

Scenario_3–Roman_Alphabet

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1 Scenario 3: Generating English Words

1. Create a die of letters from ‘A’ to ‘Z’ with weights based on their frequency of usage. See ‘Letters_and_Weights.txt’ for a tab-separated data frame of letters and weights.

```
[1]: import numpy as np
import pandas as pd

data_frame_of_letters_and_weights = pd.read_csv('Letter_Weights.txt', delimiter_
↪= '\t', header = None).rename(columns = {0: 'letter', 1: 'weight'}).
↪astype({'letter': str, 'weight': np.float64})
data_frame_of_letters_and_weights
```

```
[1]:   letter  weight
0      A    8.4966
1      B    2.0720
2      C    4.5388
3      D    3.3844
4      E   11.1607
5      F    1.8121
6      G    2.4705
7      H    3.0034
8      I    7.5448
9      J    0.1965
10     K    1.1016
11     L    5.4893
12     M    3.0129
13     N    6.6544
14     O    7.1635
15     P    3.1671
16     Q    0.1962
17     R    7.5809
18     S    5.7351
19     T    6.9509
20     U    3.6308
21     V    1.0074
22     W    1.2899
23     X    0.2902
```

```
24      Y    1.7779
25      Z    0.2722
```

```
[2]: from montecarlosimulator import Die

def generate_die():
    array_of_faces = np.array(data_frame_of_letters_and_weights['letter'],
                               dtype = str)
    die = Die(array_of_faces)
    list_of_weights = data_frame_of_letters_and_weights['weight'].to_list()
    for i in range(0, len(array_of_faces)):
        face = array_of_faces[i]
        weight = list_of_weights[i]
        die.change_weight(face, weight)
    return die

die = generate_die()
die.show()
```

```
[2]:   face  weight
0     A  8.4966
1     B  1.0000
2     C  1.0000
3     D  1.0000
4     E  1.0000
5     F  1.0000
6     G  1.0000
7     H  1.0000
8     I  1.0000
9     J  1.0000
10    K  1.0000
11    L  1.0000
12    M  1.0000
13    N  1.0000
14    O  1.0000
15    P  1.0000
16    Q  1.0000
17    R  1.0000
18    S  1.0000
19    T  1.0000
20    U  1.0000
21    V  1.0000
22    W  1.0000
23    X  1.0000
24    Y  1.0000
25    Z  1.0000
```

2. Play a game involving rolling 5 dice of letters from 'A' to 'Z' with weights based on their

frequency of usage and 1000 rolls.

```
[3]: from montecarlosimulator import Game

list_of_dice = []
for i in range(0, 5):
    die = generate_die()
    list_of_dice.append(die)
game = Game(list_of_dice)
game.play(1000)
data_frame_of_rolls_and_dice = game.show('wide')
data_frame_of_rolls_and_dice
```

```
[3]:      0  1  2  3  4
roll_index
0      A  O  A  A  V
1      Y  A  A  Z  S
2      A  A  M  X  A
3      A  B  L  G  J
4      I  A  P  O  R
...
995    G  D  U  K  C
996    S  U  D  I  A
997    C  T  A  R  X
998    A  K  Q  A  Y
999    J  E  M  C  F
```

[1000 rows x 5 columns]

3. Generate 10 random samples of 10 rows each from the data frame of rolls and dice returned by the game showing a data frame of rolls and dice. Keep a running count; this will result in an estimate of the percent of English words in the data.

```
[4]: for i in range(0, 10):
    sample = data_frame_of_rolls_and_dice.sample(n = 10, replace = True,
↪weights = None, random_state = None, axis = None, ignore_index = False)
    print(sample)
    print()
```

```
      0  1  2  3  4
roll_index
228    S  A  L  H  B
305    Q  A  L  X  W
253    A  U  A  G  D
75     Z  A  Y  A  T
490    D  K  Z  A  A
517    G  B  C  R  A
820    U  Q  P  A  H
121    A  D  L  I  J
```

598	S	Q	J	Y	B
659	M	N	Y	J	A

	0	1	2	3	4
--	---	---	---	---	---

roll_index

915	M	A	N	Q	S
42	J	G	R	A	Z
467	Z	A	A	A	L
6	K	V	A	A	A
86	E	V	X	K	A
376	C	D	H	E	A
947	O	V	H	A	Y
531	X	G	O	A	J
843	V	E	R	F	M
646	A	A	M	T	J

	0	1	2	3	4
--	---	---	---	---	---

roll_index

847	Y	E	K	J	D
36	D	E	A	V	N
676	H	Q	J	R	G
999	J	E	M	C	F
800	I	A	H	P	H
964	A	A	S	T	Y
768	Y	U	I	C	V
217	M	G	A	Z	H
868	A	J	V	B	N
161	Q	I	U	L	A

	0	1	2	3	4
--	---	---	---	---	---

roll_index

451	M	D	A	S	C
663	A	A	B	W	A
136	A	Y	E	A	A
107	B	J	T	L	R
832	Z	A	B	A	A
549	A	O	A	F	A
341	U	V	K	D	A
748	S	A	W	K	E
273	R	Y	C	F	A
548	O	S	A	X	G

	0	1	2	3	4
--	---	---	---	---	---

roll_index

972	G	A	M	Q	T
427	I	A	Z	E	M
638	W	A	X	J	P
481	A	H	D	Z	T

927	S	E	A	M	B
109	P	D	A	T	R
107	B	J	T	L	R
14	A	U	R	U	A
876	I	A	D	Q	A
196	E	L	S	A	K

0	1	2	3	4
---	---	---	---	---

roll_index

678	A	H	U	Y	K
285	Q	C	K	A	G
916	J	Y	A	A	A
849	X	A	I	E	U
538	D	T	X	E	Q
605	A	Q	T	J	A
581	Z	R	A	T	Y
668	Z	Q	V	X	D
58	C	O	P	A	A
984	M	Z	U	Z	J

0	1	2	3	4
---	---	---	---	---

roll_index

859	D	E	A	U	F
686	A	C	O	R	B
477	W	L	V	S	Q
423	P	W	A	W	X
112	U	W	M	C	S
909	Q	W	A	V	N
631	T	R	K	J	A
685	S	A	U	K	A
284	B	C	F	A	A
87	S	L	A	X	A

0	1	2	3	4
---	---	---	---	---

roll_index

319	D	N	A	O	Y
738	C	C	Q	A	I
851	J	H	U	I	I
614	A	Q	H	A	Q
259	A	G	W	X	A
995	G	D	U	K	C
691	R	Q	Y	F	I
240	T	E	W	H	B
723	A	A	B	M	R
83	L	A	A	F	A

0	1	2	3	4
---	---	---	---	---

roll_index

958	R	U	A	X	A
177	J	W	N	H	I
699	K	I	T	M	M
704	K	A	A	W	G
684	A	I	A	C	A
921	B	D	D	I	C
564	Y	B	I	A	A
409	A	U	C	A	Z
631	T	R	K	J	A
332	U	Z	U	U	B

	0	1	2	3	4
roll_index					
435	D	K	U	P	A
405	X	O	C	N	L
754	M	V	A	G	C
272	A	A	J	A	A
961	E	O	S	A	W
53	E	P	A	X	A
146	A	T	V	Y	T
234	A	R	Z	A	A
229	A	H	M	J	C
187	Q	H	U	A	F

By inspection and comparison with Tom Lever’s vocabulary, the number of English words in the samples is 0. The estimated probability of a word in the samples being English

$$P = \frac{0}{100} = 0$$

[]: