

P.138 (12) Maximum likelihood for Poisson RV:

$$f(k) = P(X=k) = e^{-\lambda} \frac{\lambda^k}{k!}$$

Given k , what value of λ maximizes this quantity?

$$g(\lambda) := e^{-\lambda} \frac{\lambda^k}{k!}$$

$$g'(\lambda) = \frac{1}{k!} [-e^{-\lambda} \lambda^k + e^{-\lambda} k \lambda^{k-1}]$$

$$= \frac{e^{-\lambda} \lambda^{k-1}}{k!} [k - \lambda] \stackrel{!}{=} 0 \Rightarrow \lambda = k$$

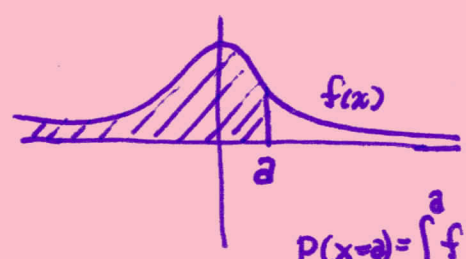
Observe that for Poisson to approximate Binomial, we said $\lambda = np$. Thus for this maximizer $\lambda = k$, we'd have $k = np$ or $p = \frac{k}{n}$ which we found previously. \square

ch 5 Continuum RVs

Here we have a pdf $f_X: \mathbb{R} \rightarrow [0, \infty)$ which is a "density" or height fcn for the probability of the pts in the domain (Note: any single pt has @0!)

$$P(a \leq X \leq b) := \int_a^b f(x) dx$$

$$\text{cdf } P(X \leq a) =: F(a) = \int_{-\infty}^a f(x) dx$$

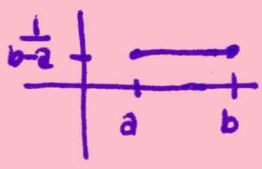


$$P(X=a) = \int_a^a f = 0$$

$$\int_{-\infty}^{\infty} f = 1$$

Uniform RV

$$f(x) = \begin{cases} \frac{1}{b-a} & x \in [a,b] \\ 0 & \text{otherwise} \end{cases}$$



P.153 example: Bus stop
P.181 #9 I arrive at bus stop at 10:00 AM. The time the bus actually arrives is a unif RV over interval [10:00, 10:30] or in minutes [0, 30]

Let $X :=$ time I have to wait

$$P(X > 10) = P(\text{bus arrives in } [10, 30]) = \int_{10}^{30} \frac{1}{30} dx = \frac{20}{30} = \frac{2}{3}$$

But the real question would be: The bus is scheduled at 10:15 and randomly arrives between [10:00, 10:30]. I randomly arrive between [10, 10:30]. What is prob I will catch the bus? (ie bus arrives after I do)

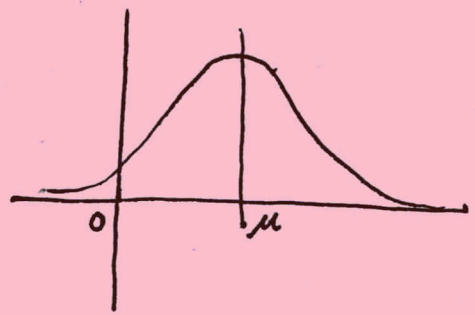
To solve this we need joint pdfs see p.196

Normal RVs

$X \sim \mathcal{N}(\mu, \sigma)$ pdf $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$

Lets show $\int_{-\infty}^{\infty} f(x) dx = 1$:

$I := \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}y^2} dy$
 COV $y := \frac{x-\mu}{\sigma}$
 $dy = \frac{1}{\sigma} dx$

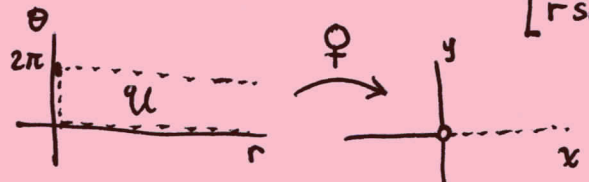


P.155

square both sides:

$I^2 = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-\frac{1}{2}y^2} dy \int_{-\infty}^{\infty} e^{-\frac{1}{2}t^2} dt = \frac{1}{2\pi} \iint e^{-\frac{1}{2}(y^2+t^2)} dy dt$

Make polar COV $\varphi(r, \theta) = \begin{bmatrix} r \cos \theta \\ r \sin \theta \end{bmatrix} = \begin{bmatrix} y \\ t \end{bmatrix}$
 $D\varphi_{r\theta} = \begin{bmatrix} \cos \theta & -r \sin \theta \\ \sin \theta & r \cos \theta \end{bmatrix}$
 $\det D\varphi_{r\theta} = r$



