

0.1 Whether Mass shootings follow Poisson Distribution

0.1.1 Daily Data does not Follow Poisson Distribution

Using the software we get the average number of shootings for each day in the week as shown below:

Day	Average Number of Shooting
Sunday	1.73
Monday	0.61
Tuesday	0.55
Wednesday	0.55
Thursday	0.49
Friday	0.68
Saturday	1.46

Table 1: Average Number of Shooting for Each Day

We find that the average number of shootings on weekend is much larger than those on the weekdays. However, if it follows a Poisson distribution, it should be the same for weekend and weekdays. In the following part, we would use Fisher's null hypothesis testing to reject the idea that it has same probability to have mass shootings on weekday and weekend. Because our data is so big, we can regard it as a normal distribution.

$$H_0 : \mu_1 = \mu_2$$

Type	Data Number n	Average Number of Shooting \bar{X}	Standard Deviation S
Weekday	1304	0.58	0.8089
Weekend	522	1.60	1.3740

Table 2: Average Number of Shooting for Weekday and Weekend

As the variances have a big difference, we would use the pooled T-Test with unequal variances. Apply Smith-Satterthwaite to get the γ .

$$\gamma = \frac{(S_1^2/n_1 + S_2^2/n_2)^2}{\frac{(S_1^2/n_1)^2}{n_1-1} + \frac{(S_2^2/n_2)^2}{n_2-1}} = 1247.7$$

we round it down to $\gamma = 1247$, so

$$T_\gamma = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)_0}{\sqrt{S_1^2/n_1 + S_2^2/n_2}} = -8.805$$

From the table we get $t_{0.025,300} = 1.968$ so we can see $T_\gamma >> t_{0.025,300} > t_{0.025,1247}$ so we can reject H_0 at the 0.05 level of significance.

By the Fisher's null hypothesis testing, we can see the possibility for mass shootings in weekday and weekend is not same, so it would reject the idea that the number of shootings daily follows a Poisson distribution. Because if it follows a Poisson distribution, it would have same probability to have mass shootings on weekday and weekend

0.1.2 Whether Weekly Data Follow Poisson Distribution

If we consider the data weekly, because the week starts on Sunday, we can get 260 full weeks from the date, we can get the results as below.

Data Number n	Average Number of Shooting \bar{X}	Standard Deviation S
260	6.07	5.4448

Table 3: Average Number of Shooting for Each Day

If it follows a Poisson distribution, then we should get the result that $\mu = \sigma$, we still use Fisher's null hypothesis testing here.

$$H_0 : \mu = S$$

Here we would use T-test with

$$T_n = \frac{\bar{X} - \mu}{S/\sqrt{n}} = 1.850$$

From the table we get $t_{0.025,300} = 1.968$ so we can see $T_n < t_{0.025,300} < t_{0.025,259}$ so we should not reject H_0 at the 0.05 level of significance.

Though we do not reject the H_0 , we can see the T_n is only slightly smaller than $t_{0.025,300}$ so we would like to discuss further this problem. If we only consider the individual year, we let 52 weeks to become a year and get the following results; (note: here the year is not same as the year we usually discussed)

Year	Data Number n	Average Number of Shooting \bar{X}	Standard Deviation S
1	52	4.81	2.6645
2	52	5.19	2.6423
3	52	6.42	3.0828
4	52	7.27	3.7632
5	52	6.65	2.8414

Table 4: Average Number of Shooting for Week by Year

If we see it yearly, we would find the standard deviation in each individual year is much smaller than we consider it as a whole. Here we would test the fourth year as an example to see whether it following a Poisson distribution in each individual year.

If it follows a Poisson distribution, then we should get the result that $\mu = \sigma$, we still use Fisher's null hypothesis testing here.

$$H_0 : \mu = \sigma$$

Here we would use T-test with

$$T_n = \frac{\bar{X} - \mu}{S/\sqrt{n}} = 6.720$$

From the table we get $t_{0.025,50} = 2.008$ so we can see $T_n > t_{0.025,50} > t_{0.025,51}$ so we should reject H_0 at the 0.05 level of significance.

Similarly, we can also reject the idea the number of mass shootings for week follows a Poisson distribution in each individual year.

To conclude, we would say the number of mass shootings for weeks does not follow a Poisson distribution. Though if we see the 5 years together, it would appear to follow, when we see the year separately, the result suggests it does not follow.

0.1.3 Weather Yearly Data Follow Poisson Distribution

Then we consider whether the number for year follows the Poisson distribution, we collect the data as follows.

Year	Data Number n	Number of Shooting X
2013	365	252
2014	365	270
2015	365	335
2016	366	382
2016	365	346

Table 5: Average Number of Shooting for Each Year

By calculation, we get

$$\bar{X} = 317.0 \quad \sigma = 2956.00$$

Because the data number $n=5$ is too small, we can't apply the center-limit law. But the huge difference between the \bar{X} and σ suggest that it may not follow a Poisson distribution.

Finally, we would conclude that the number of mass shootings does not follow a Poisson distribution on the large timescale. It may because the shooting is not a random event. People may tend to shoot when they see others shooting. Or it may due to in some month, the economic or social security is not good, it needs further analysis to give a more concrete result.

1 Weeks with n Shootings

1.1 Predicted and Actual Number of Weeks with n Shootings

By the data above we can draw the actual grapg like below.



Figure 1: Actual Number of Weeks with n Shootings

Though we do not think the number of mass shootings follows the Poisson distribution, since the number is big enough to apply the central limit law, we would use the normal distribution to

draw the graph. We would use $\mu = \bar{X} = 6.07$ and $\sigma = S = 5.4448$ then we get the weeks with n mass shootings would be

$$f(n) = \frac{a}{\sqrt{2\pi}} e^{-((n-\mu)/\sigma)^2/2}$$

f(n) would be round up to the nearest integer, then we get the following graph. By testing, we get a=53.8 which allows the sum to be 260.

Then the graph is like below

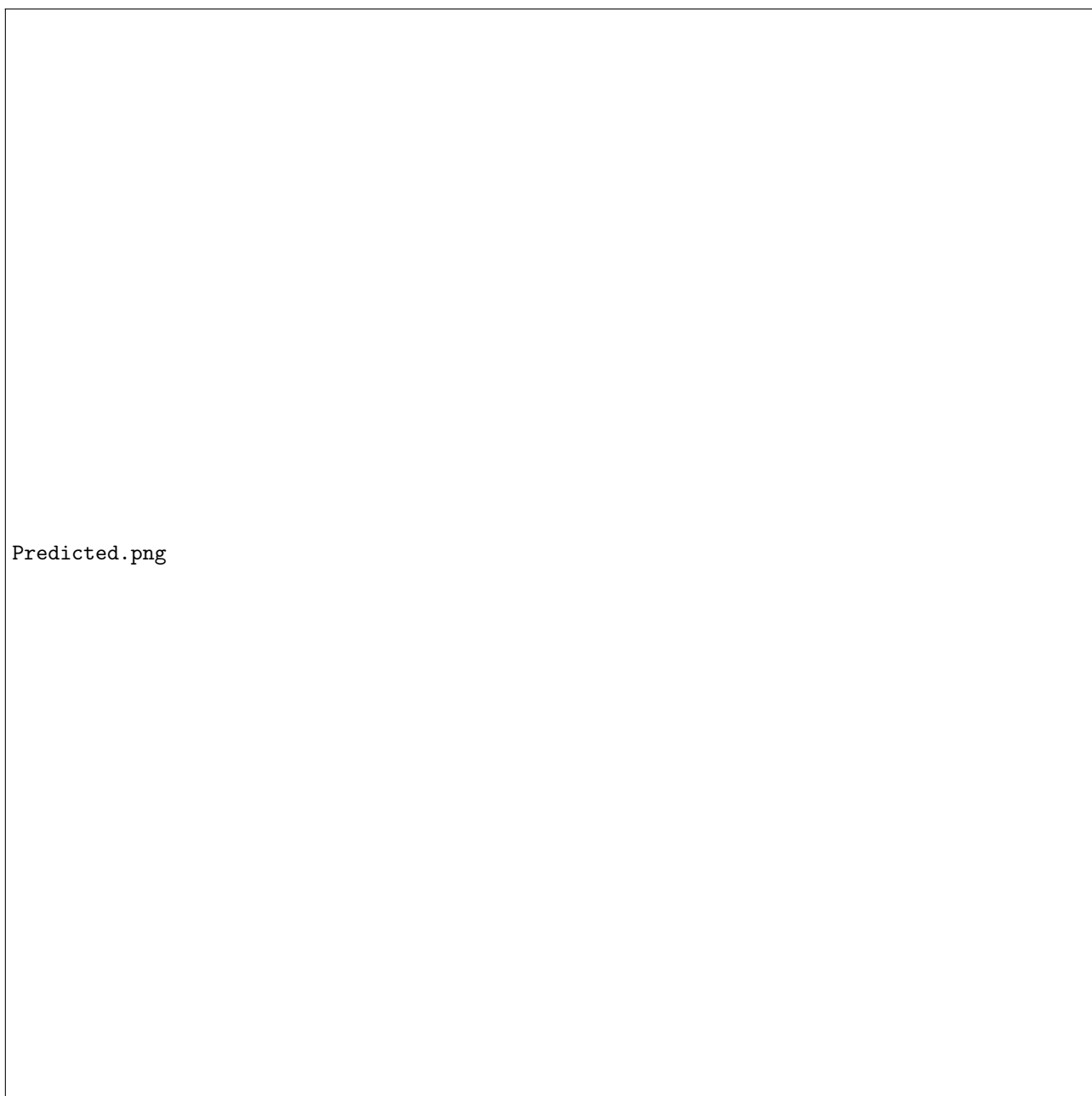


Figure 2: Predicted Number of Weeks with n Shootings

The big difference with those two picture is caused by the variance. In fact, the variance is much smaller if we consider each year separately, So the actual graph is more steep.

2 Reference