Rodriguez_Felipe_DSC530_Exercise5.2

January 15, 2023

Exercise 5.1

```
[199]: # Copied from book to download scripts
       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)
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/
        ⇔thinkstats2.py")
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/thinkplot.
        ⇔py")
[200]: import scipy.stats
       import thinkstats2
       import thinkplot
[201]: # Set up values given
       men_mu = 178
       men_sigma = 7.7
       # Creates data set
       dist = scipy.stats.norm(loc=mu, scale=sigma)
[202]: # Min Length converted to cm
       length_feet_min = 5
       length_inch_min = 10
       total_cm_min = (length_feet_min * 30.48 ) + (length_inch_min * 2.54)
       total_cm_min
```

[202]: 177.8

```
[203]: # Max length converted to cm
       length_feet_max = 6
       length_inch_max = 1
       total_cm_max = (length_feet_max * 30.48 ) + (length_inch_max * 2.54)
       total_cm_max
[203]: 185.42
[204]: # Calculates amount of people in the height range
       min_amount = dist.cdf(total_cm_min) # Lowest height 5'10" amount
       max_amount = dist.cdf(total_cm_max) # Highest height 6'1" amount
       # Calculates percentage of people between height range
       print('''Percentage of people that are between 5'10" and 6'1":''', ((max_amount ⊔
        \rightarrow min_amount)*100))
      Percentage of people that are between 5'10" and 6'1": 34.274683763147365
      Exercise 5.2
[205]: import scipy.stats
[206]: # Set up values given
       alpha = 1.7
       xmin = 1 # meter
       # Creates Data
       human_height = scipy.stats.pareto(b=alpha, scale=xmin)
[207]: # Calculates Mean
       human_height.mean()
[207]: 2.428571428571429
[208]: # Calculation for people taller 1 km out of 7 billion
       (1 - human height.cdf(1000)) * 7e9
[208]: 55602.976430479954
[209]: human_height.ppf(1 - 1 / 7e9)
[209]: 618349.6106759505
[210]: human_height.sf(600000) * 7e9
[210]: 1.0525455861201714
```

Exercise 6.1

```
[211]: # Copied from book to download scripts
       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)
[212]: download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/hinc.py")
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/hinc06.
       download("https://github.com/AllenDowney/ThinkStats2/raw/master/code/
        ⇔thinkstats2.py")
[213]: import hinc
       import numpy as np
[214]: # Reads Data
       income_df = hinc.ReadData()
[215]: # Copied from book to use function
       def InterpolateSample(df, log_upper=6.0):
           """Makes a sample of log10 household income.
           Assumes that log10 income is uniform in each range.
           df: DataFrame with columns income and freq
           log_upper: log10 of the assumed upper bound for the highest range
           returns: NumPy array of log10 household income
           # compute the log10 of the upper bound for each range
           df['log_upper'] = np.log10(df.income)
           # get the lower bounds by shifting the upper bound and filling in
           # the first element
           df['log_lower'] = df.log_upper.shift(1)
           df.loc[0, 'log_lower'] = 3.0
           # plug in a value for the unknown upper bound of the highest range
           df.loc[41, 'log_upper'] = log_upper
           # use the freq column to generate the right number of values in
```

```
# each range
           arrays = []
           for _, row in df.iterrows():
               vals = np.linspace(row.log_lower, row.log_upper, int(row.freq))
               arrays.append(vals)
           # collect the arrays into a single sample
           log_sample = np.concatenate(arrays)
           return log_sample
[216]: # Converts data
       sample_data_log = InterpolateSample(income_df)
[217]: data = np.power(10, sample_data_log)
[218]: # Calculates Mean
       means = thinkstats2.Mean(data)
       means
[218]: 74278.7075311872
[219]: # Calculates Median
       thinkstats2.Median(data)
[219]: 51226.45447894046
[220]: # Calculates Skewness
       thinkstats2.Skewness(data)
[220]: 4.949920244429583
[221]: # Calculates Pearson Median Skewness
       thinkstats2.PearsonMedianSkewness(data)
[221]: 0.7361258019141782
[222]: # Creates CDF of Data
       cdf = thinkstats2.Cdf(data)
[223]: # Calculatation of what people make below mean, close to 66%
       cdf.Prob(means)
[223]: 0.660005879566872
```

How do the results depend on the assumed upper bound? The Mean, Skewness, Pearson Median Skewness, and people who make below the mean all change because the upper bound is increased. The data set shifts to the right. The value that remains consistent is the median of the data, which does not change based on upper bound.