Final test - Salla Vesterinen

Probability, Statistics and Discrete Mathematics, Spring 2019 9.5.2019

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```
In [1]: # Import necessary Libraries
    %pylab inline
    import numpy.random as rnd
    import scipy.stats as stats
    import pandas as pd
    from pandas import Series
    from matplotlib import pyplot
    from scipy.stats import norm
```

Populating the interactive namespace from numpy and matplotlib

Problem 1

Here is how I solved this problem...

```
In [2]: #Value array
speed=np.array([199.49, 199.33, 199.23, 200.01, 200.15, 199.67, 200.25, 199.51, 200.6, 199.94])

#Average = mean
mn=speed.mean()
print("Mean:",mn)

#Variance = standard deviation
sd=speed.std()
print("Standard deviation:",sd)

#95% falls in between two standard deviations of the mean
l=mn-2*sd
print("Low:",l)
h=mn+2*sd
print("High:",h)
```

Mean: 199.8179999999998

Standard deviation: 0.42051872728809525

Low: 198.9769625454238 High: 200.65903745457618

Final answers:

- (a) 199.818
- (b) 0.421
- (c) 198.877 to 200.659

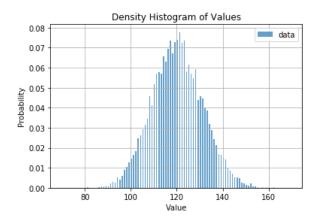
Problem 2

Here is how I solved this problem...

```
In [3]: #Array of random values following Poisson distribution and with mean of 120
x = np.random.poisson(size=10000, lam=120)
#Check mean
print(x.mean())
119.9399
```

```
In [4]:
        #Density histogram
        hist(x, bins = arange(70, 170, 0.5), alpha = 0.7, density = True, label = "data")
        xlabel('Value')
        ylabel('Probability')
        plt.title('Density Histogram of Values')
        legend()
        grid()
        #Probability value<100
        print("Less than 100:",(x<100).sum())</pre>
        #Probability 120<value<140
        print("Between 120 and 140:",(x<120).sum()+(x>140).sum())
        #Quantile 25% and 75% limits
        low=np.quantile(x, 0.25)
        print("25% limit:",low)
        high=np.quantile(x, 0.75)
        print("75% limit:",high)
        #Verification that 50% falls between range
        print("50%:",10000*0.5)
        res=(x<low).sum()+(x>high).sum()
        print("Count:",res)
```

Less than 100: 295
Between 120 and 140: 5219
25% limit: 112.0
75% limit: 127.0
50%: 5000.0
Count: 4721



Final answers:

- (a) Histogram above, "Density Histogram of Values"
- (b) Between 268 and 302 (mean of 286.6), when repeated 5 times
- (c) Between 4702 and 4864 (mean of 4799.6), when repeated 5 times
- (d) 25% quantile limit: 112 or 113 (depending on run), 75% quantile limit: 127 or 128

Problem 3

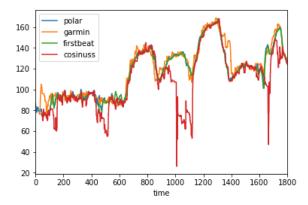
Here is how I solved this problem...

```
In [5]: # Read the example data into Python
    file=open("heartrate.csv", "r")
    sep = ","
    data = pd.read_csv(file, sep, index_col='time')
    data.head()
```

Out[5]:

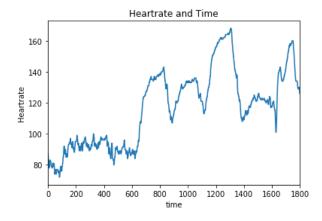
	polar	garmin	firstbeat	cosinuss
time				
0	87.0	NaN	NaN	NaN
1	85.0	NaN	NaN	NaN
2	84.0	NaN	NaN	NaN
3	83.0	NaN	NaN	NaN
4	83.0	NaN	NaN	NaN

```
In [6]: #Graphical plot of heartrates against time (time on x-axis)
    series = pd.read_csv('heartrate.csv', header=0, index_col='time')
    series.plot()
    pyplot.show()
```



