# **Final Project Report**

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- This URL: a URL to the notebook source of this document

# Instructions

Follow the instructions in the Final Project isntructions and put your work in this notebook.

Total points for each subsection under **Deliverables** and **Scenarios** are given in parentheses.

Breakdowns of points within subsections are specified within subsection instructions as bulleted lists.

This project is worth **50 points**.

### **Deliverables**

## The Monte Carlo Module (10)

- URL included, appropriately named (1).
- Includes all three specified classes (3).
- Includes at least all 12 specified methods (6; .5 each).

Put the URL to your GitHub repo here.

Repo URL:https://github.com/AddisonGambhir/ajg7pk\_ds5100\_montecarlo/tree/main

Paste a copyy of your module here.

NOTE: Paste as text, not as code. Use triple backticks to wrap your code blocks.

```
import pandas as pd
import numpy as np
import random

class Die:
    """
    Fibricates a 'die' object that has specified faces and weights.

    Attributes:
        _die_data (pandas.DataFrame): This is the dataframe that stores
face values and weights.
    """
```

```
def __init__(self, faces: np.ndarray):
       Initializes the die object with specified 'faces'.
            faces (numpy.ndarray): An array containing distinct face
values of the die.
       Raises:
            TypeError: If faces is not a NumPy array.
            ValueError: If faces do not have distinct values.
       if not isinstance(faces, np.ndarray):
            raise TypeError("Input must be a NumPy array.")
       if len(faces) != len(np.unique(faces)):
            raise ValueError("Faces must have distinct values.")
       weights = np.ones_like(faces, dtype=float)
       data = {'Weights': weights}
       if np.issubdtype(faces.dtype, np.number):
            self. die data = pd.DataFrame(data, index=faces)
       else:
            self. die data = pd.DataFrame(data, index=faces.astype(str))
   def set weight(self, face, weight):
       This controls the weight for a specific face value.
       Arguments:
            face: The face value for which the weight is to be set.
           weight: The weight value assigned.
       Raises:
            IndexError: If the face value is invalid.
            TypeError: If the weight is not a numeric value.
       if face not in self. die data.index:
            raise IndexError("Invalid face value.")
        if not (isinstance(weight, (int, float)) or
np.issubdtype(type(weight), np.number)):
            raise TypeError("Weight must be a numeric value or castable
as numeric.")
       self. die data.loc[face, 'Weights'] = weight
   def roll(self, num rolls: int = 1) -> list:
       This method simulates rolling of the die multiple times.
       Argument:
            num rolls (int): The number of times the die is rolled.
```

```
Returns:
            list: A list of outcomes from rolling the die.
        outcomes = random.choices(self._die_data.index,
weights=self._die_data['Weights'], k=num_rolls)
        return outcomes
    def get die state(self) -> pd.DataFrame:
        This will return a copy of the die's data frame.
        Returns:
            pandas.DataFrame: A copy of the die data frame.
        return self._die_data.copy()
class Game:
    The game class contains a game consisting of rolling multiple dice.
    Attributes:
        dice: A list of Die objects representing the dice in the game.
        play data: Data frame that stores the results of the most recent
play.
    def __init__(self, dice):
        This initializes the Game class with a list of similar dice.
        This also checks if the list contains Die objects
        Argusment:
            dice (list): A list of already instantiated similar dice (Die
objects).
        # Check if the list contains Die objects
        if not all(isinstance(die, Die) for die in dice):
            raise ValueError('All elements in the list must be instances
of the Die class.')
        # Check if all dice have the same faces
        faces = set(dice[0].get_die_state().index)
        if not all(set(die.get die state().index) == faces for die in
dice):
            raise ValueError('All dice must have the same faces.')
        self.dice = dice
        self.play data = pd.DataFrame() # Private data frame to store
play results
    def play(self, times):
        This method simulates playing the game by rolling the dice a
```

