

$$1) \begin{aligned} H_0: \lambda &\leq 0.5 \\ H_1: \lambda &> 0.5 \end{aligned}$$

The equivalent test statistic  $T(\underline{X}) = \sum_{i=1}^n X_i \sim \text{Poisson}(n\lambda)$

Now  $T$  does not depend on  $\lambda$  so it is the UMP test statistic for the situation above.

We have

$$T(x) = 0 \times 109 + 1 \times 65 + 2 \times 22 + 3 \times 3 + 4 \times 1 = 122$$

$$\begin{aligned} p(x) &= P\left(\sum_{i=1}^n X_i \geq 122; \lambda = 0.5\right) \\ &= 1 - P\text{Pois}(122, 0.5 \times 200) \\ &= 0.0143 \quad (35\text{f}) \end{aligned}$$

$$\begin{aligned} 2a) T(x) \sim \text{Poisson}(n\lambda) &\approx \text{Normal}(n\lambda, n\lambda) \\ \Rightarrow P(x) &= P\left(Z \geq \frac{122 - 100}{\sqrt{100}}\right) \\ &= P(Z \geq 2.2) \\ &= 1 - p_{\text{norm}}(2.2) \\ &= 0.0139 \quad (35\text{f}) \end{aligned}$$

b) Monte Carlo Integration:

$$F = \int_a^b f(x) dx$$

$$\hat{E}\hat{F}_i = (b-a) \frac{1}{N} \sum_{j=0}^{N-1} f(a + \xi_j(b-a))$$

where  $\xi_j$  = random number uniformly distributed between  $[0, 1]$

[From Internet]

In this case

$$F = \int_{-\infty}^{2.2} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} dx$$

Make substitution  $u = e^x$  to get rid of  $-\infty$ .

$$\begin{aligned} du &= e^x dx = u dx \\ F &= \int_0^{e^{2.2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(\ln u)^2} \frac{1}{u} du \end{aligned}$$

Approx p-value = 0.132. See R-code.

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$$\begin{aligned} 3a) H_0: \lambda &= 0.55 \\ H_1: \lambda &\neq 0.55 \end{aligned}$$

$$\begin{aligned} \Lambda_n(x) &= \frac{\sup_{\theta \in \{0.55\}} f_n(x; \theta)}{\sup_{\theta \in \mathbb{R}_{>0}} f_n(x; \theta)} \\ &= \frac{f_n(x; 0.55)}{f_n(x; \hat{\lambda}_n)} \end{aligned}$$

$\hat{\lambda}_n$  is ML Estimate .

$$\begin{aligned} T_n(x) &= -2 \log \Lambda_n \\ &= -2 [\ell(0.55; x) - \ell(\hat{\lambda}_n; x)] \xrightarrow{D} \chi^2_1 \end{aligned}$$

So  $T_n(x) = 1.263926$  from R.

$$p(x) = 1 - \text{pchisq}(T_n(x), df=1) = 0.2609 \dots$$

So Retain  $H_0$ .

As  $T_n \xrightarrow{D} \chi^2_1 = Z^2$  by Rule of Thumb over 95% CI is

$$\begin{aligned} \{ \theta \in \mathbb{R}_{>0} : \ell(\theta; x) \geq \ell(\hat{\lambda}_n; x) - 2 \} \\ = \{ \theta \in \mathbb{R}_{>0} : \ell(\theta; x) \geq -184.304 \} \end{aligned}$$

By R we get CI =  $\{ \theta \in \mathbb{R}_{>0} : 0.51 \leq \theta \leq 0.72 \}$

b) At 10% we still retain  $H_0$  as our p-value is  $0.26 \dots > 0.1$ .