$\int f(x) dx = \int f(\varphi(u)) |det D\varphi_{r\theta}| dr d\theta$

$\Rightarrow I^2 = \frac{1}{2\pi} \iint_{-\infty}^{\infty} e^{-\frac{1}{2}(y^2+t^2)} dy dt = \frac{1}{2\pi} \int_{r=0}^{\infty} \int_{\theta=0}^{2\pi} e^{-\frac{1}{2}r^2} r dr d\theta = \frac{1}{2\pi} \left[2\pi \int_{r=0}^{\infty} e^{-\frac{1}{2}r^2} r dr \right]$

Now $u := -\frac{r^2}{2}$ $du = -r dr$
 $r=0 \Rightarrow u=0$
 $r=\infty \Rightarrow u = -\frac{1}{2}\infty^2 = -\infty$

$= -\int_{u=0}^{-\infty} e^u du = \int_{u=-\infty}^0 e^u du = e^u \Big|_{-\infty}^0 = 1 - 0 = 1$

$I^2 = 1$

$I = \pm 1$ but its +1 because curve is above x axis, so pos area. \square

Thm a linear combination of Normal RVs is a Normal RV:
 P.156 $\left. \begin{matrix} X \sim \mathcal{N}(\mu, \sigma) \\ Y := aX + b, a > 0 \end{matrix} \right\} \Rightarrow Y \sim \mathcal{N}(a\mu + b, a\sigma)$

Pf. $F_Y(r) = P(Y \leq r) = P(aX + b \leq r) = P(X \leq \frac{r-b}{a})$ because $aX + b \leq r \Leftrightarrow aX \leq r - b$

$= F_X\left(\frac{r-b}{a}\right) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\frac{r-b}{a}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} dx$

make COV $y = ax + b$ $dy = a dx$
 $\Rightarrow \frac{y-b}{a} = x$ $x = \frac{r-b}{a} \Rightarrow y = r$

$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^r e^{-\frac{1}{2}\left(\frac{y-b}{a\sigma}\right)^2} \frac{1}{a} dy$

$= \frac{1}{(a\sigma)\sqrt{2\pi}} \int_{-\infty}^r e^{-\frac{1}{2}\left(\frac{y-(a\mu+b)}{a\sigma}\right)^2} dy$

This is the form for $\mathcal{N}(a\mu + b, a\sigma)$ \square

Normalizing a Normal RV

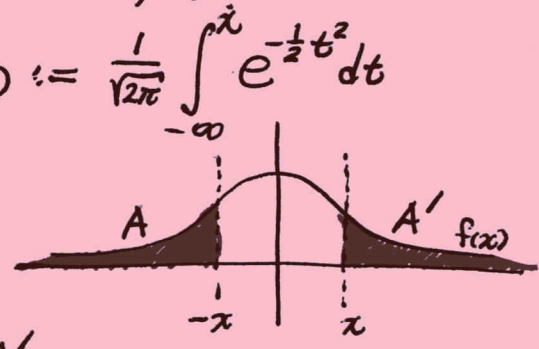
$a > 0$

Given $X \sim \mathcal{N}(\mu, \sigma)$ we want to define $Z := aX + b$ so $Z \sim \mathcal{N}(0, 1)$

we just showed $aX + b \sim \mathcal{N}(a\mu + b, a\sigma)$ so take $a := \frac{1}{\sigma}$ $b := -\frac{\mu}{\sigma}$

$\Rightarrow Z := \frac{1}{\sigma}X - \frac{\mu}{\sigma} = \frac{X - \mu}{\sigma}$ is distributed $\mathcal{N}(0, 1)$

The CDF for the ^{std} Normal distrib is $\Phi(x) := \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}t^2} dt$



Lemma Symmetry of Φ : $\Phi(-x) = 1 - \Phi(x)$

pf Since f has reflection symm about y axis, the shaded area under the curve, A , reflects to A'

$\Rightarrow \underbrace{\mathbb{P}(A)}_{\Phi(-x)} = \underbrace{\mathbb{P}(A')}_{1 - \Phi(x)}$ because $\Phi(x) + \mathbb{P}(A') \stackrel{!}{=} 1$ \square

p.159 example 3a $X \sim \mathcal{N}(\mu, \sigma)$
 $(3, 3)$

(i) Find $P(2 < X < 5)$

Normalize to $\mathcal{N}(0, 1)$ so we can use table on p.158 $Z := \frac{X - \mu}{\sigma}$

$2 < X < 5 \Rightarrow \frac{2-3}{3} < \frac{X-3}{3} < \frac{5-3}{3} \Rightarrow -\frac{1}{3} < Z < \frac{2}{3}$

$P(-\frac{1}{3} < Z < \frac{2}{3}) = \Phi(\frac{2}{3}) - \Phi(-\frac{1}{3})$
 $= \Phi(\frac{2}{3}) - [1 - \Phi(\frac{1}{3})]$ Symm Lemma
 $= \underbrace{\Phi(0.667)}_{0.7486} - [1 - \underbrace{\Phi(0.333)}_{0.6293}] = 0.3779 \checkmark$