specified number of times.

```
Argsuments:
            times (int): The number of times instructed to roll the dice.
        rolls = []
        for i in range(times):
            roll result = []
            for die in self.dice:
                roll = die.roll()[0]
                roll result.append(roll)
            rolls.append(roll_result)
        columns = [f"Die_{i}" for i in range(1, len(self.dice) + 1)]
        self.play data = pd.DataFrame(rolls, columns=columns)
    def show results(self, form='wide'):
        This method displays the results of the most recent play.
       Args:
            Format options: Options include: 'wide' (default) or 'narrow'
        Returns:
            pandas.DataFrame: The play results data frame.
        Raises:
            ValueError: If an invalid option is provided for the format.
        if form == 'wide':
            return self.play data.copy()
        elif form == 'narrow':
            narrow data = pd.DataFrame()
            for idx, col in enumerate(self.play_data.columns):
                rolls = self.play data[col].tolist()
                roll_numbers = list(range(1, len(rolls) + 1))
                multi index = pd.MultiIndex.from tuples([(r, col) for r
in roll_numbers], names=['Roll', 'Die'])
                df = pd.DataFrame(rolls, index=multi index, columns=
['Outcome'])
                narrow data = pd.concat([narrow data, df])
            return narrow_data
        else:
            raise ValueError("Invalid option for 'form'. Please choose
'wide' or 'narrow'.")
class Analyzer:
    11 11 11
    Analyzes the results of a single game and computes various
descriptive properties.
    Attributes:
        game (Game): A Game object whose results will be analyzed.
```

.....

```
def __init__(self, game):
        Initializes an Analyzer object with a Game object.
        Args:
            game (Game): A Game object whose results will be analyzed.
        Raises:
            ValueError: If the input is not a Game object.
        if not isinstance(game, Game):
            raise ValueError("Input must be a Game object.")
        self.game = game
    def jackpot(self):
        .. .. ..
        Jackpot computes the number of jackpot occurrences in the game.
        Returns:
            int: The number of jackpots in the game.
        jackpot_count = 0
        results = self.game.show_results('wide')
        for i, row in results.iterrows():
            if all(value == row.iloc[0] for value in row):
                jackpot count += 1
        return jackpot_count
    def face_counts_per_roll(self):
        Counts how many times a given face is rolled in each event.
        Returns: A data frame with an index of the roll number, face
values as columns,
        and count values in the cells (in the 'wide' format).
        # Getting the distinct faces
        faces = self.game.show results().melt().value.unique()
        # Initializing a DataFrame to store the counts
        counts df = pd.DataFrame(columns=faces,
index=range(self.game.show_results().shape[0]))
        # Looping through the rolls and computing the counts
        for roll_number, row in self.game.show_results().iterrows():
            counts = row.value counts()
            for face in faces:
                counts df.loc[roll number, face] = counts.get(face, 0)
        return counts df
```

```
def combo_count(self):
       Determines unique combinations of faces rolled, along with their
counts.
       Returns:
            pandas DataFrame with distinct combinations and associated
counts.
       results = self.game.show results('narrow')
       combo_counts = results.groupby('Roll')
['Outcome'].apply(tuple).value counts().reset index()
        combo_counts.columns = ['Combo', 'Count']
       return combo_counts
   def permutation_count(self):
       This method will compute unique permutations of faces rolled,
along with their counts.
       Returns:
            pandas DataFrame with distinct permutations and associated
counts.
       results = self.game.show_results('narrow')
       permutation counts = results.groupby('Roll')
['Outcome'].apply(list).value counts().reset index()
       permutation counts.columns = ['Permutation', 'Count']
       return permutation counts
```

### **Unitest Module (2)**

Paste a copy of your test module below.

NOTE: Paste as text, not as code. Use triple backticks to wrap your code blocks.

