

3A) 8.10.10

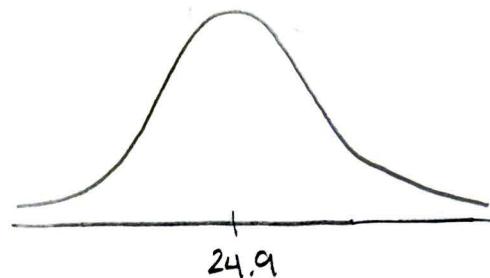
$$\hat{\lambda} = \frac{\sum x_i}{n} \text{ and } x_i \text{ are independent Poisson}$$

Sum of independent Poisson is Poisson:

$$P(\hat{\lambda} = v) = \frac{(n\lambda_0)^{nv} e^{-\lambda_0}}{(nv)!}$$

$$E(\hat{\lambda}) = \lambda_0 \approx \hat{\lambda} = 24.9$$

$$\text{Var}(\hat{\lambda}) = \frac{\lambda_0}{n} \approx 1.04^2$$



Normal approximation:

$$\hat{\lambda} \approx N(24.9, 1.04^2)$$

According to this:

$$P(|\lambda_0 - \hat{\lambda}| > .5) = P\left(\left|\frac{\lambda_0 - \hat{\lambda}}{\sigma_{\hat{\lambda}}}\right| > .5\right) = 2\Phi\left(\frac{-.5}{1.04}\right) = 0.631$$

$$P(|\lambda_0 - \hat{\lambda}| > 1) = 2\Phi\left(\frac{-1}{1.04}\right) = 0.336$$

$$P(|\lambda_0 - \hat{\lambda}| > 1.5) = 0.149$$

$$P(|\lambda_0 - \hat{\lambda}| > 2) = 0.054$$

$$P(|\lambda_0 - \hat{\lambda}| > 2.5) = 0.016$$

3B) $n = 300$, $x_i \sim \text{Pois}(\lambda_0)$

$$\hat{\lambda} = \frac{\sum x_i}{n} = \frac{30 + 2(36) + \dots + 12}{300} = \frac{1168}{300} = 3.893$$

Expected frequency of 0 right turns:

$$P(X=0) \cdot 300 = \frac{(3.893)^0 e^{-3.893}}{0!} \cdot 300 = 6.12 \quad (\text{Use this formula for each})$$

n	Observed	Expected
0	14	6.12
1	30	23.8
2	36	46.3
3	68	60.1
4	43	58.5
5	43	45.6
6	30	29.6
7	14	16.4
8	10	8.0
9	6	3.5
10	4	1.3
11	1	.5
12	1	.2
13	0	.05

← This fit is not very good.
The observed data skews more to the left than the expected distribution

3C

$$E(\hat{\mu}_k) = E\left(\frac{1}{n} \sum_{i=1}^n x_i^k\right) = \frac{1}{n} E\left(\sum_{i=1}^n x_i^k\right)$$

$$= \frac{1}{n} \sum_{i=1}^n E(x_i^k) \quad \text{because the } x_i \text{ are independent}$$

$$= \frac{1}{n} E(x_1^k) \cdot n \quad \text{because they are identically distributed}$$

$$= E(x_1^k) = \mu_k \quad \square$$

3D

$$1) \text{ 1st moment: } \mu_1 = E(X) = \frac{1}{p}$$

$$2) \text{ Invert: } p = \frac{1}{\mu_1}$$

$$3) \text{ Add hats: } \hat{p}_{\text{Mom}} = \frac{1}{\hat{\mu}_1} = \frac{1}{\bar{x}} \quad \text{where } \bar{x} = \frac{1}{n} \sum_{i=1}^n x_i$$

3E

$$a) \Gamma(r+1) = \int_0^\infty t^{r+1-1} e^{-t} dt$$

$$= \int_0^\infty t^r e^{-t} dt$$

$$= -t^r e^{-t} \Big|_0^\infty + r \int_0^\infty t^{r-1} e^{-t} dt$$

$$= r \Gamma(r)$$

$$u = t^r \quad dv = e^{-t} dt$$

$$du = r t^{r-1} dt \quad v = -e^{-t}$$

the first term = 0
because $e^{-t} \rightarrow 0$
faster than $t^r \rightarrow \infty$

$$b) \text{ Chi square density } \rightarrow f_T(t) = \begin{cases} \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{t}} e^{-\frac{t}{2}} & t \geq 0 \\ 0 & \text{else} \end{cases}$$

From homework 2H:

$$\frac{1}{\sqrt{2\pi}} t^{-1/2} e^{-\frac{1}{2}t} = \frac{(\frac{1}{2})^{1/2}}{\Gamma(\frac{1}{2})} t^{-1/2} e^{-\frac{1}{2}t} \leftarrow \text{density of Gamma}(\frac{1}{2}, \frac{1}{2})$$

$$\downarrow$$

$$\frac{1}{\sqrt{2\pi}} = \frac{1}{\sqrt{2} \Gamma(\frac{1}{2})}$$

$$\downarrow$$

$$\sqrt{\pi} = \Gamma(\frac{1}{2}) \quad \square$$