Week12_Final_KanaparthiVenkata

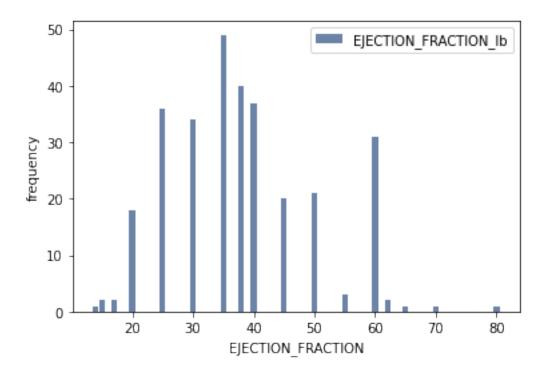
March 3, 2022

```
[47]: from __future__ import print_function, division
    import thinkstats2
    import sys
    import numpy as np
    import pandas
    import random

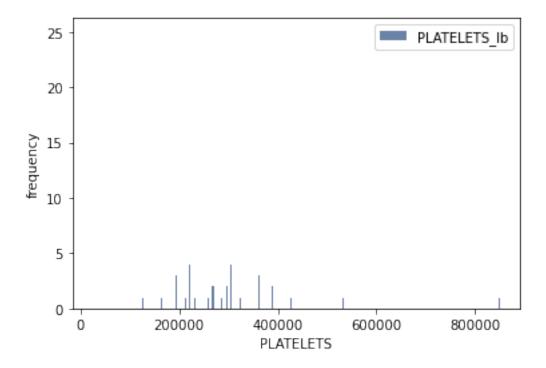
[4]: def ReadClinicalData():
        df = pandas.read_csv('heart_failure_clinical_records_dataset.csv')
        return df

[12]: clincDat = ReadClinicalData()
```

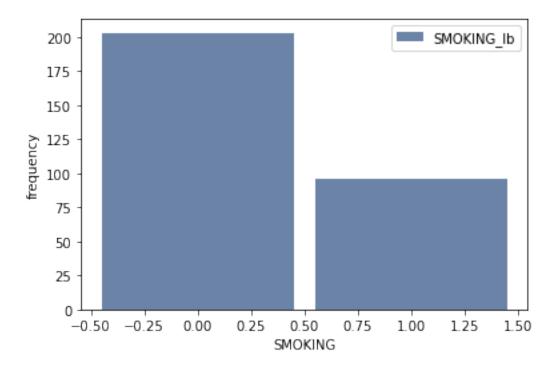
3.) Include a histogram of each of the 5 variables – in your summary and analysis, identify any outliers and explain the reasoning for them being outliers and how you believe they should be handled (Chapter 2).



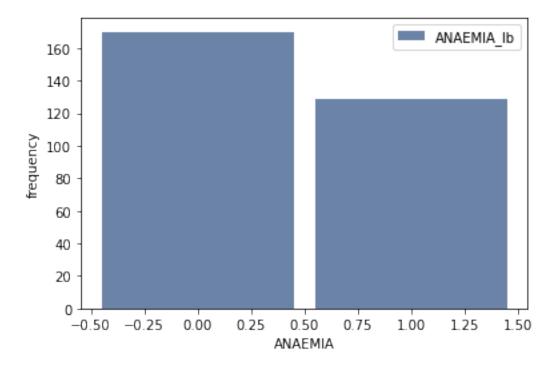
```
[14]: hist = thinkstats2.Hist(clincDat.PLATELETS, label='PLATELETS_lb')
    thinkplot.Hist(hist)
    thinkplot.Show(xlabel='PLATELETS', ylabel='frequency')
```

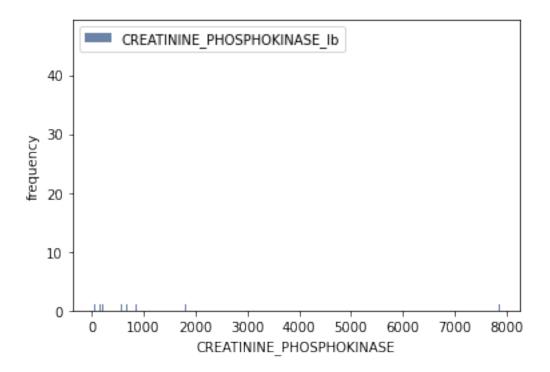


```
[15]: hist = thinkstats2.Hist(clincDat.SMOKING, label='SMOKING_lb')
thinkplot.Hist(hist)
thinkplot.Show(xlabel='SMOKING', ylabel='frequency')
```



```
[17]: hist = thinkstats2.Hist(clincDat.ANAEMIA, label='ANAEMIA_lb')
    thinkplot.Hist(hist)
    thinkplot.Show(xlabel='ANAEMIA', ylabel='frequency')
```





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4.) Include the other descriptive characteristics about the variables: Mean, Mode, Spread, and Tails (Chapter 2).

```
[20]: mean_ejection_fraction = clincDat.EJECTION_FRACTION.mean()
   var_ejection_fraction = clincDat.EJECTION_FRACTION.var()
   std_ejection_fraction = clincDat.EJECTION_FRACTION.std()
   mean_ejection_fraction, var_ejection_fraction, std_ejection_fraction
```

