

Annual Precipitation Analysis and How it Might be used to Quantify the Effect on Calhoun Brake

The Table below holds Annual Precipitation data from 1892 to 2009. Data from 1892 to 2009 were used in this analysis.

The values in each monthly cell are inches*100, the values in Annual and Summer cells are in inches.

out1
out2
out3
out4
out5

	MAR	APR	MAY	JUN	JUL	AUG	SEP	OCT	NOV	DEC	Annual	Summer			
	643	1006	260	537	492	438	501	150	573	923	65.5	19.68			
	234	722	689	782	213	328	28	31	478	254	43.21	13.51			
	822	563	335	170	660	325	203	397	132	617	55.25	13.58			
	407	186	494	1315	355	469	107	165	423	299	48.35	22.46			
	464	247	153	78	28	307	371	516	561	33	37.26	7.84			
	622	200	270	393	332	247	162	249	112	917	45.54	11.34			
	403	462	137	861	381	226	462	528	718	191	53.59	19.3			
	485	169	136	194	158	222	49	248	245	510	30.74	6.23			
	433	1189	341	803	848	147	390	243	325	197	57.67	21.88			

$i := 1 .. 118$

$year_i := out1_i$

Output the annual precipitation for each year: $A_P_i := out2_i \cdot in$

Output January precipitation for each year : $J_P_i := \frac{out3_i}{100} \cdot in$

Output June precipitation for each year: $Ju_P_i := \frac{out4_i}{100} \cdot in$

Output summer precipitation for each year: $summer_P_i := out5_i \cdot in$

$mean(A_P) = 51.036 \cdot in$ $median(A_P) = 49.78 \cdot in$

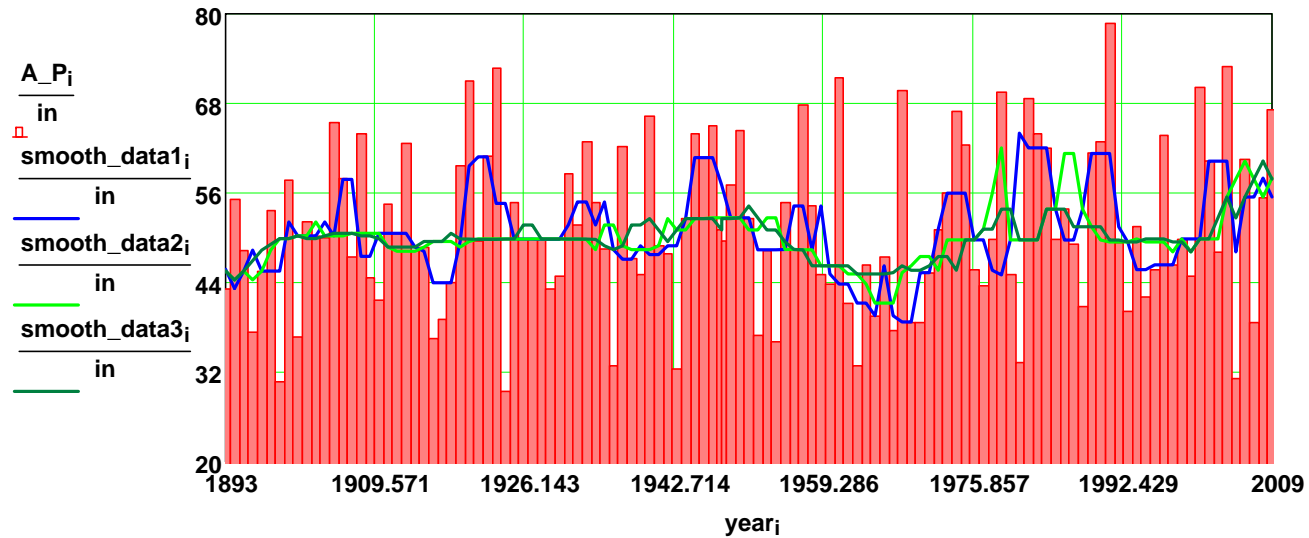
Note that the mean and median of the data set are very close together suggesting that the data may exhibit a symmetrical distribution about the mean.

Below I have plotted the data as solid bars and I have used a median smoothing algorithm with windows of 5, 10 and 15 years. The median smoothing algorithm replaces each point with the median of the n points centered on that value. The window width, n, is contracted near the ends of the vector.