# Problem Sheet 8

## Problem Sheet 8

### Question 1

```
p = 1 - ppois(122,0.5*200)
p
```

```
## [1] 0.01430626
```

### Question 2a

```
1 - pnorm(2.2)
```

```
## [1] 0.01390345
```

### Question 2b

```
f <- function(u) (1/sqrt(2*pi)) * exp(-(log(u)^2) / 2) / u
a <- 0
b <- exp(2.2)

mc.int <- function(f, a, b, N) {
  F <- 0
  for (i in 0:(N-1)) {
    xi <- runif(1, 0, 1)

    F <- F + f(a + xi*(b - a))
  }
  F <- F * (b - a) / N

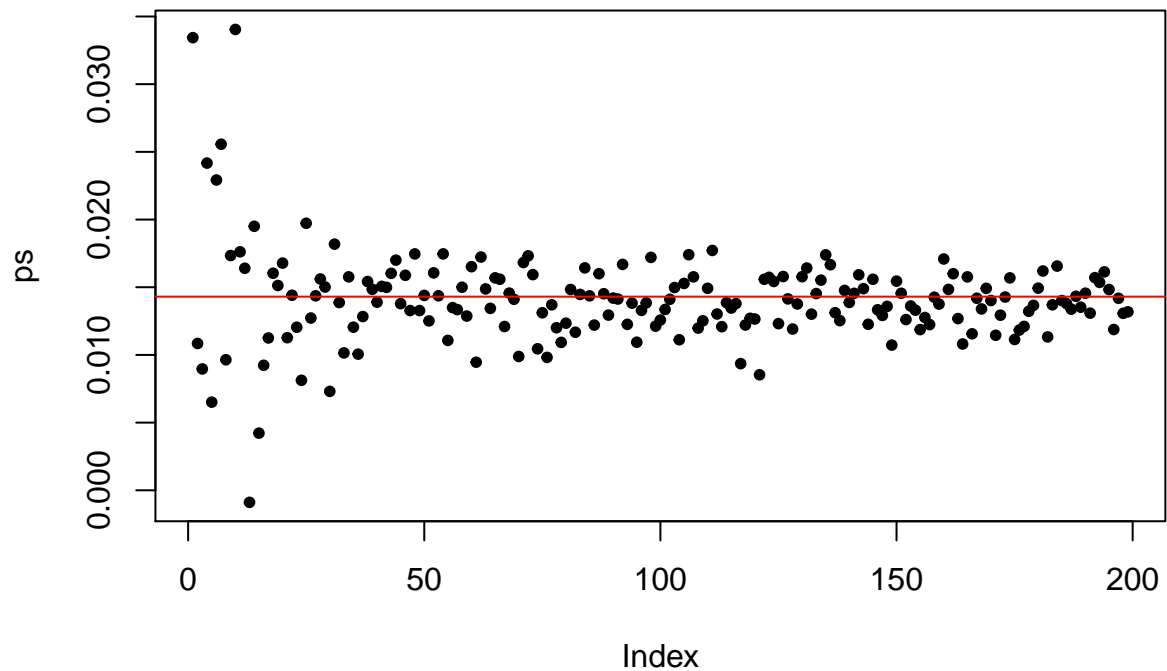
  1-F
}

ps <- vector()
for (N in seq(10000, 1000000, 5000)) {
  ps <- c(ps, mc.int(f, a, b, N))
}

tail(ps, 1)
```

```
## [1] 0.01318355
```

```
plot(ps, pch = 20)  
abline(h=p, col="red")
```



### Question 3a

```
e11 <- function(lambda, xobs) {  
  stopifnot(all(lambda > 0))  
  n <- length(xobs)  
  n * (-lambda + mean(xobs) * log(lambda))  
}  
  
xobs <- c(rep(0, 109), rep(1, 65), rep(2, 22), rep(3, 3), rep(4, 1))  
  
lambda.hat <- mean(xobs)  
  
Tn <- -2*(e11(0.55, xobs) - e11(lambda.hat, xobs))  
  
Tn
```

```
## [1] 1.263926
```

```
1-pchisq(Tn, df=1)
```

```
## [1] 0.2609093
```

```
l = ell(lambda.hat, xobs)-2
```

```
C <- vector()
```

```
for (i in seq(0.01,1,0.01)) {  
  if (ell(i, xobs) >= 1) {  
    C <- c(C, i)  
  }  
}
```

```
c(min(C), max(C))
```

```
## [1] 0.51 0.72
```