[20]: (38.08361204013378, 140.06345536576063, 11.834840741039173)

```
[21]: mean_platelets = clincDat.PLATELETS.mean()
   var_platelets = clincDat.PLATELETS.var()
   std_platelets = clincDat.PLATELETS.std()
   mean_platelets, var_platelets, std_platelets
```

[21]: (263358.02926421416, 9565668749.44888, 97804.23686859828)

```
[22]: mean_smoking = clincDat.SMOKING.mean()
  var_smoking = clincDat.SMOKING.var()
  std_smoking = clincDat.SMOKING.std()
  mean_smoking, var_smoking, std_smoking
```

[22]: (0.3210702341137124, 0.21871562927880447, 0.4676704280567721)

```
[23]: mean_anaemia = clincDat.ANAEMIA.mean()
   var_anaemia = clincDat.ANAEMIA.var()
   std_anaemia = clincDat.ANAEMIA.std()
   mean_anaemia, var_anaemia, std_anaemia
```

[23]: (0.431438127090301, 0.2461224214944669, 0.49610726813307915)

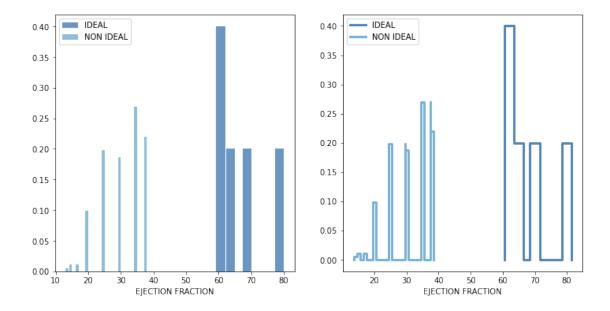
```
[24]: mean_creatinine_phosphokinase = clincDat.CREATININE_PHOSPHOKINASE.mean()
var_creatinine_phosphokinase = clincDat.CREATININE_PHOSPHOKINASE.var()
std_creatinine_phosphokinase = clincDat.CREATININE_PHOSPHOKINASE.std()
mean_creatinine_phosphokinase, var_creatinine_phosphokinase, u
→std_creatinine_phosphokinase
```

- [24]: (581.8394648829432, 941458.5714574311, 970.2878807124363)
 - 5.) Using pg. 29 of your text as an example, compare two scenarios in your data using a PMF. Reminder, this isn't comparing two variables against each other it is the same variable, but a different scenario. Almost like a filter. The example in the book is first babies compared to all other babies, it is still the same variable, but breaking the data out based on criteria we are exploring (Chapter 3).

```
[32]: ideal = clincDat[clincDat.EJECTION_FRACTION > 60]
    notIdeal = clincDat[clincDat.EJECTION_FRACTION < 40]

ideal_pmf = thinkstats2.Pmf(ideal.EJECTION_FRACTION, label='IDEAL')
    notIdeal_pmf = thinkstats2.Pmf(notIdeal.EJECTION_FRACTION, label='NON IDEAL')

thinkplot.PrePlot(2, cols=2)
    thinkplot.Hist(ideal_pmf, align='right')
    thinkplot.Hist(notIdeal_pmf, align='right')
    thinkplot.Config(xlabel = 'EJECTION FRACTION')
    thinkplot.PrePlot(2)
    thinkplot.SubPlot(2)
    thinkplot.Pmfs([ideal_pmf, notIdeal_pmf])
    thinkplot.Show(xlabel='EJECTION FRACTION')</pre>
```



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6.) Create 1 CDF with one of your variables, using page 41-44 as your guide, what does this tell you about your variable and how does it address the question you are trying to answer (Chapter 4).

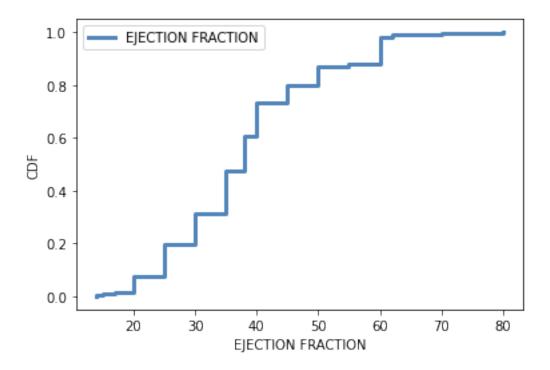
```
[37]: ejectFrac_cdf = thinkstats2.Cdf(clincDat.EJECTION_FRACTION, label='EJECTION_L

→FRACTION')

thinkplot.PrePlot(2)

thinkplot.Cdf(ejectFrac_cdf)

thinkplot.Show(xlabel='EJECTION FRACTION', ylabel='CDF')
```