Note that when a window of 31 is used the plot is nearly horizontal. This suggests no trends in the data extending over 30 years.

```
smooth_data1 := medsmooth(A_P, 5)
smooth_data2 := medsmooth(A_P, 11)
smooth_data3 := medsmooth(A_P, 15)
```

Annual Precipitation in inches, Monroe, La. 1892-2008



- ▣ Bar chart Annual Precipitation
- Median Smoothing 5 years
- Median Smoothing 10 years
- Median Smoothing 15 years

The statements below are used to create the probability plots for Monroe Rainfall, shown below

$$\text{mean}(A_P) = 51.036 \cdot \text{in}$$

$\text{sorted_precip_data} := \text{sort}(A_P)$

$\text{sorted_summer_data} := \text{sort}(\text{summer_P})$

$x := \text{sorted_precip_data}$

$xx := \text{sorted_summer_data}$

```

Rankdemo(x) := | sorted ← sort(x)
                | for i ∈ 0 .. length(x) - 1
                |   | m ← match(xi, sorted) + 1
                |   | ranki ← mean(m)
                | rank

```

precip_ranks := Rankdemo(x)

```

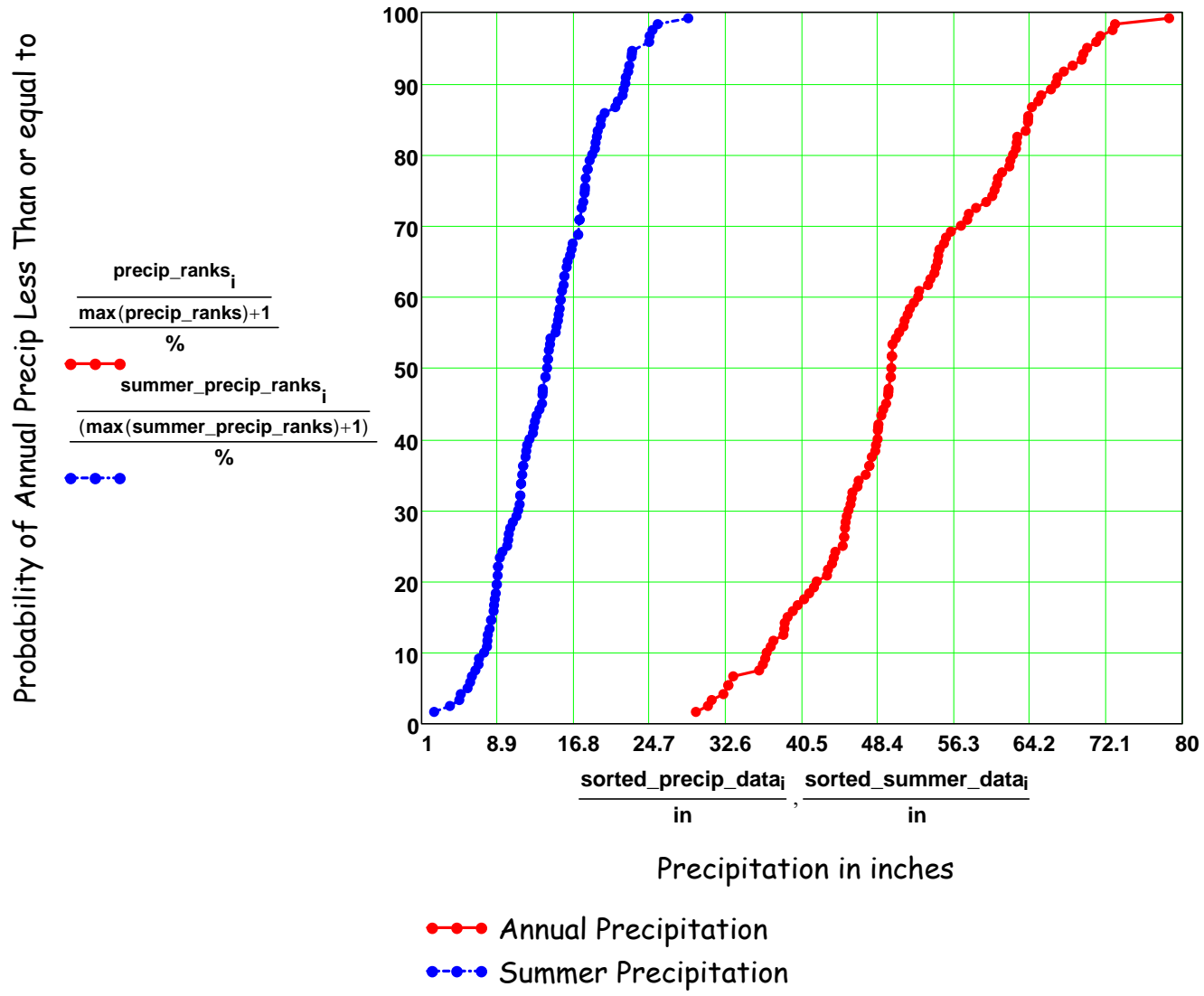
Rankdemo(xx) := | sorted ← sort(xx)
                 | for i ∈ 0 .. length(xx) - 1
                 |   | m ← match(xi, sorted) + 1
                 |   | ranki ← mean(m)
                 | rank

