

Lec 19 MATH 340/640

Let $Y|X=x \sim \text{Bin}(n, x)$, $X \sim \text{Beta}(\alpha, \beta)$. Full p.f. Y is "conjugate" distr. $S_X = \mathbb{H}_p$

$$p(y) = \int_{\mathbb{R}} P_{Y|X}(y|x) f(x) dx = \int_{\mathbb{R}} \binom{n}{y} x^y (1-x)^{n-y} \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} \mathbb{1}_{x \in (0,1)} dx$$

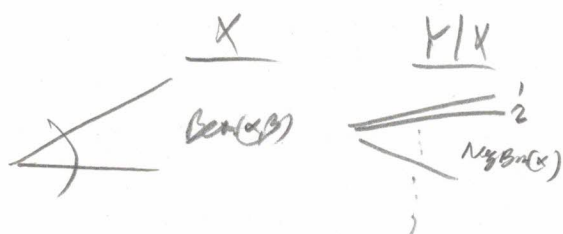
\swarrow \nwarrow
 $\text{Beta}(\alpha, \beta)$ $\text{Bin}(n, x)$

$$= \frac{\binom{n}{y}}{B(\alpha, \beta)} \int_0^1 x^{y+\alpha-1} (1-x)^{n-y+\beta-1} dx = \frac{\binom{n}{y} B(y+\alpha, n-y+\beta)}{B(\alpha, \beta)} = \text{Beta Binomial}(n, \alpha, \beta)$$

take the place of p

Let $Y|X=x \sim \text{ExpNegBin}(k, x)$, $X \sim \text{Beta}(\alpha, \beta)$. Y is a conjugate distr. Full p.f. $S_X = \mathbb{H}_p$

$$p(y) = \int_{\mathbb{R}} P_{Y|X}(y|x) f(x) dx = \int_{\mathbb{R}} \frac{\Gamma(k+y)}{\Gamma(k) y!} (1-x)^y x^k \mathbb{1}_{y \in \mathbb{N}_0} \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} \mathbb{1}_{x \in (0,1)} dx$$



$$= \frac{\Gamma(k+y)}{\Gamma(k) y! B(\alpha, \beta)} \mathbb{1}_{y \in \mathbb{N}_0} \int_0^1 x^{k+\alpha-1} (1-x)^{y+\beta-1} dx = \frac{\Gamma(k+y) B(k+\alpha, y+\beta)}{\Gamma(k) y! B(\alpha, \beta)} \mathbb{1}_{y \in \mathbb{N}_0}$$

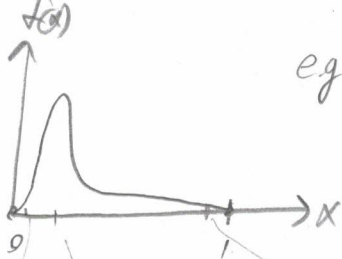
$$= \text{Beta Negative Binomial}(k, \alpha, \beta)$$

take the place of p

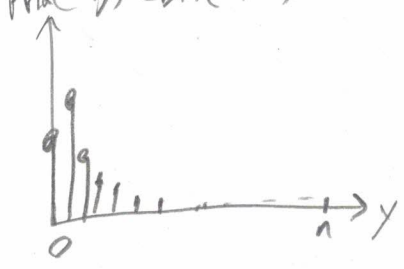
Let $Y|X=x \sim \text{Exp}(\alpha)$, $X \sim \text{Gamma}(\alpha, \beta)$. Y is a conjugate distr. $S_X = \mathbb{H}_\lambda$

Let $Y \sim \text{Lomax}(\beta, \alpha)$, shorter waiting time distr.