```
In [7]: #Another way of doing it
    series = Series.from_csv('heartrate.csv', header=0)
    ylabel('Heartrate')
    plt.title('Heartrate and Time')
    series.plot()
    pyplot.show()
```

C:\Users\Salla\Anaconda3\lib\site-packages\pandas\core\series.py:4141: FutureWarning: from_csv is deprecated.
Please use read_csv(...) instead. Note that some of the default arguments are different, so please refer to t
he documentation for from_csv when changing your function calls
 infer_datetime_format=infer_datetime_format)



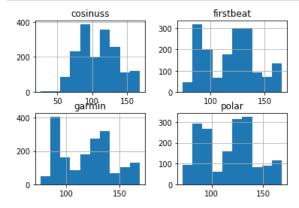
Final answers (based only on graph!):

(a) Typical range: 80-160(b) Min: 75, Max: 165(c) Average: 120

Problem 4

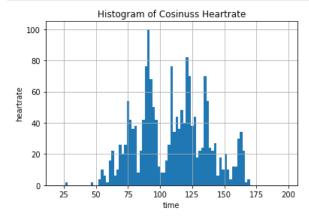
Here is how I solved this problem...

```
In [8]: #Histogram plots of heartrates and calculate descriptive statistics of the results
#Overview of all
series = pd.read_csv('heartrate.csv', header=0, index_col='time')
series.hist()
pyplot.show()
```

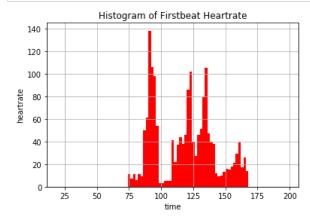


```
In [9]: #Data
    series = pd.read_csv('heartrate.csv', header=0, index_col='time')
    a=data['cosinuss']
    b=data['firstbeat']
    c=data['garmin']
    d=data['polar']
    bins=np.arange(20,200,2)
```

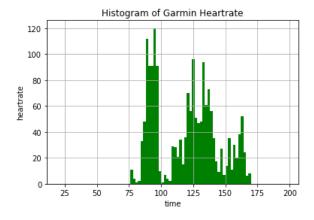
```
In [10]: #Cosinuss
    plt.hist(a, bins)
    xlabel('time')
    ylabel('heartrate')
    plt.title('Histogram of Cosinuss Heartrate')
    grid()
```



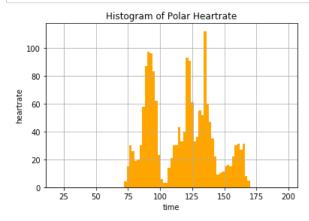
```
In [11]: #Firstbeat
    plt.hist(b, bins, color='red')
    xlabel('time')
    ylabel('heartrate')
    plt.title('Histogram of Firstbeat Heartrate')
    grid()
```



```
In [12]: #Garmin
    plt.hist(c, bins, color='green')
    xlabel('time')
    ylabel('heartrate')
    plt.title('Histogram of Garmin Heartrate')
    grid()
```



```
In [13]: #Polar
    plt.hist(d, bins, color='orange')
    xlabel('time')
    ylabel('heartrate')
    plt.title('Histogram of Polar Heartrate')
    grid()
```



```
In [14]: #Data
    print("Cosinuss:",a.mean(), a.std())
    print("Firstbeat:",b.mean(), b.std())
    print("Garmin:",c.mean(), c.std())
    print("Polar:",d.mean(), d.std())
```

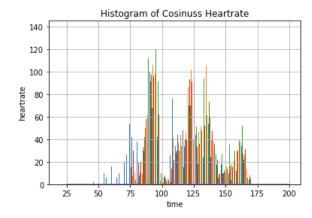
Cosinuss: 109.07797381900967 27.21012387789403 Firstbeat: 118.09766081871345 23.337284370811943 Garmin: 119.2697072072 24.159724490637693 Polar: 116.6627777777778 24.42094602676549

```
In [15]: #Probability distribution that fits the histograms
    plt.hist([a, b, c, d], bins)
    xlabel('time')
    ylabel('heartrate')
    plt.title('Histogram of Cosinuss Heartrate')
    grid()

# Create a normal distribution with default values loc = 1.0, and scale = 1.0

rv = norm()
    # Create x-axis
    x = arange(25, 200, 0.1)
# Draw the probability density function (pdf)
    plot(x, rv.pdf(x))
```

Out[15]: [<matplotlib.lines.Line2D at 0x2375db4b898>]



Final answers:

- (a) Number of samples: 5
- (b) Device with highest average heartrate: Garmin
- (c) Device with largest deviation: Cosinuss this is because Cosinuss started when the time started (lower heartrate) unlike the others
- (d) Probability distribution that fits: From afar it looks like a normal distribution with a slope up followed by the peak and then the descent down. In reality though it goes down and up again after the initial peak therefore looking like a multimodal distribution function with three peaks.