```

summer_precip_ranks := Rankdemo(xx)

Probability plots for annual precipitation data as well as summertime precipitation data are shown on the next page

Probability Plot - Rainfall - Monroe 1892-2009



These curves give the relationship between precipitation and the probability of that or a lesser amount falling in a subsequent year. For example there is a 50 % chance of an annual precipitation amount of 49.8 inches OR LESS falling in any subsequent year. There is a 27% (about 1 in 4) chance of getting a summer rain total greater than the 2010 summer rain total of 17.46 inches (curve on the left)

The median summer precipitation amount is 14.03 in. The summer 2010 amount was 17.41 in.

How to use the probability plot above to assess the effect on the break

The mean annual precipitation based on the data used here is 50.9 inches. Thus, there is a 50% chance that in any subsequent year either more or less than this amount will occur. Lets assume that in any year where less than the mean annual precipitation occurs there is opportunity for Cypress germination. Now, the annual plant cooling water discharge is equivalent to 6 inches over the break. So, we can construct an equivalent scenario in terms of total depth on the break if, in a year when $50.9 \text{ in} - 6 \text{ in} = 44.9 \text{ in}$ falls, the probability of this amount of precipitation or less occurring is approximately 26%. Thus, given the assumptions made, the plant's discharge causes the break to dry up 1 out of every 4 years instead of 1 in every 2 years.

Histogram of Annual Precipitation Amounts

The purpose of this analysis is to see if the annual precipitation values appear to follow any known statistical distribution

The algorithm below constructs a 20 bin histogram of the annual precipitation data

$$\text{Annual_Precip_histogram} := \text{histogram}\left(20, \frac{\mathbf{A_P}}{\text{in}}\right) =$$

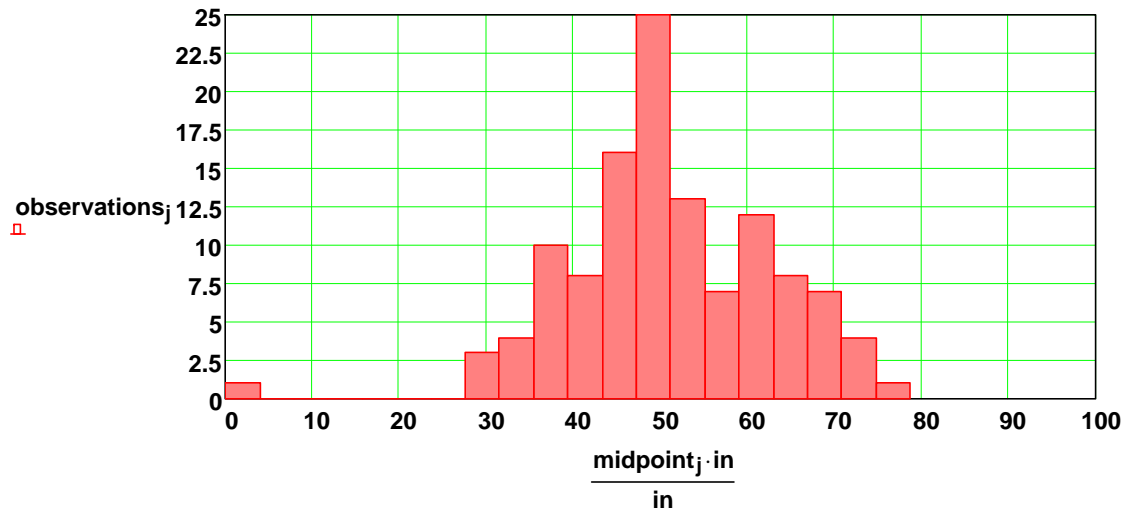
	0	1
0	1.966	1
1	5.898	0
2	9.83	0
3	13.762	0
4	17.694	0
5	21.626	0
6	25.558	0
7	29.49	3
8	33.422	4
9	37.354	10
10	41.286	8
11	45.218	16
12	49.15	25
13	53.082	13
14	57.014	7
15	60.946	...

$j := 0..19$

$\text{observations}_j := \text{Annual_Precip_histogram}_j,1$

$\text{midpoint}_j := \text{Annual_Precip_histogram}_j,0$

Histogram of Annual Precipitation Values



The histogram looks reasonably symmetrical about the mean suggesting that annual precipitation can be described by a normal distribution

Next, I examine the data to see if it is "autocorrelated", that is "does the precipitation amount in a given year imply anything about values to be expected in future years."