- All methods have at least one test method (1).
- Each method employs one of Unittest's Assert methods (1).

```
import unittest
import numpy as np
from montecarlo import Die, Game, Analyzer
import sys

class TestDieGameAnalyzer(unittest.TestCase):
    # Test methods for Die class
    def test_die_creation(self):
        faces = np.array([1, 2, 3, 4, 5, 6])
        die = Die(faces)
```

```
self.assertIsNotNone(die)
def test set weight(self):
   faces = np.array([1, 2, 3, 4, 5, 6])
    die = Die(faces)
    die.set_weight(3, 5)
    self.assertEqual(die.get die state().loc[3, 'Weights'], 5)
def test roll(self):
   faces = np.array([1, 2, 3, 4, 5, 6])
   die = Die(faces)
    outcomes = die.roll(5)
    self.assertEqual(len(outcomes), 5)
def test get die state(self):
   faces = np.array([1, 2, 3, 4, 5, 6])
    die = Die(faces)
    state = die.get die state()
    self.assertTrue((state.index == faces).all())
# Test methods for Game class
def test_game_creation(self):
    die1 = Die(np.array([1, 2, 3, 4, 5, 6]))
    die2 = Die(np.array([1, 2, 3, 4, 5, 6]))
    game = Game([die1, die2])
    self.assertIsNotNone(game)
def test_game_play(self):
    die1 = Die(np.array([1, 2, 3, 4, 5, 6]))
    die2 = Die(np.array([1, 2, 3, 4, 5, 6]))
    game = Game([die1, die2])
    game.play(5)
    self.assertEqual(game.play data.shape[0], 5)
def test show results(self):
   die1 = Die(np.array([1, 2, 3, 4, 5, 6]))
    die2 = Die(np.array([1, 2, 3, 4, 5, 6]))
    game = Game([die1, die2])
    game.play(5)
    results = game.show results('wide')
    self.assertEqual(results.shape[0], 5)
# Test methods for Analyzer class
def test analyzer creation(self):
    die1 = Die(np.array([1, 2, 3, 4, 5, 6]))
    die2 = Die(np.array([1, 2, 3, 4, 5, 6]))
    game = Game([die1, die2])
    analyzer = Analyzer(game)
    self.assertIsNotNone(analyzer)
def test analyzer jackpot(self):
    die1 = Die(np.array([1, 2, 3]))
    die2 = Die(np.array([1, 2, 3]))
```

```
game = Game([die1, die2])
        game.play(10)
        analyzer = Analyzer(game)
        self.assertTrue(analyzer.jackpot() >= 0)
    def test_analyzer_face_counts_per_roll(self):
        die1 = Die(np.array([1, 2, 3]))
        die2 = Die(np.array([1, 2, 3]))
        game = Game([die1, die2])
        game.play(10)
        analyzer = Analyzer(game)
        counts = analyzer.face_counts_per_roll()
        self.assertEqual(counts.shape[0], 10)
    def test_analyzer_combo_count(self):
        die1 = Die(np.array([1, 2, 3]))
        die2 = Die(np.array([1, 2, 3]))
        game = Game([die1, die2])
        game.play(10)
        analyzer = Analyzer(game)
        combos = analyzer.combo count()
        self.assertTrue(combos.shape[0] > 0)
    def test_analyzer_permutation_count(self):
        die1 = Die(np.array([1, 2, 3]))
        die2 = Die(np.array([1, 2, 3]))
        game = Game([die1, die2])
        game.play(10)
        analyzer = Analyzer(game)
        permutations = analyzer.permutation_count()
        self.assertTrue(permutations.shape[0] > 0)
# Execute the tests
#if __name__ == '__main__':
     unittest.main()
#this command didn't push the results of the test to the .txt file so I
had to check stack overflow
if name == ' main ':
unittest.TextTestRunner(stream=sys.stdout).run(unittest.TestLoader().loadTestsFr
```

#### **Unittest Results (3)**

Put a copy of the results of running your tests from the command line here.

Again, paste as text using triple backticks.

• All 12 specified methods return OK (3; .25 each).