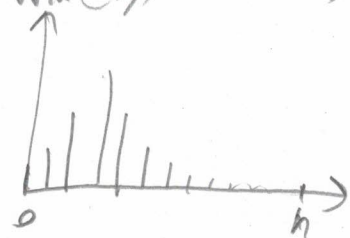
eg $\text{Beta}(2, 8)$



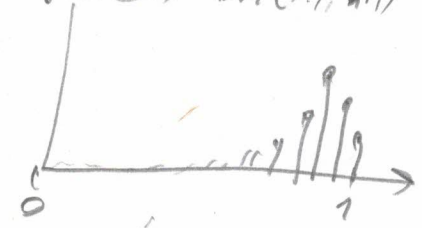
$P_{Y|X}(0.1, y) = \text{Bin}(n, 0.1)$



$P_{Y|X}(0.2, y) = \text{Bin}(n, 0.2)$

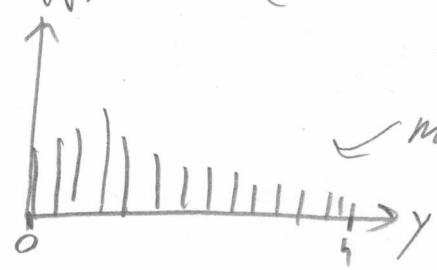


$P_{Y|X}(0.9, y) = \text{Bin}(n, 0.9)$



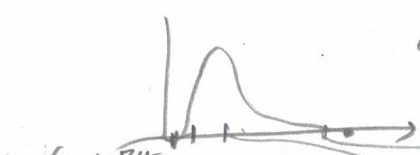
$\text{Var}[Y|X=0.9] = n \cdot 0.9 \cdot 0.1$

$P_Y(y) = \text{BetaBinom}(n, 2, 8)$

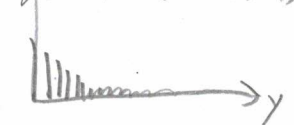


Compound distr
 more dispersed variance larger than
 a Binomial rv

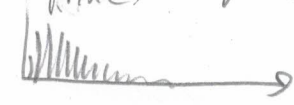
eg $\text{Beta}(2, 8)$



$P_{Y|X}(0.1, y) = \text{NegBin}(k, 0.1)$



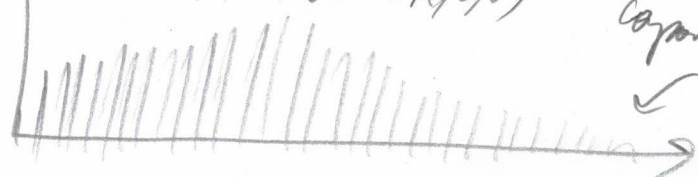
$P_{Y|X}(0.2, y) = \text{NegBin}(k, 0.2)$



$P_{Y|X}(0.9, y) = \text{NegBin}(0.9, k)$



$P_Y(y) = \text{BetaNegBin}(k, 2, 8)$



Compound distr
 more dispersed, variance
 larger than Neg Binomial

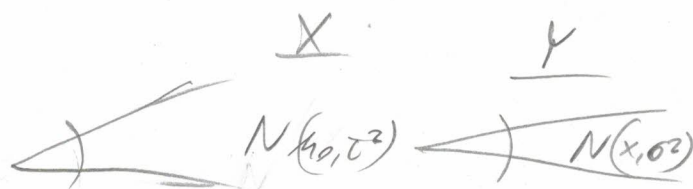
let's prove a very useful kernel... let $a \in \mathbb{R}, b > 0$

$$\begin{aligned}
 X \sim N\left(\frac{a}{2b}, \frac{1}{2b}\right) &= \frac{1}{\sqrt{2\pi\left(\frac{1}{2b}\right)}} e^{-\frac{1}{2\left(\frac{1}{2b}\right)}\left(x - \frac{a}{2b}\right)^2} \\
 &= \sqrt{\frac{b}{\pi}} e^{-b\left(x - \frac{a}{2b}\right)^2} \\
 &= \sqrt{\frac{b}{\pi}} e^{-b\left(x^2 - \frac{a}{b}x + \frac{a^2}{4b^2}\right)} \\
 &= \sqrt{\frac{b}{\pi}} e^{-bx^2 + ax - \frac{a^2}{4b}} \\
 &= \sqrt{\frac{b}{\pi}} e^{-\frac{a^2}{4b}} e^{ax - bx^2} \propto e^{ax - bx^2} \propto N\left(\frac{a}{2b}, \frac{1}{2b}\right)
 \end{aligned}$$

