count))
 if word count > 0:
     print(words list)
print("Percentage of English words : "+str(word_count)+"%")
English words in data : 2
['PATES', 'CHINA']
Percentage of English words : 2%
```

Directory Listing

!ls -lRF -o

A code block that executes the following bash command:

```
drwxr-sr-x 4 nyc2xu 3072 Nov 29 14:05 pkg_mc/
drwxr-sr-x 3 nyc2xu 3072 Nov 29 14:01 pkg_mc.egg-info/
-rw-r--r-- 1 nyc2xu 6043 Nov 29 13:23 README.md
-rw-r--r-- 1 nyc2xu
                   318 Nov 27 13:37 setup.py
-rw-r--r-- 1 nyc2xu 278 Nov 29 11:15 weights.csv
./pkg_mc:
-rw-r--r 1 nyc2xu 8575 Nov 29 13:59 montecarlo.py
-rw-r--r-- 1 nyc2xu 10173 Nov 29 14:05 montecarlo.pyc
drwxr-sr-x 2 nyc2xu 1024 Nov 29 14:09 pycache
./pkg_mc/__pycache__:
total 20
-rw-r--r-- 1 nyc2xu 186 Nov 29 13:56 __init__.cpython-38.pyc
-rw-r--r-- 1 nyc2xu 8569 Nov 29 14:09 montecarlo.cpython-38.pyc
./pkg_mc.egg-info:
total 20
-rw-r--r 1 nyc2xu 1 Nov 29 13:54 dependency_links.txt
-rw-r--r- 1 nyc2xu 243 Nov 29 14:01 PKG-INFO
-rw-r--r-- 1 nyc2xu 15 Nov 29 13:54 requires.txt
```

```
-rw-r--r-- 1 nyc2xu 207 Nov 29 13:54 SOURCES.txt
-rw-r--r-- 1 nyc2xu 7 Nov 29 13:54 top_level.txt
```

Installation Output Listing

A code block that executes the code to install your your package and outputs a successful installation.

```
In [15]:
```

```
# Installation commands
!pip install -e .
```

Defaulting to user installation because normal site-packages is not writeable Obtaining file:///sfs/qumulo/qhome/nyc2xu/Documents/MSDS/DS5100/DS5100-2022-08-nyc2xu/Project Requirement already satisfied: pandas>=0.15.1 in /sfs/applications/202206/software/standard/compiler/gcc/9.2.0/jupyter_co $\verb|nda/2020.11-py3.8/lib/python3.8/site-packages (from pkg-mc==0.1) (1.1.3)|$ $Requirement\ already\ satisfied:\ python-date util>=2.7.3\ in\ /sfs/applications/202206/software/standard/compiler/gcc/9.2.0/ju$ $pyter_conda/2020.11-py3.8/lib/python3.8/site-packages \ (from pandas>=0.15.1->pkg-mc==0.1) \ (2.8.1)$ Requirement already satisfied: pytz>=2017.2 in /sfs/applications/202206/software/standard/compiler/gcc/9.2.0/jupyter_cond $a/2020.11-py3.8/lib/python3.8/site-packages \ (from pandas>=0.15.1->pkg-mc==0.1) \ (2020.1)$ Requirement already satisfied: numpy>=1.15.4 in /sfs/applications/202206/software/standard/compiler/gcc/9.2.0/jupyter_con $\label{lib_python3.8} $$ da/2020.11-py3.8/lib/python3.8/site-packages (from pandas>=0.15.1->pkg-mc==0.1) (1.19.2) $$ da/2020.11-py3.8/lib/python3.8/site-packages (from pandas>=0.15.1->pkg-mc==0.1) $$ da/2020.11-py3.8/lib/python3.8/lib$ Requirement already satisfied: six>=1.5 in /sfs/applications/202206/software/standard/compiler/gcc/9.2.0/jupyter_conda/20 20.11-py3.8/lib/python3.8/site-packages (from python-dateutil>=2.7.3->pandas>=0.15.1->pkg-mc==0.1) (1.15.0) Installing collected packages: pkg-mc Attempting uninstall: pkg-mc Found existing installation: $pkg-mc \ 0.1$ Uninstalling pkg-mc-0.1: Successfully uninstalled pkg-mc-0.1Running setup.py develop for pkg-mc Successfully installed pkg-mc