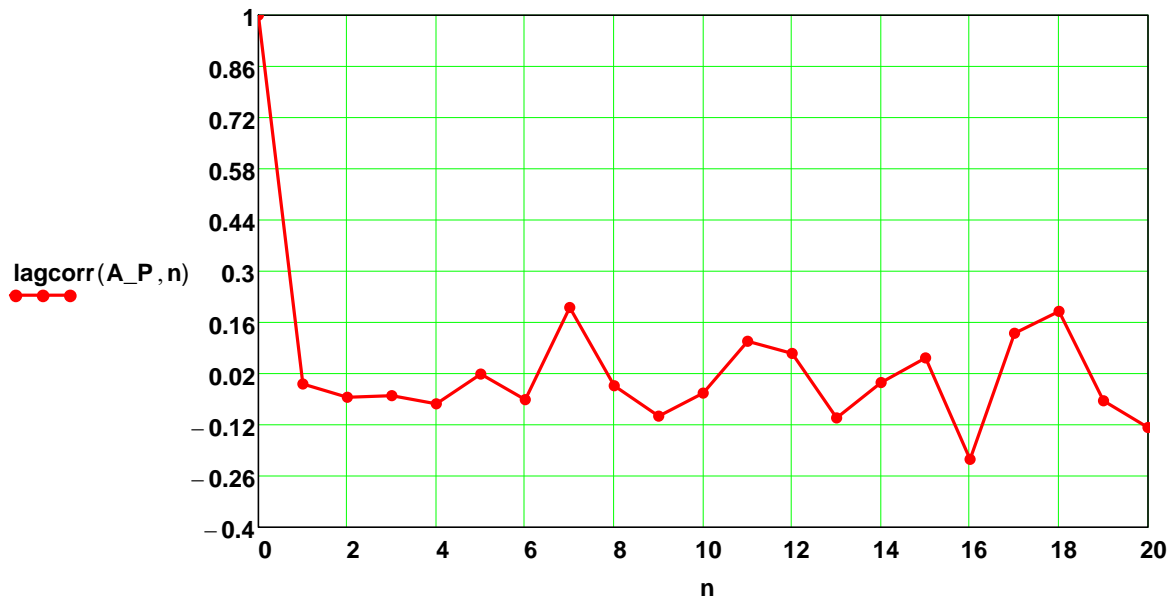
Autocorrelation Analysis

An autocorrelation analysis tests for correlation between successive values of annual precipitation in time, that is does the fact that you have a low value for precip in one year imply that you will have a low (or high) value the next year.

n := 0..20

$$\text{lagcorr}(A_P, n) := \frac{1}{\text{length}(A_P) \cdot \text{stdev}(A_P)^2} \cdot \sum_{i=n}^{\text{last}(A_P)} [(A_P_i - \text{mean}(A_P)) \cdot (A_P_{i-n} - \text{mean}(A_P))]$$

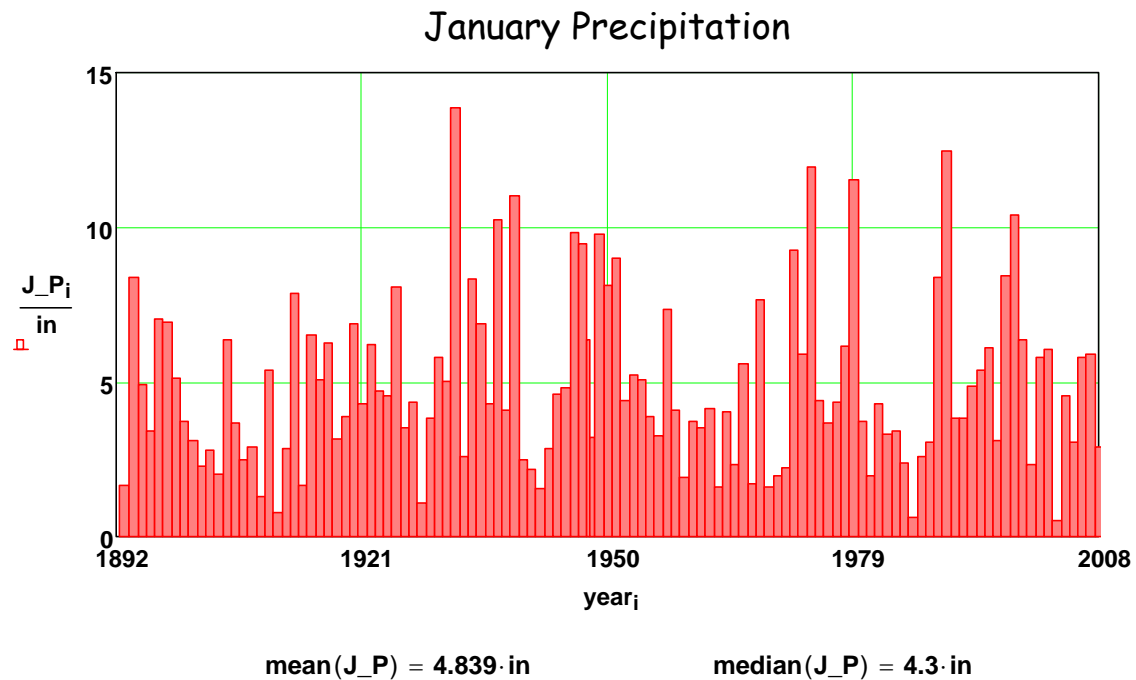
Correlogram - Annual Precipitation Calhoun Break