# Import (1)

Import your module here. This import should refer to the code in your package directory.

• Module successufly imported (1).

```
In [1]: from montecarlo.montecarlo import Die,Game,Analyzer
# e.g. import montecarlo.montecarlo
```

## Help Docs (4)

Show your docstring documentation by applying help() to your imported module.

- All methods have a docstring (3; .25 each).
- All classes have a docstring (1; .33 each).

```
help(Die)
In [2]:
         help(Game)
         help(Analyzer)
        Help on class Die in module montecarlo.montecarlo:
        class Die(builtins.object)
            Die(faces: numpy.ndarray)
            Fibricates a 'die' object that has specified faces and weights.
            Attributes:
                 _die_data (pandas.DataFrame): This is the dataframe that stores face values and
        weights.
            Methods defined here:
             init (self, faces: numpy.ndarray)
                Initializes the die object with specified 'faces'.
                Args:
                    faces (numpy.ndarray): An array containing distinct face values of the die.
                    TypeError: If faces is not a NumPy array.
                    ValueError: If faces do not have distinct values.
            get_die_state(self) -> pandas.core.frame.DataFrame
                This will return a copy of the die's data frame.
                Returns:
                     pandas.DataFrame: A copy of the die data frame.
```

```
roll(self, num rolls: int = 1) -> list
        This method simulates rolling of the die multiple times.
        Argument:
            num rolls (int): The number of times the die is rolled.
            list: A list of outcomes from rolling the die.
    set_weight(self, face, weight)
        This controls the weight for a specific face value.
        Arguments:
            face: The face value for which the weight is to be set.
            weight: The weight value assigned.
        Raises:
            IndexError: If the face value is invalid.
            TypeError: If the weight is not a numeric value.
    Data descriptors defined here:
    dict
        dictionary for instance variables (if defined)
     weakref
        list of weak references to the object (if defined)
Help on class Game in module montecarlo.montecarlo:
class Game(builtins.object)
    Game(dice)
    The game class contains a game consisting of rolling multiple dice.
    Attributes:
        dice: A list of Die objects representing the dice in the game.
        play_data: Data frame that stores the results of the most recent play.
   Methods defined here:
    __init__(self, dice)
        This initializes the Game class with a list of similar dice.
        This also checks if the list contains Die objects
            dice (list): A list of already instantiated similar dice (Die objects).
    play(self, times)
        This method simulates playing the game by rolling the dice a specified number of
times.
        Argsuments:
            times (int): The number of times instructed to roll the dice.
    show_results(self, form='wide')
        This method displays the results of the most recent play.
        Args:
            Format options: Options include: 'wide' (default) or 'narrow'
        Returns:
            pandas.DataFrame: The play results data frame.
        Raises:
```

```
ValueError: If an invalid option is provided for the format.
   Data descriptors defined here:
    dict
        dictionary for instance variables (if defined)
     weakref
        list of weak references to the object (if defined)
Help on class Analyzer in module montecarlo.montecarlo:
class Analyzer(builtins.object)
    Analyzer(game)
   Analyzes the results of a single game and computes various descriptive properties.
   Attributes:
        game (Game): A Game object whose results will be analyzed.
   Methods defined here:
    init (self, game)
        Initializes an Analyzer object with a Game object.
        Args:
            game (Game): A Game object whose results will be analyzed.
        Raises:
            ValueError: If the input is not a Game object.
    combo count(self)
        Determines unique combinations of faces rolled, along with their counts.
        Returns:
            pandas DataFrame with distinct combinations and associated counts.
    face_counts_per_roll(self)
        Counts how many times a given face is rolled in each event.
        Returns: A data frame with an index of the roll number, face values as columns,
        and count values in the cells (in the 'wide' format).
    jackpot(self)
        Jackpot computes the number of jackpot occurrences in the game.
            int: The number of jackpots in the game.
    permutation count(self)
        This method will compute unique permutations of faces rolled, along with their c
ounts.
        Returns:
            pandas DataFrame with distinct permutations and associated counts.
   Data descriptors defined here:
        dictionary for instance variables (if defined)
     weakref
```

list of weak references to the object (if defined)

## README.md File (3)

Provide link to the README.md file of your project's repo.

