

**MODELING CA  
WILDFIRES w/POISSON  
PROCESS**

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STAT 482**



# BACKGROUND







Chapparal



Coastal Bluffs



Grasslands



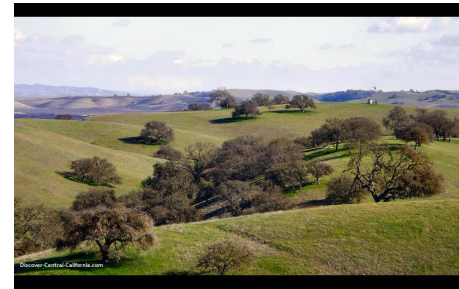
Sloughs



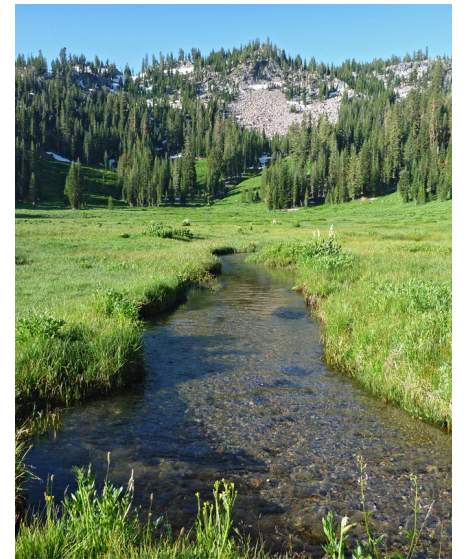
Joshua Tree  
Woodlands



Coniferous Forests



Oak Savannas



Wet Meadows

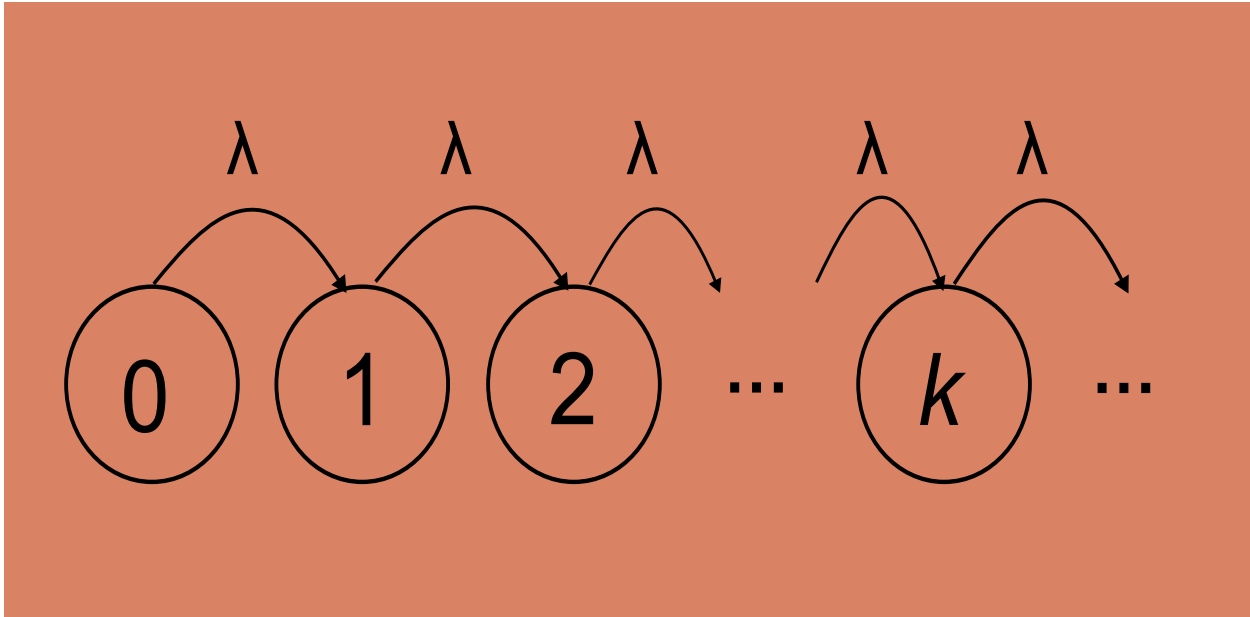


# Wildfires



- Wildfires are a natural process in the ecosystem, especially in the Western United States
- Fires remove low-growing underbrush, cleans the floor of debris, opens it up to sunlight, and nourishes the soil
- Provides habitat making way for new generations of plants, some of which require fire to germinate (sprout)
- Fire kills diseases and insects that prey on trees and provides valuable nutrients that enrich the soil
- This is all good, however, the frequency with which wildfires are occurring now negate all these benefits





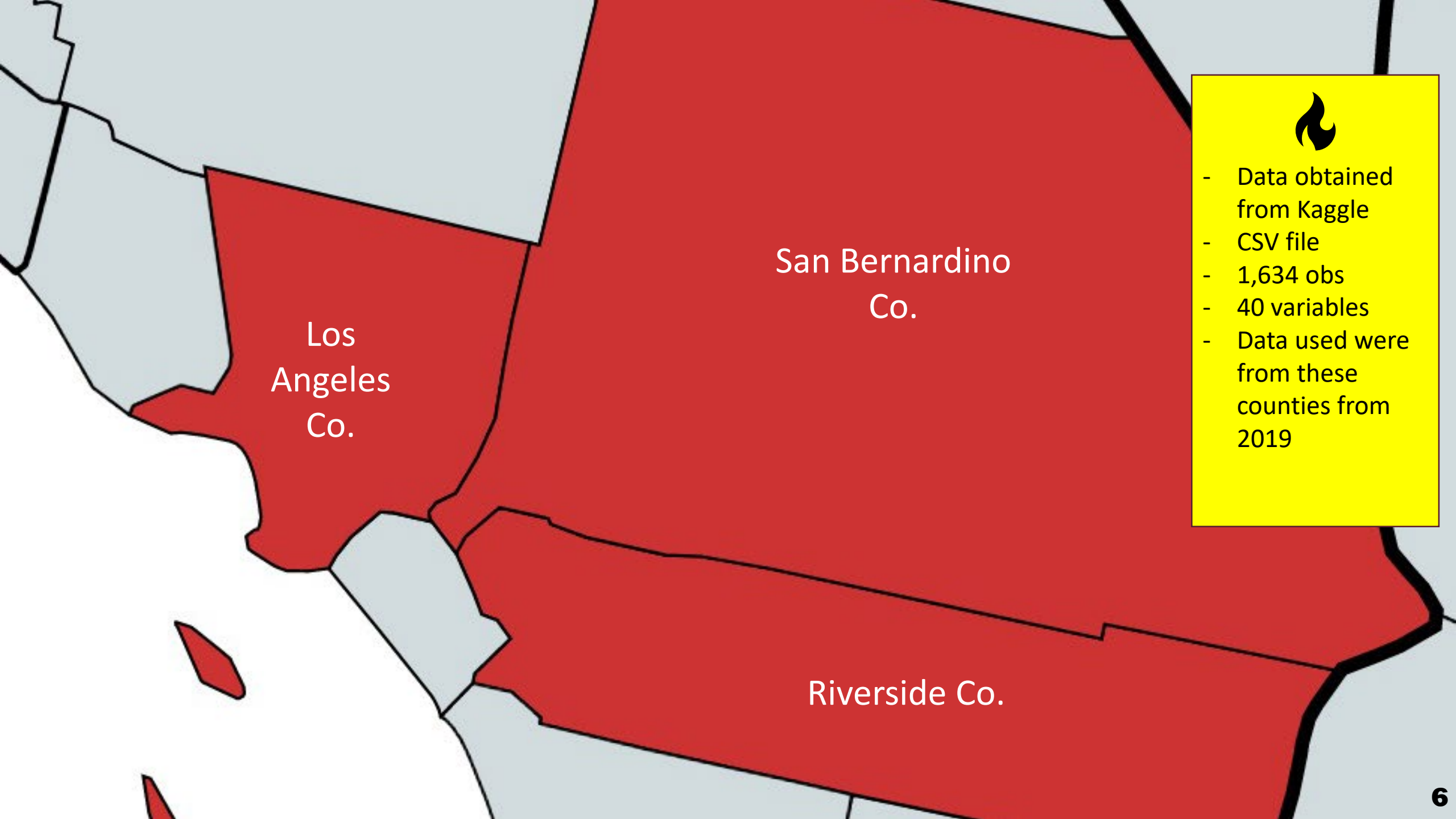
# Poisson Process

- A stochastic process  $\{N(t), t \geq 0\}$  is called a counting process if  $N(t)$  gives the total number of events occurring by time  $t$
- A counting process  $\{N(t), t \geq 0\}$  is called a Poisson process with rate  $\lambda$ , if:
  - (i) no events occur at time 0, i.e.,  $N(0) = 0$
  - (ii) it has independent increments,
  - (iii) it has stationary increments
  - (iv)  $P(N(t) = n) = \frac{(\lambda t)^n}{n!} e^{-\lambda t}$









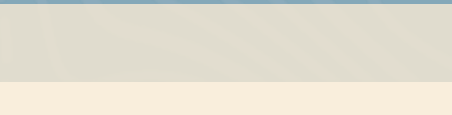
Los Angeles Co.

San Bernardino Co.

Riverside Co.



- Data obtained from Kaggle
- CSV file
- 1,634 obs
- 40 variables
- Data used were from these counties from 2019

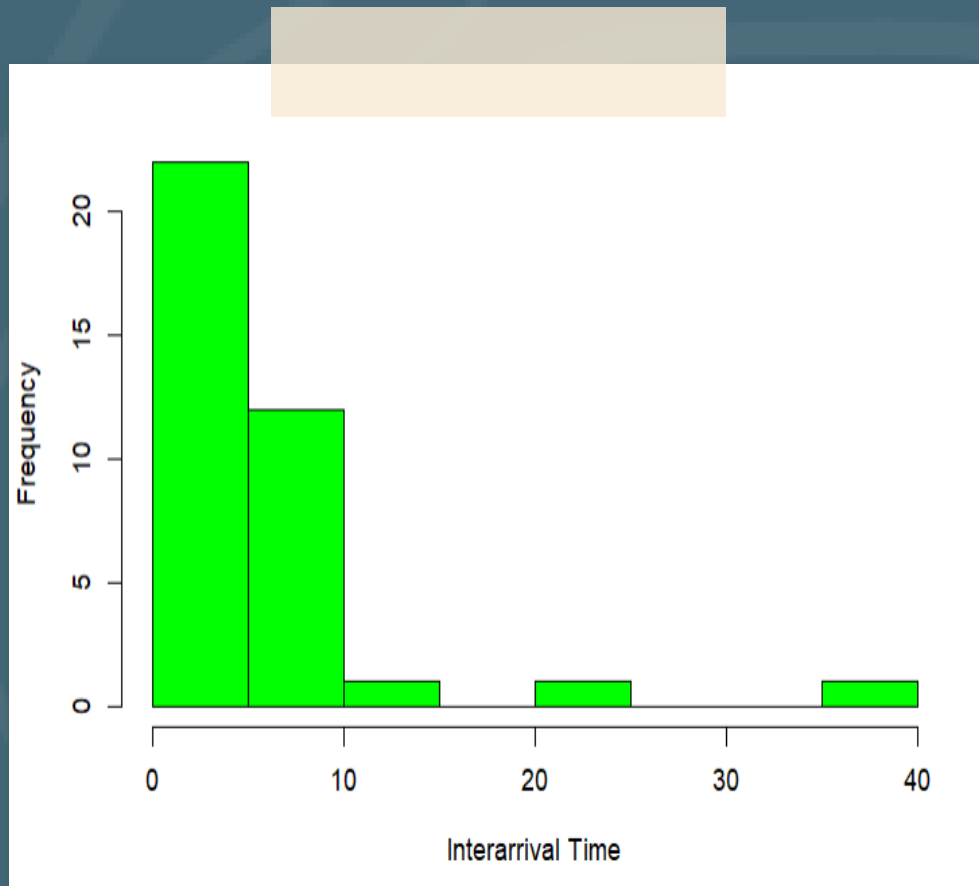


**Excel data used in modeling after filtering and cleaning the original**

5	5/29/2019	San Bernardino					
6	5/30/2019	Riverside					
7	6/21/2019	Los Angeles					
8	6/27/2019	Los Angeles					
9	6/29/2019	Riverside					
10	7/11/2019	Los Angeles					
11	7/12/2019	Riverside					
12	7/13/2019	Riverside					
13	7/16/2019	Riverside					
14	7/22/2019	Riverside					
15	7/28/2019	Riverside					
16	8/3/2019	Riverside					
17	8/5/2019	San Bernardino					
18	8/12/2019	Riverside					
19	8/13/2019	Los Angeles					
20	8/22/2019	San Bernardino					
21	8/25/2019	San Bernardino					
22	8/27/2019	Riverside					
23	9/2/2019	Los Angeles					







## HISTOGRAM



```
incidents <- read.csv(file = "...", header = TRUE, sep = ",")
```

```
# creating date-time variable
datetime<- as.POSIXct(as.Date(incidents$Date, "%m/%d/%Y"))
datetime
```

```
# computing lag
datetime.lag<- c(0,head(datetime, -1))
```

```
# computing inter arrival times (in hours) and removing 1st val
int.time<- (as.numeric(datetime)-as.numeric(datetime.lag))/(3600*24)
int<- int.time[-1]
```

```
# plotting histogram
hist(int, main="", col="green", xlab="Interarrival Time")
```

```
#binning inter arrival times
binned.int<- as.factor(ifelse(int<5,"1",
  ifelse(int>=5 & int<10,"2",ifelse(int>=10 & int<15,"3","4"))))
```

```
#computing observed frequencies
obs<- table(binned.int)
obs
```

```
#estimating mean for exponential distribution
mean.est<- mean(int)
```

```
#computing expected frequencies
exp<- c(1:4)
exp[1]<- length(int) * (1-exp(-5/mean.est))
exp[2]<- length(int) * (exp(-5/mean.est)-exp(-10/mean.est))
exp[3]<- length(int) * (exp(-10/mean.est)-exp(-15/mean.est))
exp[4]<- length(int) * exp(-15/mean.est)
round(exp,1)
```

```
#computing chi-squared statistic
print(chi.sq<- sum((obs-exp)^2/exp))
```

```
#computing p-value
print(p.value<- 1-pchisq(chi.sq, df=2))
```

```

> obs
binned.int
 1  2  3  4
22 11  2  2
> round(exp,1)
[1] 21.9  8.9  3.6  2.5
> #computing chi-squared statistic
> print(chi.sq<- sum((obs-exp)^2/exp))
[1] 1.317218
> #computing p-value
> print(p.value<- 1-pchisq(chi.sq, df=2))
[1] 0.5175706

```

## Goodness-of-Fit Test

The p-value is larger than 0.05, indicating that the wildfires in the given time frame occurred according to a Poisson process.



```

incidents <- read.csv(file = "...", header = TRUE, sep = ",")

# creating date-time variable
datetime<- as.POSIXct(as.Date(incidents$Date, "%m/%d/%Y"))
datetime

# computing lag
datetime.lag<- c(0,head(datetime, -1))

# computing inter arrival times (in hours) and removing 1st val
int.time<- (as.numeric(datetime)-as.numeric(datetime.lag))/(3600*24)
int<- int.time[-1]

# plotting histogram
hist(int, main="", col="green", xlab="Interarrival Time")

#binning inter arrival times
binned.int<- as.factor(ifelse(int<5,"1",
  ifelse(int>=5 & int<10,"2",ifelse(int>=10 & int<15,"3","4"))))

#computing observed frequencies
obs<- table(binned.int)
obs

#estimating mean for exponential distribution
mean.est<- mean(int)

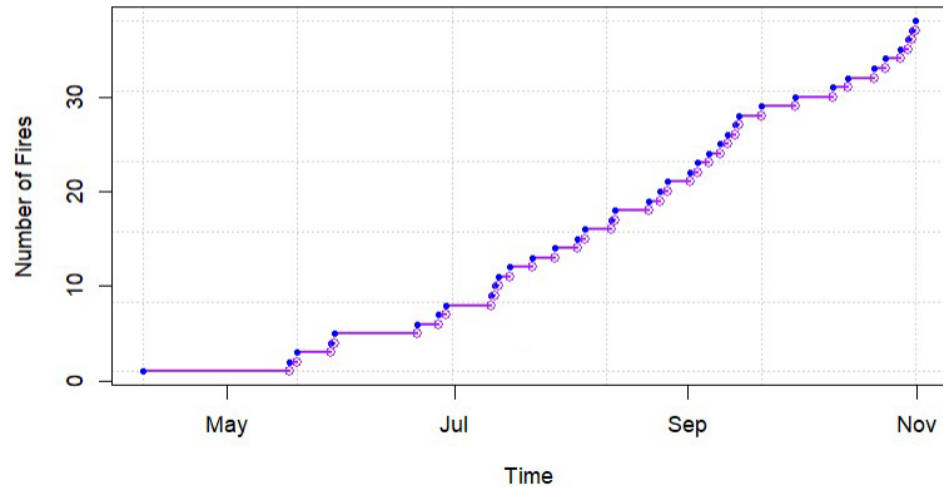
#computing expected frequencies
exp<- c(1:4)
exp[1]<- length(int)*(1-exp(-5/mean.est))
exp[2]<- length(int)*(exp(-5/mean.est)-exp(-10/mean.est))
exp[3]<- length(int)*(exp(-10/mean.est)-exp(-15/mean.est))
exp[4]<- length(int)*exp(-15/mean.est)
round(exp,1)

#computing chi-squared statistic
print(chi.sq<- sum((obs-exp)^2/exp))

#computing p-value
print(p.value<- 1-pchisq(chi.sq, df=2))

