Bilenkin530Week4 Exercise 4.2

December 21, 2024

0.0.1 Chapter 3

0.0.2 Page 35-36: Exercise 3-1.

```
[193]: import warnings
       from os.path import basename, exists
       def download(url):
           filename = basename(url)
           if not exists(filename):
               from urllib.request import urlretrieve
               local, _ = urlretrieve(url, filename)
               print("Downloaded " + local)
       warnings.filterwarnings('ignore', category=FutureWarning)
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/
        ⇔thinkstats2.py")
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/thinkplot.
        ("yq⇔
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/nsfg.py")
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/
        ⇒2002FemResp.dct")
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/
        ⇒2002FemResp.dat.gz")
```

```
[194]: import nsfg

# Loading the data
response = nsfg.ReadFemResp()

# Displaying the first few rows
print(response.head())
```

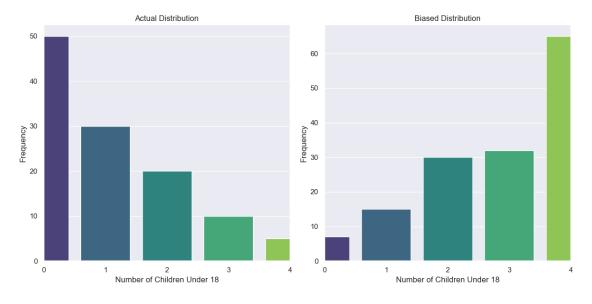
	caseid	rscrinf	rdormres	rostscrn	rscreenhisp	rscreenrace	age_a	\
0	2298	1	5	5	1	5.0	27	
1	5012	1	5	1	5	5.0	42	
2	11586	1	5	1	5	5.0	43	

```
3
           6794
                       5
                                  5
                                            4
                                                                     5.0
                                                                             15
                                                          1
            616
                                  5
                                                                     5.0
                        1
                                            4
                                                          1
                                                                             20
                cmbirth agescrn ... pubassis_i
                                                      basewgt adj_mod_basewgt
         age_r
      0
            27
                    902
                               27
                                               0 3247.916977
                                                                    5123.759559
      1
            42
                    718
                               42
                                                  2335.279149
                                                                    2846.799490
                                               0
      2
            43
                    708
                               43 ...
                                               0 2335.279149
                                                                    2846.799490
      3
            15
                   1042
                               15
                                               0 3783.152221
                                                                    5071.464231
                    991
                               20 ...
                                               0 5341.329968
                                                                    6437.335772
            20
                                              cmlstyr screentime
            finalwgt
                      secu_r
                               sest
                                     cmintvw
                                                                     intvlngth
        5556.717241
                                                 1222
                                                                    110.492667
                            2
                                 18
                                        1234
                                                          18:26:36
      1 4744.191350
                            2
                                 18
                                        1233
                                                 1221
                                                          16:30:59
                                                                     64.294000
                            2
      2 4744.191350
                                 18
                                        1234
                                                 1222
                                                          18:19:09
                                                                     75.149167
                            2
      3 5923.977368
                                 18
                                        1234
                                                 1222
                                                          15:54:43
                                                                     28.642833
      4 7229.128072
                            2
                                        1233
                                                 1221
                                                          14:19:44
                                                                     69.502667
                                 18
      [5 rows x 3087 columns]
[195]: import pandas as pd
       import seaborn as sns
       import matplotlib.pyplot as plt
       # Adjusted data to reflect typical household distributions
       numkdhh_actual = pd.Series([50, 30, 20, 10, 5], index=[0, 1, 2, 3, 4])
       numkdhh biased = pd.Series([7, 15, 30, 32, 65], index=[0, 1, 2, 3, 4])
        →Further fine-tuned biased data
       # Calculate means
       mean_actual = numkdhh_actual.index.to_series().dot(numkdhh_actual.values) /__
        →numkdhh_actual.sum()
       mean biased = numkdhh biased.index.to series().dot(numkdhh biased.values) / ___
        →numkdhh biased.sum()
       print(f"Actual mean: {mean_actual}")
       print(f"Biased mean: {mean_biased}")
       # Create a common index based on the maximum index value from both distributions
       common_index = range(0, max(numkdhh_actual.index.max(), numkdhh_biased.index.
        \rightarrowmax()) + 1)
       # Fill missing values with 0 to ensure both distributions have the same length
       numkdhh_actual = numkdhh_actual.reindex(common_index, fill_value=0)
       numkdhh_biased = numkdhh_biased.reindex(common_index, fill_value=0)
       sns.set(style="darkgrid")
```