What is

$$\int_{\mathbb{R}} e^{ax - bx^2} dx = \sqrt{\frac{\pi}{b}} e^{\frac{a^2}{4b}} \int_{\mathbb{R}} \sqrt{\frac{b}{\pi}} e^{-\frac{a^2}{4b}} e^{ax - bx^2} dx = \sqrt{\frac{\pi}{b}} e^{\frac{a^2}{4b}} \quad \text{cool result!}$$

Let $Y|X=x \sim N(x, \sigma^2)$, $X \sim N(\mu_0, \tau^2)$ where $\mu_0 \in \mathbb{R}$, $\tau^2 > 0$, $\sigma^2 > 0$
 Y is a conditional cont. distr



$$f_Y(y) = \int_{\mathbb{R}} f_{Y|X}(x, y) f_X(x) dx$$

$$= \int_{\mathbb{R}} \left(\frac{1}{\sqrt{2\pi}\sigma^2} e^{-\frac{1}{2\sigma^2}(y-x)^2} \right) \left(\frac{1}{\sqrt{2\pi}\tau^2} e^{-\frac{1}{2\tau^2}(x-\mu_0)^2} \right) dx$$

$$\propto \int_{\mathbb{R}} e^{-\frac{1}{2\sigma^2}(y-x)^2 - \frac{1}{2\tau^2}(x-\mu_0)^2} dx$$

$$= \int_{\mathbb{R}} e^{-\frac{1}{2\sigma^2}y^2 + \frac{y}{\sigma^2}x - \frac{1}{2\sigma^2}x^2 - \frac{1}{2\tau^2}x^2 + \frac{\mu_0}{\tau^2}x - \frac{\mu_0^2}{2\tau^2}} dx$$

$$\propto e^{-\frac{1}{2\sigma^2}y^2} \int_{\mathbb{R}} e^{\underbrace{\left(\frac{y}{\sigma^2} + \frac{\mu_0}{\tau^2}\right)x}_{a} - \underbrace{\left(\frac{1}{2\sigma^2} + \frac{1}{2\tau^2}\right)x^2}_{b}} dx$$

$$b = \frac{1}{2} \left(\frac{1}{\sigma^2} + \frac{1}{\tau^2} \right) = \frac{1}{2} \frac{\sigma^2 + \tau^2}{\sigma^2 \tau^2}$$

$$= e^{-\frac{1}{2\sigma^2}y^2} \sqrt{\frac{\pi}{b}} e^{\frac{a^2}{4b}}$$

$$\frac{a^2}{4b} = \frac{\frac{y^2}{\sigma^4} + \frac{2\mu_0 y}{\sigma^2 \tau^2} + \frac{\mu_0^2}{\tau^4}}{\frac{2(\sigma^2 + \tau^2)}{\sigma^2 \tau^2}} = \frac{y^2 \tau^2}{2\sigma^2(\sigma^2 + \tau^2)} + \frac{\mu_0 y}{\sigma^2 + \tau^2} + C$$

$$\propto e^{-\frac{1}{2\sigma^2}y^2} e^{\frac{a^2}{4b}}$$

$$\propto e^{-\frac{1}{2\sigma^2}y^2 + \frac{\tau^2}{2\sigma^2(\sigma^2 + \tau^2)}y^2 + \frac{\mu_0}{\sigma^2 + \tau^2}y}$$

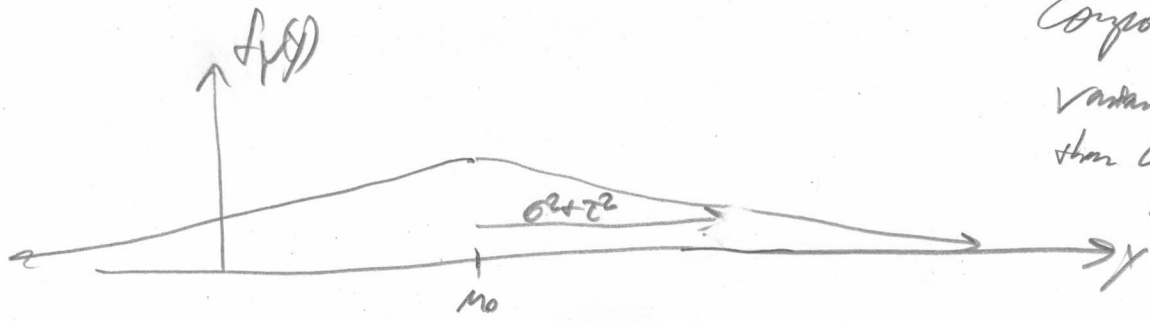
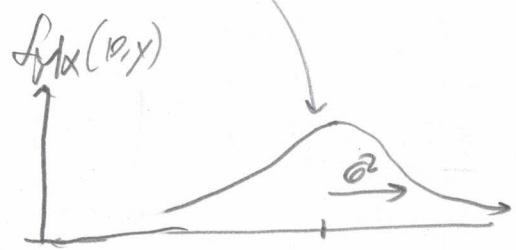
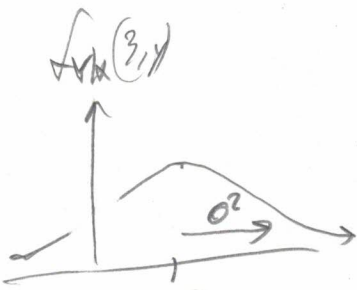
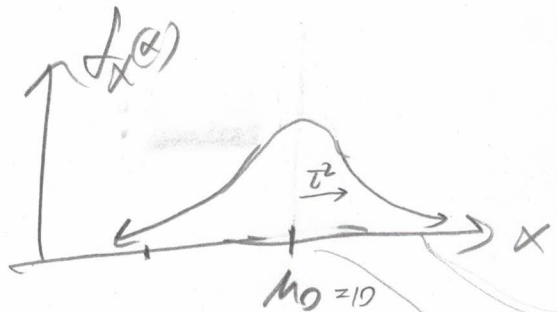
$$= e^{\underbrace{\frac{\mu_0}{\sigma^2 + \tau^2}y}_{c} - \underbrace{\left(\frac{\tau^2}{2\sigma^2(\sigma^2 + \tau^2)} + \frac{1}{2\sigma^2}\right)y^2}_{d}}$$

$$2d = \frac{-\tau^2}{\sigma^2(\sigma^2 + \tau^2)} + \frac{1}{\sigma^2} = \frac{\sigma^2 + \tau^2 - \tau^2}{\sigma^2(\sigma^2 + \tau^2)} = \frac{\sigma^2}{\sigma^2(\sigma^2 + \tau^2)} = \frac{1}{\sigma^2 + \tau^2}$$

$$\Rightarrow \frac{1}{2d} = \sigma^2 + \tau^2$$

$$\Rightarrow \frac{c}{2d} = \frac{\mu_0}{\sigma^2 + \tau^2} (\sigma^2 + \tau^2) = \mu_0$$

$$\propto N\left(\frac{c}{2d}, \frac{1}{2d}\right) = \boxed{N(\mu_0, \sigma^2 + \tau^2)}$$



Compound dist
variance larger
than compo-
nents

let's do the Poisson now!

Then if $\lim_{n \rightarrow \infty} P_{X_n}(x) = P_X(x) \Rightarrow X_n \xrightarrow{d} X$. Proof:

$$\sum_{x \in (-\infty, y]} \lim_{n \rightarrow \infty} P_{X_n}(x) = \sum_{x \in (-\infty, y]} P_X(x) \xRightarrow{\text{interchange sum/limits}} \lim_{n \rightarrow \infty} \sum_{x \in (-\infty, y]} P_{X_n}(x) = \sum_{x \in (-\infty, y]} P_X(x)$$

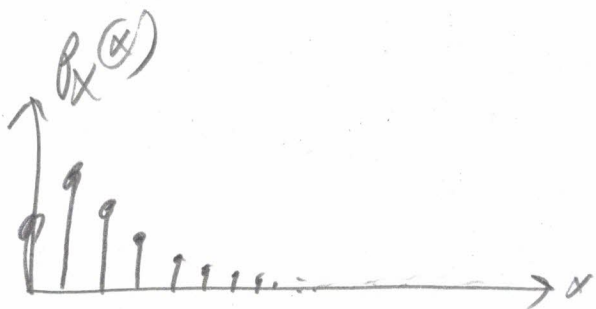
$\Rightarrow \lim_{n \rightarrow \infty} F_{X_n}(x) = F_X(x)$, i.e. the def of convergence is done.

let $X_n \sim \text{Bin}(n, \frac{\lambda}{n})$ where $\lambda := np$. Consider n large, p small just like when we derived the exponential from the geometric. Eg: call center with tons of customers and prob they call is very small.

$$\begin{aligned} \lim_{n \rightarrow \infty} P_{X_n}(x) &= \lim_{n \rightarrow \infty} \binom{n}{x} \left(\frac{\lambda}{n}\right)^x \left(1 - \frac{\lambda}{n}\right)^{n-x} = \frac{\lambda^x}{x!} \lim_{n \rightarrow \infty} \frac{n!}{(n-x)!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \\ &= \frac{\lambda^x e^{-\lambda}}{x!} \lim_{n \rightarrow \infty} \frac{n(n-1)(n-2)\dots(1)}{n \cdot n \cdot n \dots (n-x)(n-x-1)\dots(1)} \lim_{n \rightarrow \infty} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-x} \end{aligned}$$

$$\Rightarrow \text{Poisson}(\lambda) = \frac{\lambda^x e^{-\lambda}}{x!} \mathbb{1}_{x \in \mathbb{N}_0}$$

Param space: $n \in \mathbb{N}, \lambda \in (0, \infty)$



Looks like Binomial except no max $\{x\}$

let $Y|X=x \sim \text{Poisson}(x)$, $X \sim \text{Gamma}(\alpha, \beta)$

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$$P_Y(y) = \int_{\mathbb{R}} P_{Y|X}(x|y) f_X(x) dx = \int_{\mathbb{R}} \frac{x^y e^{-x}}{y!} \mathbb{1}_{y \in \mathbb{N}_0} \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \mathbb{1}_{x \in (0, \infty)} dx$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \frac{1}{y!} \mathbb{1}_{y \in \mathbb{N}_0} \int_0^\infty x^{y+\alpha-1} e^{-(\beta+1)x} dx$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \frac{1}{y!} \mathbb{1}_{y \in \mathbb{N}_0} \frac{\Gamma(y+\alpha)}{(\beta+1)^{y+\alpha}}$$

$$1-p = 1 - \frac{\beta}{\beta+1} = \frac{\beta+1-\beta}{\beta+1}$$

\uparrow

$$\text{let } p = \frac{\beta}{\beta+1} \Rightarrow 1-p = \frac{1}{\beta+1}, \text{ let } k = \alpha$$

$$= \frac{\Gamma(y+\alpha)}{\Gamma(\alpha) y!} \left(\frac{\beta}{\beta+1}\right)^\alpha \left(\frac{1}{\beta+1}\right)^y \mathbb{1}_{y \in \mathbb{N}_0} = \frac{\Gamma(y+k)}{\Gamma(k) y!} (1-p)^y p^k \mathbb{1}_{y \in \mathbb{N}_0} = \text{ExtNegBin}(k, p)$$