- Metadata section or info present (1).
- Synopsis section showing how each class is called (1). (All must be included.)
- API section listing all classes and methods (1). (All must be included.)

URL: https://github.com/AddisonGambhir/ajg7pk\_ds5100\_montecarlo/blob/main/README.md

# Successful installation (2)

Put a screenshot or paste a copy of a terminal session where you successfully install your module with pip.

If pasting text, use a preformatted text block to show the results.

- Installed with pip (1).
- Successfully installed message appears (1).

```
# Command to install:$
bash-4.2$pip3 install -e .

# Terminal Output:

Defaulting to user installation because normal site-packages is not writeable
Obtaining file:///sfs/qumulo/qhome/ajg7pk/Documents/MSDS/ds5100-ajg7pk/lessons/M15
Installing collected packages: montecarlo
   Running setup.py develop for montecarlo
Successfully installed montecarlo
```

## **Scenarios**

Use code blocks to perform the tasks for each scenario.

Be sure the outputs are visible before submitting.

# Scenario 1: A 2-headed Coin (9)

Task 1. Create a fair coin (with faces \$H\$ and \$T\$) and one unfair coin in which one of the faces has a weight of \$5\$ and the others \$1\$.

- Fair coin created (1).
- Unfair coin created with weight as specified (1).

```
import numpy as np
import pandas as pd

#fair coin
fair_faces = np.array(['H', 'T'])
fair_coin = Die(fair_faces)
print("Fair coin Info:", fair_coin.get_die_state())

#unfair coin
unfair_faces = np.array(['H', 'T'])
unfair_coin = Die(unfair_faces)
unfair_coin.set_weight('H', 5)
print("Unfair coin Info:", unfair_coin.get_die_state())
```

```
Fair coin Info: Weights
H 1.0
T 1.0
Unfair coin Info: Weights
H 5.0
T 1.0
```

Task 2. Play a game of \$1000\$ flips with two fair dice.

• Play method called correctly and without error (1).

```
In [4]: # creating fair coins
    faces = np.array(['H', 'T'])
    fair_coin = Die(faces)

unfair_coin = Die(faces)
unfair_coin.set_weight(face='H', weight=5)

game = Game([fair_coin, unfair_coin])

game.play(1000)

results_df = game.show_results()

print("Results of the game with 1000 flips:")
results_df
```

Results of the game with 1000 flips:

```
Out[4]: Die_1 Die_2

0 H H

1 H H

2 H H

3 H T

4 H H

... ... ...
```

	Die_1	Die_2
996	Н	Т
997	Т	Н
998	Т	Т
999	Н	Н

1000 rows × 2 columns

Task 3. Play another game (using a new Game object) of \$1000\$ flips, this time using two unfair dice and one fair die. For the second unfair die, you can use the same die object twice in the list of dice you pass to the Game object.

- New game object created (1).
- Play method called correctty and without error (1).

```
In [5]: #from die_game import Die, Game
import numpy as np

faces = np.array(['H', 'T'])
fair_coin = Die(faces)

unfair_coin = Die(faces)

unfair_coin.set_weight(face='H', weight=5)

new_game = Game([unfair_coin, unfair_coin, fair_coin])
new_game.play(1000)

new_results_df = new_game.show_results()
new_results_df
```

Out[5]:		Die_1	Die_2	Die_3
	0	Н	Н	Т
	1	Н	Н	Н
	2	Т	Н	Т
	3	Н	Т	Т
	4	Н	Т	Т
	995	Н	Н	Н
	996	Н	Н	Н
	997	Н	Н	Т
	998	Н	Н	Т
	999	Т	Н	Н

1000 rows × 3 columns

Task 4. For each game, use an Analyzer object to determine the raw frequency of jackpots — i.e. getting either all \$H\$s or all \$T\$s.