The analysis above suggests that annual precipitation values are independent in time over time frames from one to twenty years. Simply stated "the value of annual precipitation in a given year implies nothing about the value to be expected in subsequent years". Lack of autocorrelation is also a prerequisite for using equations based on known statistical distributions.

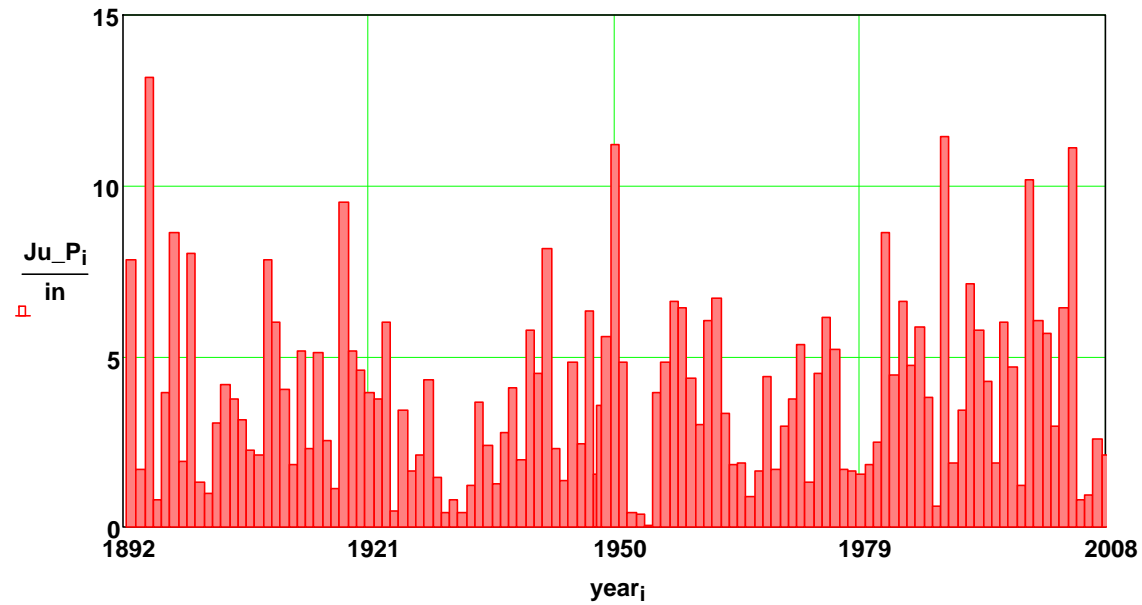
In reality, I don't think anyone knows what level of precipitation will cause the Brake to "dry up" nor have we even defined "dry up". Further, (and this is a legal question) does the fact that cypress seeds need to germinate on soil that is not inundated and the fact that the plant's discharge may affect the availability of such conditions constitute any real, meaningful, damage in this case, or is this simply a straw man ? Does drying up 1 in 4 years instead of 1 in 2 years have any meaningful effect on the Brake's ability to produce Cypress trees ?

In the statements and plots below I examine the January and July precipitation amounts. In both months no trends are evident and the mean and median of the data are close together suggesting the distribution of both January and July precipitation amounts is symmetrical.



