

# Lab13

October 15, 2020

## 1 Laboratory 13 Probability Modeling

1.1 Full name:

1.2 R#:

1.3 HEX:

1.4 Title of the notebook

1.5 Date:

1.5.1 Important Terminology:

**Population:** In statistics, a population is the entire pool from which a statistical sample is drawn. A population may refer to an entire group of people, objects, events, hospital visits, or measurements. **Sample:** In statistics and quantitative research methodology, a sample is a set of individuals or objects collected or selected from a statistical population by a defined procedure. The elements of a sample are known as sample points, sampling units or observations. **Distribution (Data Model):** A data distribution is a function or a listing which shows all the possible values (or intervals) of the data. It also (and this is important) tells you how often each value occurs.

*From <https://www.investopedia.com/terms> <https://www.statisticshowto.com/data-distribution/>*

```
[0]: ### Important Steps:
1. __Get descriptive statistics- mean, variance, std. dev.__
2. __Use plotting position formulas (e.g., weibull, gringorten, cunnane) and__
   ↳plot the SAMPLES (data you already have)__
3. __Use different data models (e.g., normal, log-normal, Gumbell) and find the__
   ↳one that better FITs your samples- Visual or Numerical__
4. __Use the data model that provides the best fit to infer about the__
   ↳POPULATION__
```

## 2 Estimate the magnitude of the annual peak flow at Spring Ck near Spring, TX.

The file 08068500.pkf is an actual WATSTORE formatted file for a USGS gage at Spring Creek, Texas. The first few lines of the file look like:

Z08068500

USGS

H08068500	3006370952610004848339SW12040102409	409	72.6
N08068500	Spring Ck nr Spring, TX		
Y08068500			
308068500	19290530 483007	34.30	1879
308068500	19390603 838	13.75	
308068500	19400612 3420	21.42	
308068500	19401125 42700	33.60	
308068500	19420409 14200	27.78	
308068500	19430730 8000	25.09	
308068500	19440319 5260	23.15	
308068500	19450830 31100	32.79	
308068500	19460521 12200	27.97	

The first column are some agency codes that identify the station , the second column after the fourth row is a date in YYYYMMDD format, the third column is a discharge in CFS, the fourth and fifth column are not relevant for this laboratory exercise. The file was downloaded from

[https://nwis.waterdata.usgs.gov/tx/nwis/peak?site\\_no=08068500&agency\\_cd=USGS&format=hn2](https://nwis.waterdata.usgs.gov/tx/nwis/peak?site_no=08068500&agency_cd=USGS&format=hn2)

In the original file there are a couple of codes that are manually removed:

- 19290530 483007; the trailing 7 is a code identifying a break in the series (non-sequential)
- 20170828 784009; the trailing 9 identifies the historical peak

The laboratory task is to fit the data models to this data, decide the best model from visual perspective, and report from that data model the magnitudes of peak flow associated with the probabilities below (i.e. populate the table)

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value (in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

The first step is to read the file, skipping the first part, then build a dataframe:

```
[6]: # Read the data file
amatrix = [] # null list to store matrix reads
```

```

rowNumA = 0
matrix1=[]
col0=[]
col1=[]
col2=[]
with open('08068500.pkf','r') as afile:
    lines_after_4 = afile.readlines()[4:]
afile.close() # Disconnect the file
howmanyrows = len(lines_after_4)
for i in range(howmanyrows):
    matrix1.append(lines_after_4[i].strip().split())
for i in range(howmanyrows):
    col0.append(matrix1[i][0])
    col1.append(matrix1[i][1])
    col2.append(matrix1[i][2])
# col2 is date, col3 is peak flow
#now build a datafranem

```

```

[4]: import pandas
df = pandas.DataFrame(col0)
df['date']= col1
df['flow']= col2

```

```

[5]: df.head()

```

```

[5]:
      0      date  flow
0  308068500  19290530  48300
1  308068500  19390603    838
2  308068500  19400612   3420
3  308068500  19401125  42700
4  308068500  19420409  14200

```



Now explore if you can plot the dataframe as a plot of peaks versus date.

```

[7]: # Plot here

```

From here on you can proceed using the lecture notebook as a go-by, although you should use functions as much as practical to keep your work concise

```

[87]: # Descriptive Statistics

```

```

[88]: # Weibull Plotting Position Function

```

```

[89]: # Normal Quantile Function

```

```

[90]: # Fitting Data to Normal Data Model

```



## 2.1 Normal Distribution Data Model

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value (in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

[91]: *# Log-Normal Quantile Function*

[92]: *# Fitting Data to Normal Data Model*



## 2.2 Log-Normal Distribution Data Model

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value (in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

[93]: *# Gumbell EV1 Quantile Function*

[94]: # Fitting Data to Gumbell EV1 Data Model



## 2.3 Gumbell Double Exponential (EV1) Distribution Data Model

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value (in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

[95]: # Gamma (Pearson Type III) Quantile Function



[96]: # Fitting Data to Pearson (Gamma) III Data Model

# This is new, in lecture the fit was to log-Pearson, same procedure, but not  
→ log transformed



## 2.4 Pearson III Distribution Data Model

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value (in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)

Exceedence Probability	Flow Value	Remarks
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

[97]: # Fitting Data to Log-Pearson (Log-Gamma) III Data Model



## 2.5 Log-Pearson III Distribution Data Model

Exceedence Probability	Flow Value	Remarks
25%	????	75% chance of greater value
50%	????	50% chance of greater value
75%	????	25% chance of greater value
90%	????	10% chance of greater value
99%	????	1% chance of greater value
		(in flood statistics, this is the 1 in 100-yr chance event)
99.8%	????	0.002% chance of greater value (in flood statistics, this is the 1 in 500-yr chance event)
99.9%	????	0.001% chance of greater value (in flood statistics, this is the 1 in 1000-yr chance event)

## 3 Summary of “Best” Data Model based on Graphical Fit

