

# Discrete Response Model

## Lecture 5

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Models for Count Response, Discrete Response Model Evaluation, and Model Selection

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# An Example (continue)

# Example

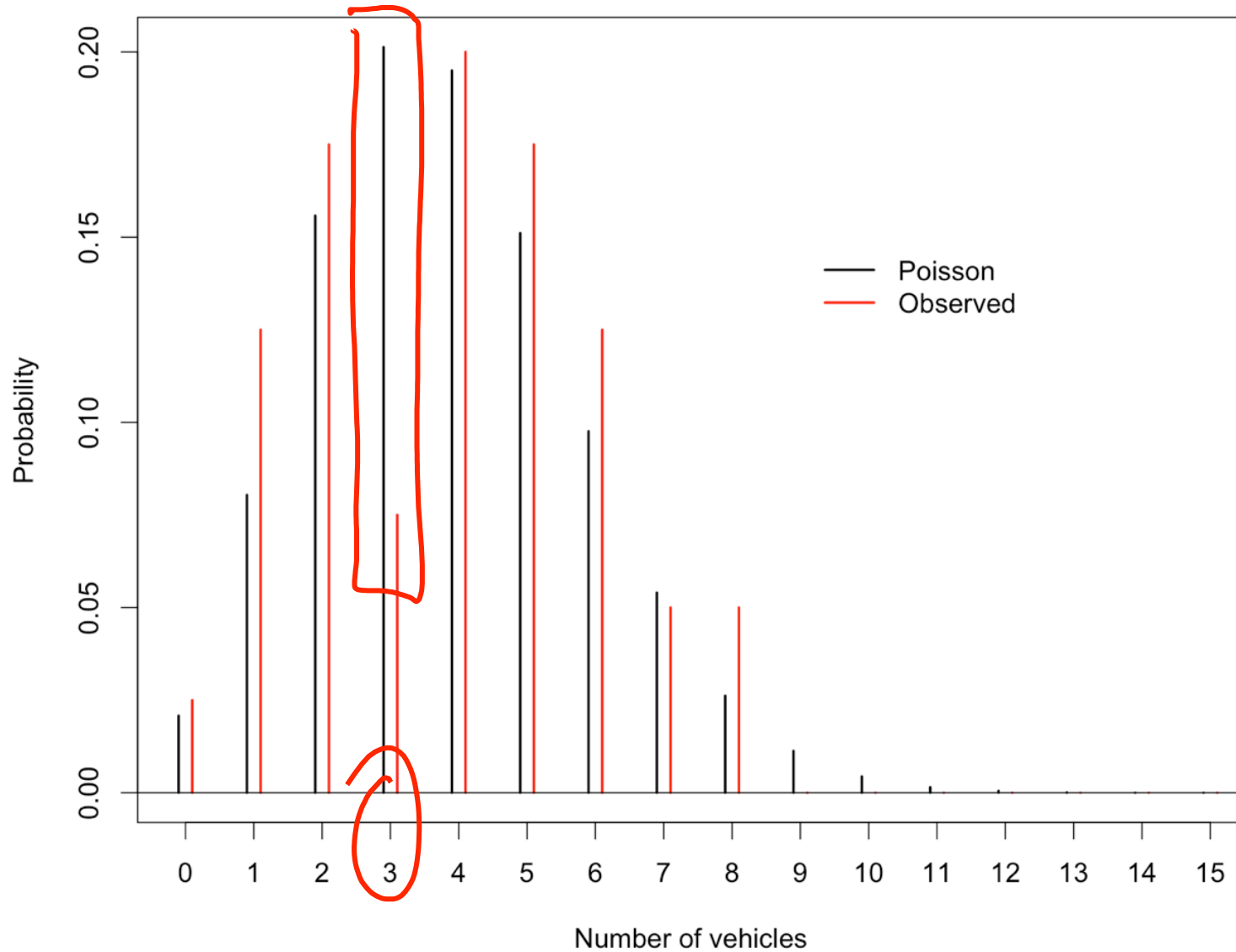
```
> mean(stoplight$vehicles)
[1] 3.875
> var(stoplight$vehicles)
[1] 4.317308
> table(stoplight$vehicles) #Note that y = 0, 1, ..., 8 all have positive counts
```

```
0 1 2 3 4 5 6 7 8
1 5 7 3 8 7 5 2 2
```

```
> rel.freq <- table(stoplight$vehicles)/length(stoplight$vehicles)
> rel.freq2 <- c(rel.freq, rep(0, times = 7))
> y <- 0:15
> prob <- round(dpois(x = y, lambda = mean(stoplight$vehicles)), 4)
> data.frame(y, prob, rel.freq = rel.freq2)
```

	y	prob	rel.freq
1	0	0.0208	0.025
2	1	0.0804	0.125
3	2	0.1558	0.175
4	3	0.2013	0.075
5	4	0.1950	0.200
6	5	0.1511	0.175
7	6	0.0976	0.125
8	7	0.0540	0.050
9	8	0.0262	0.050
10	9	0.0113	0.000
11	10	0.0044	0.000
12	11	0.0015	0.000
13	12	0.0005	0.000
14	13	0.0001	0.000
15	14	0.0000	0.000
16	15	0.0000	0.000

# Example



# Example

## Wald confidence interval

```
> mu.hat <- mean(stoplight$vehicles)
> mu.hat + qnorm(p = c(alpha/2, 1 - alpha/2))*sqrt(mu.hat/n)
[1] 3.264966 4.485034
```

Note that the Wald interval using the  $\log(\mu)$  transformation is

$$e^{\log(\hat{\mu}) \pm Z_{1-\alpha/2} \sqrt{1/(\hat{\mu}n)}}$$

Exponentiate the  $\log()$  transformation.

```
> exp(log(mu.hat) + qnorm(p = c(alpha/2, 1-alpha/2)) * sqrt(1/(mu.hat*n)))
[1] 3.310561 4.535674
```

## Score C.I.

```
> (mu.hat + qnorm(p = c(alpha/2, 1 - alpha/2))/(2*n)) + qnorm(p = c(alpha/2, 1 - alpha/2)) * sqrt((mu.hat + qnorm(p = 1 - alpha/2)/(4*n))/n)
[1] 3.239503 4.510497
```

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