The mean of the precipitation amount falling in January is 4.8 inches while the median is 4.3 inches

June Precipitation



$$\text{mean}(Ju_P) = 3.858 \cdot \text{in}$$

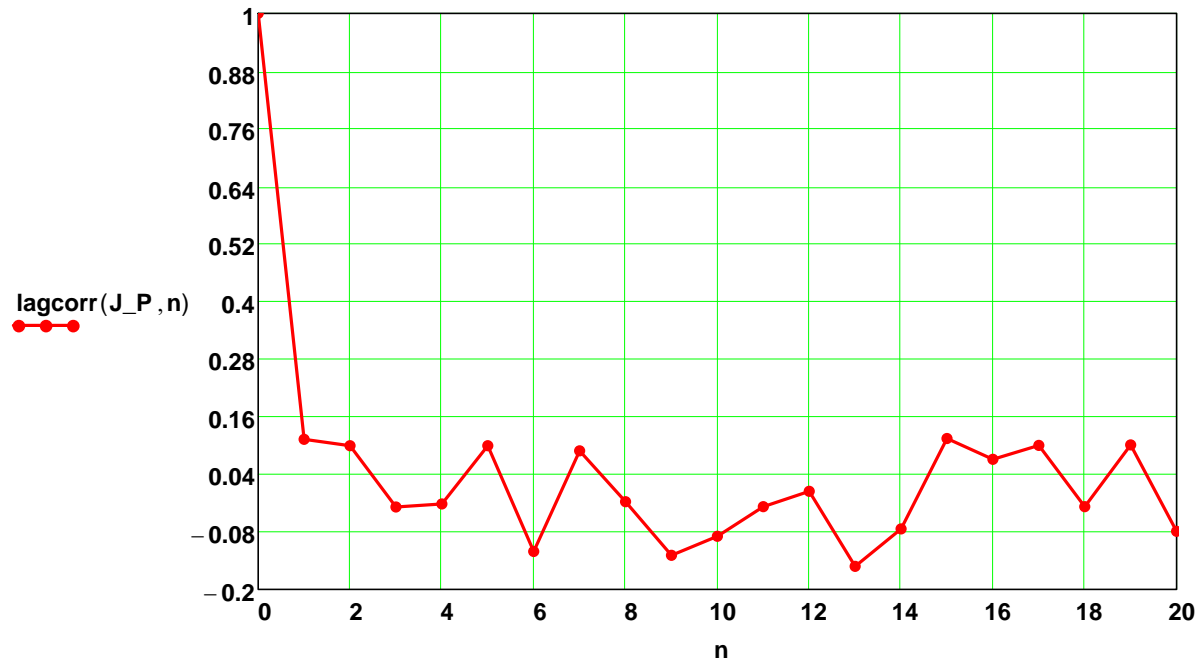
$$\text{median}(Ju_P) = 3.57 \cdot \text{in}$$

The mean of the precipitation amount falling in July is 3.9 inches while the median is 3.6 inches

The statements and plots below show that the precipitations for January and July are not autocorrelated. This equations based on known statistical distributions can be used make predictions about future precipitation values in these months.

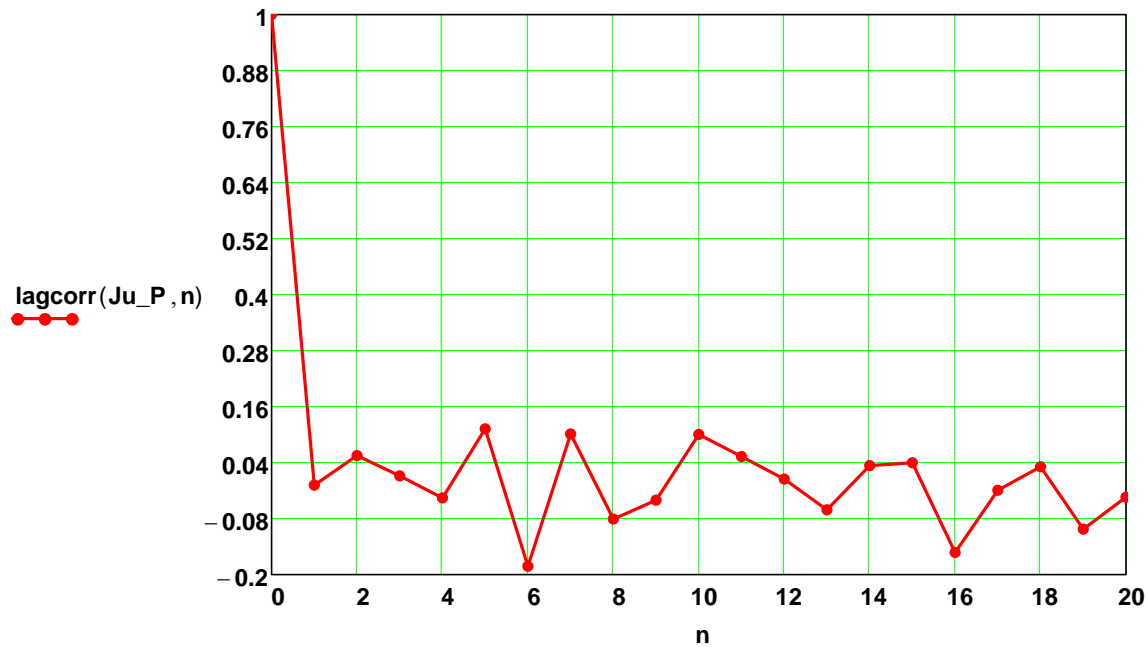
$$\text{lagcorr}(J_P, n) := \frac{1}{\text{length}(J_P) \cdot \text{stdev}(J_P)^2} \cdot \sum_{i=n}^{\text{last}(J_P)} [(J_P_i - \text{mean}(J_P)) \cdot (J_P_{i-n} - \text{mean}(J_P))]$$