- Analyzer objecs instantiated for both games (1).
- Raw frequencies reported for both (1).

```
In [6]: analyzer_results = Analyzer(game) # Game results from task 2
    analyzer_new_results = Analyzer(new_game) # Game results from task 3

    raw_frequency_results = analyzer_results.jackpot()
    raw_frequency_new_results = analyzer_new_results.jackpot()

# Print raw frequencies
    print("Raw frequency of jackpots for task 2:", raw_frequency_results)
    print("Raw frequency of jackpots for task3:", raw_frequency_new_results)
```

Raw frequency of jackpots for task 2: 513 Raw frequency of jackpots for task3: 369

Task 5. For each analyzer, compute relative frequency as the number of jackpots over the total number of rolls.

• Both relative frequencies computed (1).

```
In [7]: total_rolls = 1000 # Since each game consists of 1000 flips

# Compute relative frequencies for both games
relative_frequency_results = raw_frequency_results / total_rolls
relative_frequency_new_results = raw_frequency_new_results / total_rolls

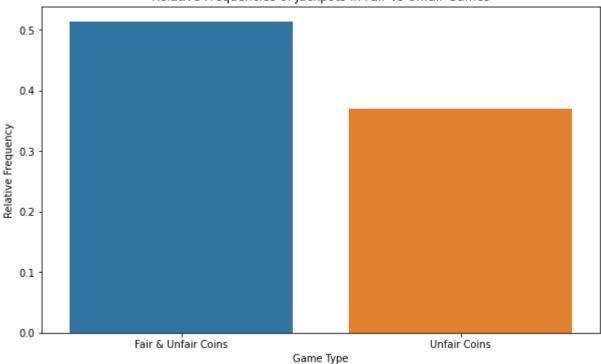
print("Relative frequency of jackpots for the fair game:", relative_frequency_results)
print("Relative frequency of jackpots for the unfair game:", relative_frequency_new_res
```

Relative frequency of jackpots for the fair game: 0.513 Relative frequency of jackpots for the unfair game: 0.369

Task 6. Show your results, comparing the two relative frequencies, in a simple bar chart.

• Bar chart plotted and correct (1).

#### Relative Frequencies of Jackpots in Fair vs Unfair Games



# Scenario 2: A 6-sided Die (9)

Task 1. Create three dice, each with six sides having the faces 1 through 6.

• Three die objects created (1).

```
faces=np.array([1,2,3,4,5,6])
In [9]:
         die1 = Die(faces)
         die2 = Die(faces)
         die3 = Die(faces)
         print("die1 Info:", die1.get_die_state())
         print("die2 Info:", die2.get_die_state())
         print("die3 Info:", die3.get_die_state())
        die1 Info:
                       Weights
                1.0
         2
                1.0
         3
                1.0
         4
                1.0
         5
                1.0
        6
                1.0
        die2 Info:
                       Weights
                1.0
                1.0
        3
                1.0
        4
                1.0
         5
                1.0
                1.0
        die3 Info:
                       Weights
        1
                1.0
         2
                1.0
         3
                1.0
        4
                1.0
         5
                1.0
                1.0
```

Task 2. Convert one of the dice to an unfair one by weighting the face \$6\$ five times more than the other weights (i.e. it has weight of 5 and the others a weight of 1 each).

• Unfair die created with proper call to weight change method (1).

```
In [10]: die1.set_weight(face=6, weight=5)
    die1.get_die_state()
```

# Out[10]: Weights 1 1.0 2 1.0 3 1.0 4 1.0 5 1.0 6 5.0

Task 3. Convert another of the dice to be unfair by weighting the face \$1\$ five times more than the others.