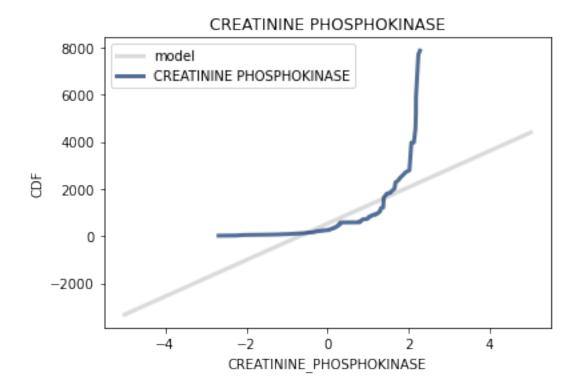
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7.) Plot 1 analytical distribution and provide your analysis on how it applies to the dataset you have chosen (Chapter 5).

```
def MakeNormalPlot(creatPhos):
    mean, var = thinkstats2.TrimmedMeanVar(creatPhos, p=0.01)
    std = np.sqrt(var)

    xs = [-5, 5]
    xs, ys = thinkstats2.FitLine(xs, mean, std)
    thinkplot.Plot(xs, ys, color='0.8', label='model')

    xs, ys = thinkstats2.NormalProbability(creatPhos)
    thinkplot.Plot(xs, ys, label='CREATININE PHOSPHOKINASE')
```



8.) Create two scatter plots comparing two variables and provide your analysis on correlation and causation. Remember, covariance, Pearson's correlation, and Non-Linear Relationships should also be considered during your analysis (Chapter 7).

```
[41]: def Cov(xs, ys, meanx=None, meany=None):
    xs = np.asarray(xs)
    ys = np.asarray(ys)

if meanx is None:
    meanx = np.mean(xs)
    if meany is None:
        meany = np.mean(ys)

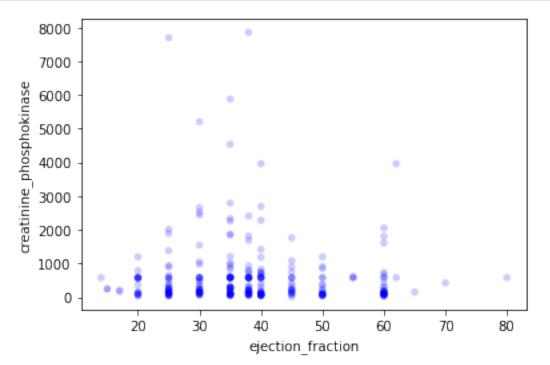
cov = np.dot(xs-meanx, ys-meany) / len(xs)
    return cov
```

```
[44]: def Corr(xs, ys):
    xs = np.asarray(xs)
    ys = np.asarray(ys)

meanx, varx = thinkstats2.MeanVar(xs)
    meany, vary = thinkstats2.MeanVar(ys)

corr = Cov(xs, ys, meanx, meany) / np.sqrt(varx * vary)
```

return corr



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```
[42]: Cov(ejection_fraction, creatinine_phosphokinase)
```

[42]: -504.4815606089415

```
[45]: Corr(ejection_fraction, creatinine_phosphokinase)
```

[45]: -0.044079554456711474

9.) Conduct a test on your hypothesis using one of the methods covered in Chapter 9.

```
[48]: class HypothesisTest(object):
    def __init__(self, data):
        self.data = data
        self.MakeModel()
```

```
[60]: class CorrelationPermute(HypothesisTest):

    def TestStatistic(self, data):
        xs, ys = data
        test_stat = abs(thinkstats2.Corr(xs, ys))
        return test_stat

    def RunModel(self):
        xs, ys = self.data
        xs = np.random.permutation(xs)
        return xs, ys
```

```
[61]: data = clincDat.CREATININE_PHOSPHOKINASE, clincDat.DEATH_EVENT
ht = CorrelationPermute(data)
pvalue = ht.PValue()
pvalue
```

[61]: 0.275

10.) For this project, conduct a regression analysis on either one dependent and one explanatory variable, or multiple explanatory variables (Chapter 10 & 11).

```
[68]: import statsmodels.formula.api as smf
formula = 'DEATH_EVENT ~ EJECTION_FRACTION'
model = smf.ols(formula, data=clincDat)
results = model.fit()
print(results.summary())
```

OLS Regression Results

Dep. Variable: Model: Method: Date: Time: No. Observations: Df Residuals: Df Model: Covariance Type:	Least Sun, 06 J	OLS Squares Tun 2021	R-squared: Adj. R-squar F-statistic: Prob (F-stat Log-Likeliho AIC: BIC:	istic):	0.072 0.069 23.09 2.45e-06 -185.33 374.7 382.1
0.975]	coef	std err	t	P> t	[0.025
Intercept 0.899 EJECTION_FRACTION -0.006	0.7253	0.088	8.235 -4.806	0.000	0.552 -0.015
Omnibus: Prob(Omnibus): Skew: Kurtosis:		0.000	Durbin-Watso Jarque-Bera Prob(JB): Cond. No.		1.114 43.084 4.41e-10 135.

Notes:

^[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.