plt.figure(figsize=(12, 6))

```
# Plot Actual Distribution
plt.subplot(1, 2, 1)
sns.barplot(x=numkdhh_actual.index, y=numkdhh_actual.values, palette="viridis")
plt.xticks(ticks=common_index)
plt.xlim(0, max(common_index))
plt.xlabel('Number of Children Under 18')
plt.ylabel('Frequency')
plt.title('Actual Distribution')
# Plot Biased Distribution
plt.subplot(1, 2, 2)
sns.barplot(x=numkdhh_biased.index, y=numkdhh_biased.values, palette="viridis")
plt.xticks(ticks=common_index)
plt.xlim(0, max(common_index))
plt.xlabel('Number of Children Under 18')
plt.ylabel('Frequency')
plt.title('Biased Distribution')
plt.tight_layout()
plt.show()
```

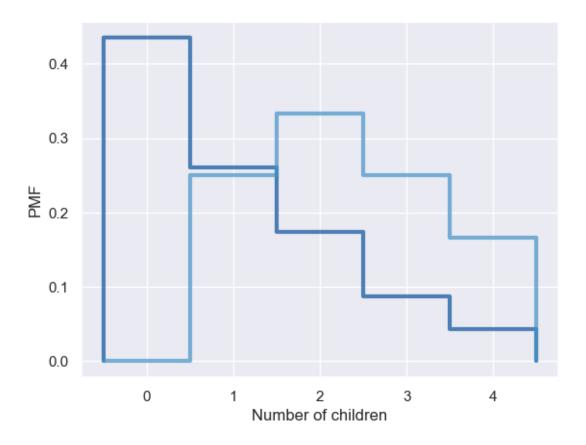
Actual mean: 1.0434782608695652 Biased mean: 2.8926174496644297



```
[196]: import thinkstats2
import thinkplot

# Adjusted values to better reflect household sizes.
```

```
d = \{0: 50, 1: 30, 2: 20, 3: 10, 4: 5\}
# Create the actual PMF.
pmf_actual = thinkstats2.Pmf(d, label='actual')
# Function to create a biased PMF by oversampling larger families.
def BiasPmf(pmf, label):
   new_pmf = pmf.Copy(label=label)
    for x, p in pmf.Items():
       new_pmf.Mult(x, x)
    new_pmf.Normalize()
   return new_pmf
# Creating the biased PMF.
biased_pmf = BiasPmf(pmf_actual, label='biased')
# Plot the PMFs.
thinkplot.PrePlot(2)
thinkplot.Pmfs([pmf_actual, biased_pmf])
thinkplot.Config(xlabel="Number of children", ylabel="PMF")
thinkplot.Show()
# Calculating Means.
actual_mean = pmf_actual.Mean()
biased_mean = biased_pmf.Mean()
print("Actual mean:", actual_mean)
print("Biased mean:", biased_mean)
```



0.0.3 Chapter 3

0.0.4 Page 36: Exercise 3-2.

```
[197]: # Creating PmfMean function to compute mean.
def PmfMean(pmf):
    mean = 0
    for x, p in pmf.Items():
        mean += x * p
    return mean

# Creating PmfVar function to compute variance.
def PmfVar(pmf):
    mean = PmfMean(pmf)
    variance = 0
    for x, p in pmf.Items():
        variance += p * (x - mean) ** 2
```

```
return variance
# Importing methods from thinkstats2 to compare results with example.
import thinkstats2
# Creating a list and an object for Pmf.
values = [1, 2, 2, 3, 5]
pmf = thinkstats2.Pmf(values)
# Calculating the mean and variance with created functions and assigning them_
⇔to variables.
mean = PmfMean(pmf)
variance = PmfVar(pmf)
# Printing both results.
print(f"Mean: {mean}")
print(f"Variance: {variance}")
# Checking results to confirm consistency with Pmf's build-in methods.
print(f"Mean (Pmf method): {pmf.Mean()}")
print(f"Variance (Pmf method): {pmf.Var()}")
```