• Unfair die created with proper call to weight change method (1).

```
In [11]: die2.set_weight(face=1, weight=5)
    die2.get_die_state()
```

```
Out[11]: Weights

1 5.0
2 1.0
3 1.0
4 1.0
5 1.0
6 1.0
```

Task 4. Play a game of \$10000\$ rolls with \$5\$ fair dice.

- Game class properly instantiated (1).
- Play method called properly (1).

```
In [12]: #5 dice
    faces = np.array([1, 2, 3, 4, 5, 6])
    fair_dice = [Die(faces) for i in range(5)]

    game_with_fair_dice = Game(fair_dice)
    game_with_fair_dice.play(10000)

    print("Game Results:")
    game_with_fair_dice.show_results()
```

Game Results:

Out[12]:		Die_1	Die_2	Die_3	Die_4	Die_5
	0	1	6	5	4	3
	1	1	1	6	5	3
	2	5	4	1	1	6
	3	2	6	6	1	2
	4	1	2	1	1	1
	•••					
	9995	5	3	3	1	2
	9996	5	2	5	3	5
	9997	4	6	5	6	6
	9998	6	1	3	6	1
	9999	3	6	2	1	4

10000 rows × 5 columns

Task 5. Play another game of \$10000\$ rolls, this time with \$2\$ unfair dice, one as defined in steps #2 and #3 respectively, and \$3\$ fair dice.

- Game class properly instantiated (1).
- Play method called properly (1).

```
In [13]: #dice
    fair_dice_for_mixed_game = [Die(faces) for i in range(3)]
    mixed_dice = fair_dice_for_mixed_game + [die1, die2]

#game
    game_with_mixed_dice = Game(mixed_dice)
    game_with_mixed_dice.play(10000)
    results_df_mixed_dice = game_with_mixed_dice.show_results()
    print("Game results:")

results_df_mixed_dice
```

Game results:

Out[13]:		Die_1	Die_2	Die_3	Die_4	Die_5	
	0	2	2	6	6	5	
	1	6	6	2	6	1	
	2	4	4	4	1	1	
	3	5	1	5	6	4	
	4	2	2	5	6	4	
	•••						
	9995	3	2	3	4	4	

	Die_1	Die_2	Die_3	Die_4	Die_5
9996	3	2	4	3	1
9997	4	3	1	4	1
9998	6	5	1	6	1
9999	1	4	4	1	1

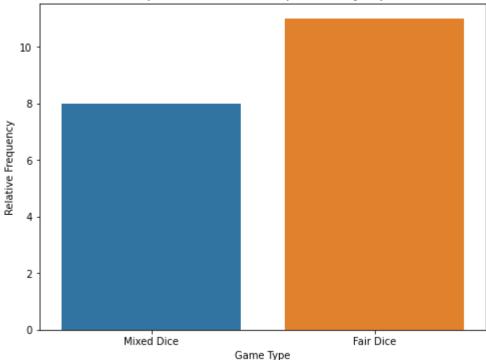
10000 rows × 5 columns

Task 6. 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.

- Jackpot methods called (1).
- Graph produced (1).

```
In [14]:
          #analyzing games
          analyzer_mixed_game = Analyzer(game_with_mixed_dice)
          analyzer_fair_game = Analyzer(game_with_fair_dice)
          relative_frequency_mixed = analyzer_mixed_game.jackpot()
          relative_frequency_fair = analyzer_fair_game.jackpot()
          #making data
          df1 = pd.DataFrame({
               'Game': ['Mixed Dice', 'Fair Dice'],
              'Relative Frequency': [relative_frequency_mixed, relative_frequency_fair]
          })
          #plot
          plt.figure(figsize=(8, 6))
          sns.barplot(x='Game', y='Relative Frequency', data=df1)
          plt.title('Comparison of Relative Frequencies of Jackpots')
          plt.ylabel('Relative Frequency')
          plt.xlabel('Game Type')
          plt.show()
```

#### Comparison of Relative Frequencies of Jackpots



## Scenario 3: Letters of the Alphabet (7)

Task 1. Create a "die" of letters from \$A\$ to \$Z\$ with weights based on their frequency of usage as found in the data file english\_letters.txt . Use the frequencies (i.e. raw counts) as weights.

- Die correctly instantiated with source file data (1).
- Weights properly applied using weight setting method (1).

```
In [15]: my_txt = 'english_letters.txt'
    letter_freq_data = pd.read_csv(my_txt, sep=" ", header=None)
    letter_freq_data.columns = ['Letter', 'Weight']

# faces und weights
    faces = letter_freq_data['Letter'].values
    weights = letter_freq_data['Weight'].values

letter_die = Die(faces)
    for face, weight in zip(faces, weights):
        letter_die.set_weight(face, weight)

print("Letter die info:")
    letter_die.get_die_state()
```

Letter die info:

#### Out[15]:

#### Weights

- **E** 529117365.0
- **T** 390965105.0
- **A** 374061888.0
- **O** 326627740.0

#### Weights

- **I** 320410057.0
- **N** 313720540.0
- **S** 294300210.0
- **R** 277000841.0
- **H** 216768975.0
- **L** 183996130.0
- **D** 169330528.0
- **C** 138416451.0
- **U** 117295780.0
- **M** 110504544.0
- **F** 95422055.0
- **G** 91258980.0
- **P** 90376747.0
- **W** 79843664.0
- **Y** 75294515.0
- **B** 70195826.0
- **V** 46337161.0
- **K** 35373464.0
- **J** 9613410.0
- **X** 8369915.0
- **Z** 4975847.0
- **Q** 4550166.0

Task 2. Play a game involving \$4\$ of these dice with \$1000\$ rolls.

• Game play method properly called (1).

```
In [16]: letter_dice = [Die(faces) for i in range(4)]

for die in letter_dice:
    for face, weight in zip(faces, weights):
        die.set_weight(face, weight)

game_4d = Game(letter_dice)
game_4d.play(1000)

# Show the results
print("Game results:")
game_4d.show_results()
```

Game results:

Out[16]:		Die_1	Die_2	Die_3	Die_4
	0	С	I	А	Т
	1	Н	Т	V	Е
	2	R	R	1	U
	3	L	Н	Е	Е
	4	Н	Т	А	Е
	•••				
	995	W	0	R	Е
	996	В	Ν	Т	R
	997	Е	V	L	Е
	998	Е	Т	0	N
	999	С	Α	Р	Е

1000 rows × 4 columns

Task 3. Determine how many permutations in your results are actual English words, based on the vocabulary found in scrabble\_words.txt.

- Use permutation method (1).
- Get count as difference between permutations and vocabulary (1).

Number of permutations that are actual English words: {45}

Task 4. Repeat steps #2 and #3, this time with \$5\$ dice. How many actual words does this produce? Which produces more?

- Successfully repreats steps (1).
- Identifies parameter with most found words (1).

```
In [18]: def play_game_and_count_words(num_dice, vocabulary):
    dice = [letter_die] * num_dice

    game_letter = Game(dice)

    game_letter.play(1000)
```

```
analyzer = Analyzer(game_letter)
    permutations = analyzer.permutation_count()

count_actual_words = sum(1 for permutation in permutations['Permutation'] if ''.joi
    return count_actual_words

count_4_dice = play_game_and_count_words(4, vocabulary)

count_5_dice = play_game_and_count_words(5, vocabulary)

print('Words produced from 4 dice:',count_4_dice)
    print('Words produced from 5 dice:',count_5_dice)

print("4 dice produce more actual words.")
```

Words produced from 4 dice: 68 Words produced from 5 dice: 7 4 dice produce more actual words.

# **Submission**

When finished completing the above tasks, save this file to your local repo (and within your project), and them push it to your GitHub repo.

Then convert this file to a PDF and submit it to GradeScope according to the assignment instructions in Canvas.

```
In [ ]:
```