```





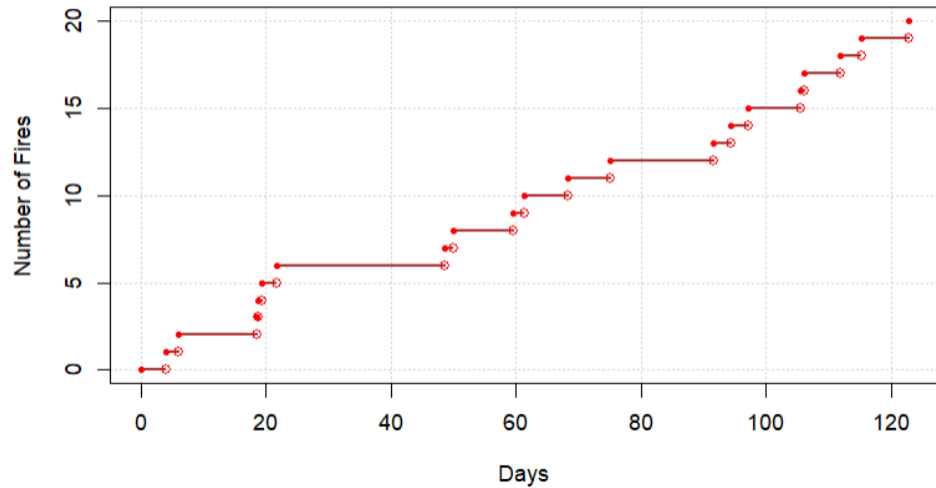
## Plotting Actual Trajectory



```
fires <- c(1:38)
date<- as.POSIXct(as.Date(incidents$Date,
"%m/%d/%Y"))
```

```
#plotting stock price against date
plot(date, fires, type="n",
      xlab="Time", ylab="Number of Fires",
      first.panel=grid())
```

```
segments(date[-length(date)], fires[-
length(date)], date[-1]-0.07, fires[-
length(date)],
         lwd=2, col="purple")
points(date, fires, pch=20, col="blue")
points(date[-1], fires[-length(date)],
       pch=1, col="purple")
```



## Plotting Simulated Trajectory



```
# creating date-time variable
datetime<- as.POSIXct(as.Date(incidents$Date, "%m/%d/%Y"))
datetime

# computing lag
datetime.lag<- c(0,head(datetime, -1))

# computing inter arrival times (in hours) and removing 1st val
int.time<- (as.numeric(datetime)-as.numeric(datetime.lag))/(3600*24)
int<- int.time[-1]

#estimating mean for exponential distribution
mean.est<- mean(int)

lambda.est <- 1 / mean.est
t<- 10
nfires <- 20

#defining states
N<- 0:nfires

#setting time as vector
time<- c()

#setting initial value for time
time[1]<- 0

#specifying seed
set.seed(483650)

#simulating trajectory
for (i in 2:(nfires+1))
  time[i]<- time[i-1]+round((-1/lambda.est)*log(runif(1)),2)

#plotting trajectory
plot(time, N, type="n", xlab="Days", ylab="Number of Fires", panel.first =
grid())
segments(time[-length(time)], N[-length(time)], time[-1]-0.07, N[-length(time)],
          lwd=2, col="brown")
points(time, N, pch=20, col="red")
points(time[-1], N[-length(time)], pch=1, col="red")
```



```
datetime<-
as.POSIXct(as.Date(incidents$Date,
"%m/%d/%Y"))

datetime.lag<- c(0,head(datetime, -1))

int.time<- (as.numeric(datetime)-
as.numeric(datetime.lag))/(3600*24)
int<- int.time[-1]

mean.est<- mean(int)
```

```
> mean.est
[1] 5.567568
```

**Wait Between Fires  
(days)**

```
datetime<-
as.POSIXct(as.Date(incidents$Date,
"%m/%d/%Y"))

datetime.lag<- c(0,head(datetime, -1))

int.time<- (as.numeric(datetime)-
as.numeric(datetime.lag))/(3600*24)
int<- int.time[-1]

mean.est<- mean(int)

lambda.est <- 1/mean.est
```


```
> lambda.est
[1] 0.1796117
```

**Estimated Lambda  
(fires per day)**

Date	Mean	Next Fire
11/1/2019	5.56	11/6/2019

**Next Fire**

**FURTHER RESULTS**

A monarch butterfly with orange and black wings is perched on a cluster of small pink flowers. The background is a soft, out-of-focus green and pink. The text "THANK YOU!" is overlaid in the center in a white, sans-serif font. The word "THANK" is in all caps and a smaller font size than "YOU!".

THANK **YOU!**