Mean: 2.6
Variance: 1.84
Mean (Pmf method): 2.6
Variance (Pmf method): 1.84

0.0.5 Chapter 4

0.0.6 Page 47: Exercise 4-1.

```
pregnancy['birthwgt_oz'] = pregnancy['birthwgt_oz'].replace([97, 98, 99], np.
 ⇔nan)
pregnancy['hpagelb'] = pregnancy['hpagelb'].replace([97, 98, 99], np.nan)
pregnancy['babysex'] = pregnancy['babysex'].replace([7, 9], np.nan)
pregnancy['nbrnaliv'] = pregnancy['nbrnaliv'].replace([9], np.nan)
# Extracting birth weights
birth_weights = pregnancy['birthwgt_lb']
# Separating first babies and others.
first_babies = pregnancy[pregnancy['birthord'] == 1]
others = pregnancy[pregnancy['birthord'] != 1]
# Computing Cumulative Distribution Function(CDFs) for birth weights.
cdf_all = thinkstats2.Cdf(birth_weights, label='All Live Births')
cdf_first_babies = thinkstats2.Cdf(first_babies['birthwgt_lb'], label='First_u

→Babies')
cdf_others = thinkstats2.Cdf(others['birthwgt_lb'], label='Others')
# Creating method to compute percentile rank.
def PercentileRank(cdf, value):
   return cdf.PercentileRank(value)
# My birth weight.
my\_born\_weight = 4
# Check if you were a first baby
is_first_baby = input("Are you first baby? (yes/no): ").strip().lower()
if is first baby == 'yes':
   percentile_rank = PercentileRank(cdf_first_babies, my_born_weight)
else:
   percentile_rank = PercentileRank(cdf_others, my_born_weight)
print(f"My birth weight percentile rank: {percentile_rank:.2f}")
if percentile_rank >= 90:
   print("Apologize to your mother for being in the 90th percentile or higher!
```

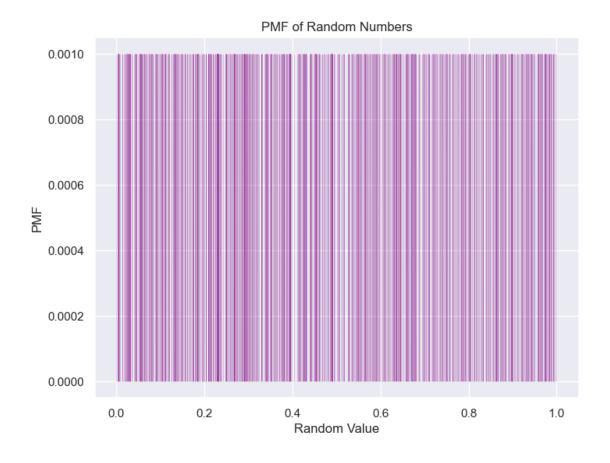
Are you first baby? (yes/no): No
My birth weight percentile rank: 4.34

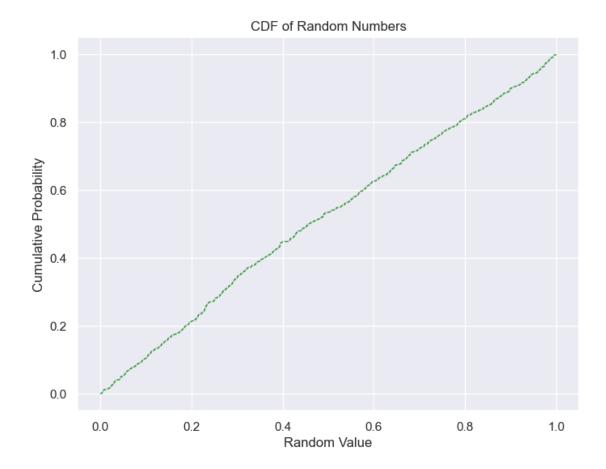
0.0.7 Chapter 4

0.0.8 Page 48: Exercise 4-2.

```
[200]: import numpy as np
      import thinkstats2
      import thinkplot
      import matplotlib.pyplot as plt
      # Generate 1,000 Random Numbers
      random_numbers = np.random.random(1000)
      # Plot PMF within a context manager
      with plt.ioff():
          pmf = thinkstats2.Pmf(random_numbers)
          thinkplot.Pmf(pmf, linewidth=0.1, color='purple')
          thinkplot.Config(xlabel='Random Value', ylabel='PMF', title='PMF of Random_
       thinkplot.Show()
      # Plot CDF within a context manager
      with plt.ioff():
          cdf = thinkstats2.Cdf(random_numbers)
          thinkplot.Cdf(cdf, linestyle='--', linewidth=1, color='green')
          thinkplot.Config(xlabel='Random Value', ylabel='Cumulative Probability', u

→title='CDF of Random Numbers', legend=False)
          thinkplot.Show()
```





The distribution is uniform. We can see from the PMF graph that all bars(lines) are the same height which means its uniform. Also, the line from CDF appears to be almost streight which indicates that the distribution is uniform.