Correlation Analysis - Jan



$$\text{lagcorr}(\text{Ju_P}, n) := \frac{1}{\text{length}(\text{Ju_P}) \cdot \text{stdev}(\text{Ju_P})^2} \cdot \sum_{i=n}^{\text{last}(\text{Ju_P})} [(\text{Ju_P}_i - \text{mean}(\text{Ju_P})) \cdot (\text{Ju_P}_{i-n} - \text{mean}(\text{Ju_P}))]$$

Correlation Analysis - June



The statements and plots below are used to produce histograms of the precipitation amounts in January and July

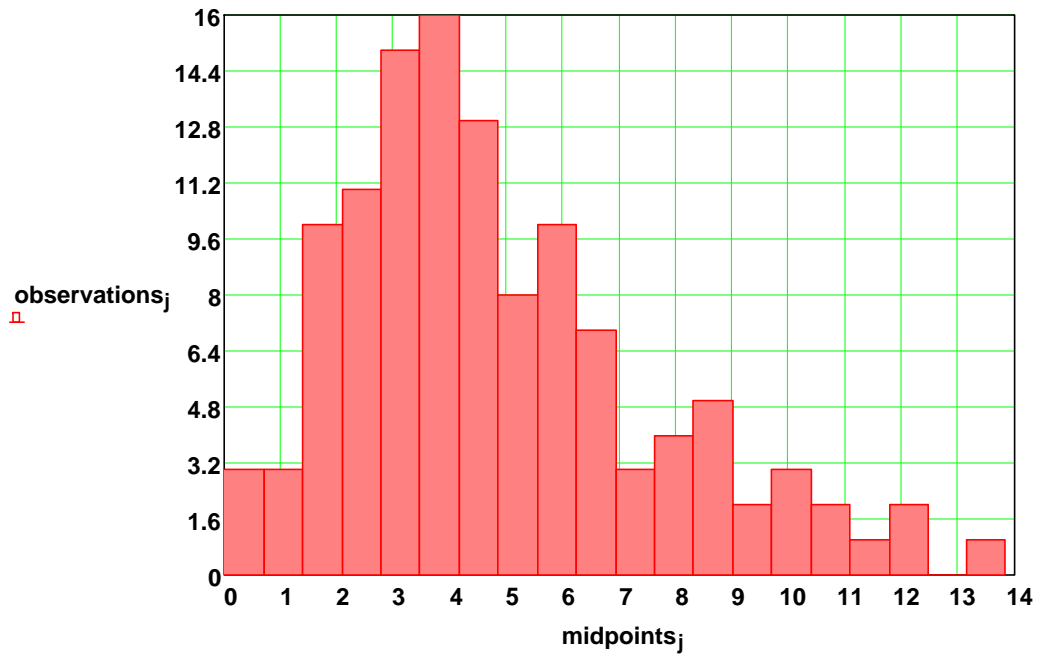
$$\text{Annual_Precip_histogram} := \text{histogram}\left(20, \frac{\text{J_P}}{\text{in}}\right) =$$

	0	1
0	0.346	3
1	1.039	3
2	1.731	10
3	2.424	11
4	3.116	15
5	3.809	16
6	4.501	13
7	5.194	8
8	5.886	10
9	6.579	7
10	7.271	3
11	7.964	4
12	8.656	5
13	9.349	2
14	10.041	3
15	10.734	...

$$\text{observations}_j := \text{histogram}\left(20, \frac{\text{J_P}}{\text{in}}\right)_{j,1}$$

$$\text{midpoints}_j := \text{histogram}\left(20, \frac{\text{J_P}}{\text{in}}\right)_{j,0}$$