(ii) $P(X > 0)$

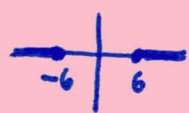
$0 < X \Rightarrow \frac{0-3}{3} < \frac{X-3}{3} \Rightarrow -1 < Z$



$P(-1 < Z) = 1 - P(Z \leq -1) = 1 - \Phi(-1) \stackrel{\text{Symm lemma}}{=} \Phi(1) = 0.8413$

(iii) $P(|X-3| > 6)$

$6 < |x-3|$
 $6 < |u|$



$u < -6$ or $6 < u$
 $(-\infty, -6) \cup (6, \infty)$

So here we get $X-3 < -6$ and $6 < X-3$

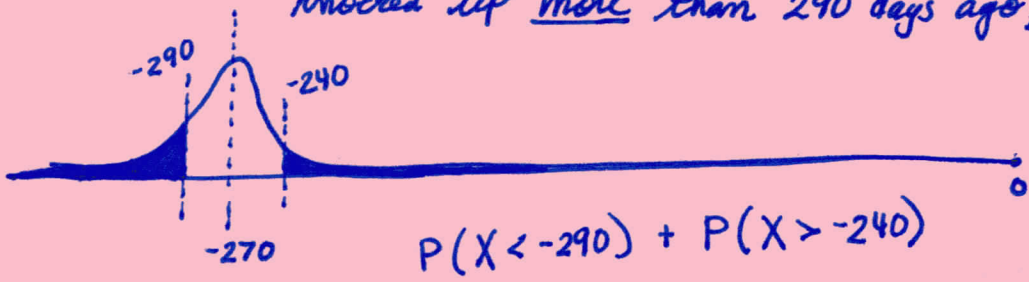
$\frac{X-3}{3} < \frac{-6}{3} = -2$ and $2 < \frac{X-3}{3}$
 $Z < -2$ and $2 < Z$

$P(Z < -2) + P(Z > 2)$
 $\Phi(-2) + 1 - P(Z \leq 2)$
 $1 - \Phi(2) + 1 - \Phi(2)$
 $2(1 - \Phi(2))$

P.160 example 3c

The length of a pregnancy is a Normal RV with $\mu = 270$ days and $\sigma = 10$

a woman gives birth today. what is the prob she was knocked up more than 290 days ago, or less than 240 days ago?



$$P(X < -290) + P(X > -240)$$

$$\frac{X - 270}{10} < \frac{-290 + 270}{10} \qquad \frac{X - 270}{10} > \frac{-240 + 270}{10}$$

$$Z < -2 \qquad Z > 3$$

$$P(Z < -2) + P(Z > 3)$$

$$= \Phi(-2) + 1 - \Phi(3) = 1 - \Phi(2) + 1 - \Phi(3) = 0.0241$$

P.161 Here are some bounds for Φ that the author said were important:

$$\frac{1}{\sqrt{2\pi}} \left(\frac{1}{x} - \frac{1}{x^3} \right) e^{-\frac{1}{2}x^2} < 1 - \Phi(x) < \frac{1}{\sqrt{2\pi}} \frac{1}{x} e^{-\frac{1}{2}x^2} \quad \forall x > 0$$

▷ Special Case of Central Lim Thm - DeMoivre-Laplace limit thm

$$S_n = \begin{matrix} \# \text{ of wins in } n \text{ trials} \\ \text{Prob } p \end{matrix} \qquad P\left(a \leq \frac{S_n - np}{\sqrt{np(1-p)}} \leq b\right) \xrightarrow{n \rightarrow \infty} \Phi(b) - \Phi(a)$$

We can approximate Binomial $\text{Bin}(n, p)$:

- with Poisson when n large np moderate
- with Normal $np(1-p)$ large (rule of thumb: $np(1-p) \geq 10$)

Here is a statement of CLT:

Seq of RVs (X_i) i.i.d.
(each has mean μ and var σ^2)

$$P\left(\frac{\sum X_i - n\mu}{\sigma\sqrt{n}} \leq a\right) \xrightarrow{n \rightarrow \infty} \underbrace{\frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-\frac{1}{2}x^2} dx}_{\Phi(a)} \quad \text{p.157}$$

example 3e

Flip coin $n=40$ times
 $p = \frac{1}{2}$ $X = \#$ of heads i.e. wins
we take $\mu = np = 40 \cdot \frac{1}{2} = 20$

$$\sigma := \sqrt{n(p)(1-p)} = \sqrt{40 \cdot \frac{1}{2} \cdot \frac{1}{2}} = \sqrt{10}$$

$$\begin{aligned} \text{Normal approx } P(X=20) &= P(19.5 < X < 20.5) \\ &= P\left(\frac{19.5-20}{\sqrt{10}} < Z < \frac{20.5-20}{\sqrt{10}}\right) \\ &= P(-0.16 < Z < 0.16) \\ &= \Phi(0.16) - \Phi(-0.16) \\ &= \Phi(0.16) - [1 - \Phi(0.16)] = 0.1272 \end{aligned}$$

□