Histogram - January Precipitation



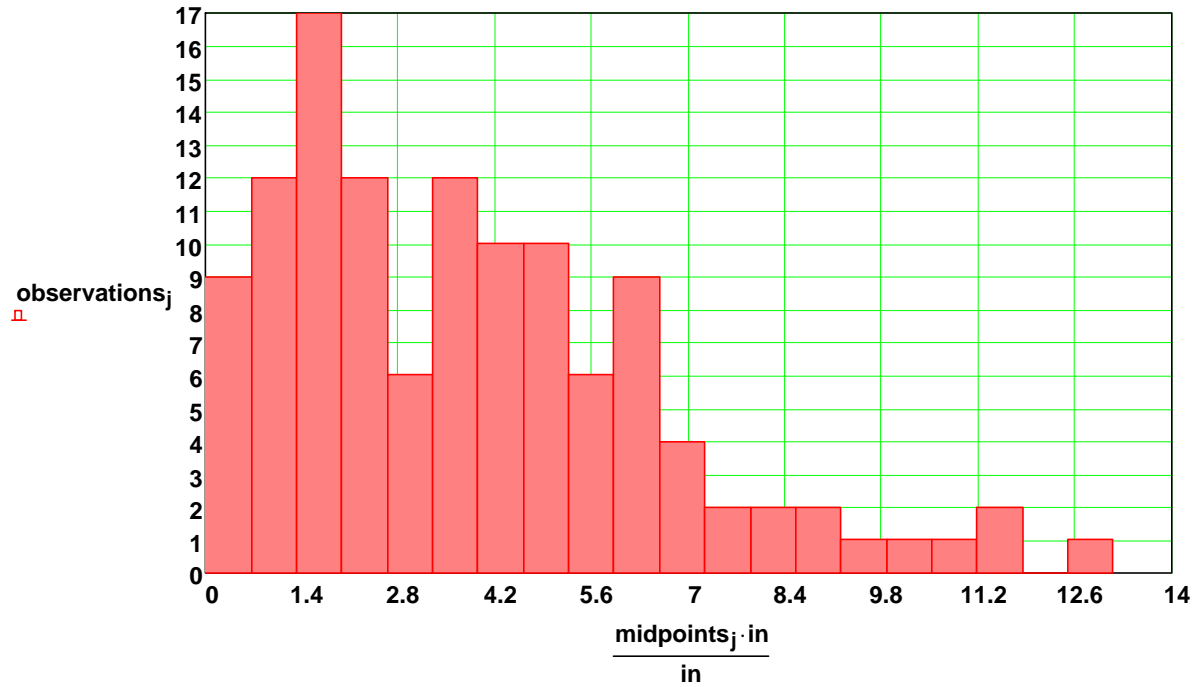
$$\text{Annual_Precip_histogram} := \text{histogram}\left(20, \frac{\text{Ju_P}}{\text{in}}\right) =$$

	0	1
0	0.329	9
1	0.986	12
2	1.644	17
3	2.301	12
4	2.959	6
5	3.616	12
6	4.274	10
7	4.931	10
8	5.589	6
9	6.246	9
10	6.904	4
11	7.561	2
12	8.219	2
13	8.876	2
14	9.534	1
15	10.191	...

$$\text{observations}_j := \text{histogram}\left(20, \frac{\text{Ju_P}}{\text{in}}\right)_{j,1}$$

$$\text{midpoints}_j := \text{histogram}\left(20, \frac{\text{Ju_P}}{\text{in}}\right)_{j,0}$$

Histogram - June Precipitation



Both histograms are skewed to the right, meaning that more "higher" values of precipitation are to be expected than would be if the data followed a normal distribution.

Summary

I am not sure if it is possible (at least for me) to make explanatory statements understandable to folks without some background in probability and statistics, sorry.

1. The mean and the median of the annual precipitation amounts are very close to each other, this suggests the distribution of annual precipitation amounts is symmetrical.
2. Plots of annual precipitation amounts do not show long term trends
3. Probability plots of annual precipitation amounts and summer precipitation amounts are shown in the figure labeled "Probability Plot - Rainfall - Monroe 1892 -2009". Using these plots it is possible to determine the probability of future rainfall amounts. Examples are given adjacent to the plot.
4. The plot labeled "Histogram of Annual Precipitation Values" shows that annual precipitation values can be described by a normal distribution
5. The correlation analysis shows that the annual precipitation amounts are not "autocorrelated", that is the precipitation value in the current year does not imply anything about the value to be expected in subsequent years.
6. Items 4 and 5 above imply that equations developed for the "Normal Distribution" can be used to predict the probability of future annual precipitation values.
7. Based on the analyses above the following statements can be made:

The most likely value of annual precipitation in any future year is 51 inches

There is a 50% chance that in any future year the annual precipitation amount will be less than 50 inches

Approximately 95% of future annual precipitation values will fall within 2 standard deviations of the mean of the data, approximately 99 % of future annual precipitation values will fall within 3 standard deviations of the mean of the data.