

Lecture 14 Math 621 10/26/17

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$$X \sim \text{Gamma}(\alpha, \beta) \quad Y|X=x \sim \text{Poisson}(x) \Rightarrow f_Y(y) =$$

$$= \frac{\Gamma(y+k)}{\Gamma(k)\Gamma(y+1)} p^k (1-p)^y = \text{Ext-Neg Bin}(p, k) \leftarrow \text{Support same}$$

If $k \in \mathbb{N}$

$$= \binom{y+k-1}{k-1} p^k (1-p)^y = \text{Neg Bin}(p, k)$$

two params

"more freedom"

AKA "overdispersion"

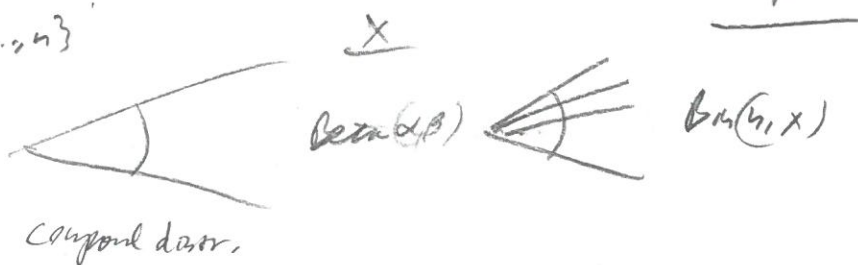
Poisson - count data

Neg bin - also count data but more overdispersed

Another Example of this.

$$Y|X \sim \text{Bin}(n, x), \quad X \sim \text{Beta}(\alpha, \beta)$$

$\text{supp}(Y) = \{0, \dots, n\}$
still!

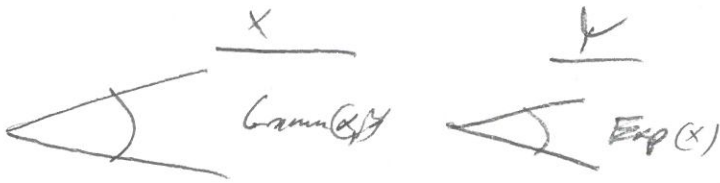


$$P_Y(y) = \int_{\text{supp}(x)} P_{Y|X}(y, x) f_X(x) dx = \int_0^1 \binom{n}{y} x^y (1-x)^{n-y} \frac{1}{B(\alpha, \beta)} x^{\alpha-1} (1-x)^{\beta-1} dx$$

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

"overdispersed binomial"

$$Y|X \sim \text{Exp}(x), \quad X \sim \text{Gamma}(\alpha, \beta) \quad f_{\text{Exp}}(y) = (0, \infty) \text{ still}$$



$$f_Y(y) = \int_{f_{\text{Exp}}(x)} f_{Y|X}(y|x) f_X(x) dx = \int_0^{\infty} x e^{-xy} \frac{\beta^\alpha x^{\alpha-1} e^{-\beta x}}{\Gamma(\alpha)} dx$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)} \int_0^{\infty} x^\alpha e^{-x(\beta+y)} dx = \frac{\beta^\alpha}{\Gamma(\alpha)} \int_0^{\infty} \frac{u^\alpha}{(\beta+y)^\alpha} e^{-u} \frac{1}{\beta+y} du$$

$$\text{let } u = x(\beta+y) \quad \frac{du}{dx} = \beta+y \Rightarrow dx = \frac{1}{\beta+y} du \\ \Rightarrow x = \frac{1}{\beta+y} u$$

$$= \frac{\beta^\alpha}{\Gamma(\alpha)(\beta+y)^{\alpha+1}} \int_0^{\infty} u^{\alpha+1-1} e^{-u} du = \frac{\beta^\alpha \overset{\alpha!}{\Gamma(\alpha+1)}}{\Gamma(\alpha)(\beta+y)^{\alpha+1}} = \frac{\alpha}{\beta} \left(\frac{\beta}{\beta+y}\right)^{\alpha+1}$$

\$(\alpha-1)!\$

$$= \frac{\alpha}{\beta} \left(1 + \frac{y}{\beta}\right)^{-(\alpha+1)} = \text{Lomax}(\beta, \alpha)$$

Another type of survival distr.
"overdispersed exponential"

\$\Rightarrow\$ New r.v.'s can be created via mixes and compounds

NEW TOPIC

Recall Complex #'s

field of complex #'s

$$a, b \in \mathbb{R} \quad z := a + bi \in \mathbb{C} \quad \text{where } i := \sqrt{-1} \Rightarrow i^2 = -1, i^3 = -i, i^4 = 1$$

$$\operatorname{Re}[z] = a, \quad \operatorname{Im}[z] = b$$

$$\text{Recall } e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!} = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \frac{x^4}{4!} + \frac{x^5}{5!} + \dots$$

$$\Rightarrow e^{itx} = \sum_{k=0}^{\infty} \frac{(itx)^k}{k!} = 1 + itx + \frac{(itx)^2}{2!} + \frac{(itx)^3}{3!} + \frac{(itx)^4}{4!} + \frac{(itx)^5}{5!} + \dots$$

$$= 1 + itx - \frac{t^2 x^2}{2!} - \frac{it^3 x^3}{3!} + \frac{t^4 x^4}{4!} + \frac{it^5 x^5}{5!} + \dots$$

$$\sin(x) = x - \frac{x^3}{3!} + \frac{x^5}{5!} - \dots$$

$$\cos(x) = 1 - \frac{x^2}{2!} + \frac{x^4}{4!} - \dots$$

pattern: i is in the odd terms and not in the even terms
2 terms $+$, 2 terms $-$, etc...

$$i \sin(tx) = itx - i \frac{t^3 x^3}{3!} + i \frac{t^5 x^5}{5!} - \dots$$

$$\cos(tx) = 1 - \frac{t^2 x^2}{2!} + \frac{t^4 x^4}{4!} - \dots$$

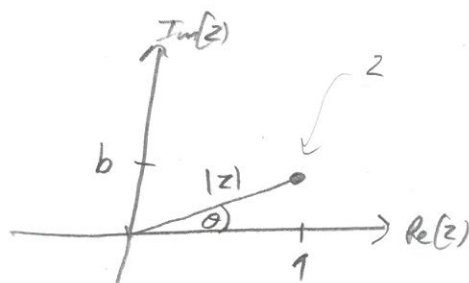
$$\Rightarrow e^{itx} = \cos(tx) + i \sin(tx)$$

$$\text{Note: if } tx = \pi \Rightarrow e^{i\pi} = -1 + 0 \Rightarrow e^{i\pi} + 1 = 0$$

$$\Rightarrow e^{i\theta} = \cos(\theta) + i \sin(\theta)$$

Euler's identity

If you imagine



$$z = |z| e^{i\theta} \quad \text{where } |z| = \sqrt{a^2 + b^2} \text{ is the "complex norm"}$$

and θ is the complex argument

$$\operatorname{Arg}(z) := \theta = \arctan\left(\frac{b}{a}\right) \in [-\pi, \pi]$$

Recall Complex #'s

$$a, b \in \mathbb{R}$$

$$z = a + bi \in \mathbb{C}$$

$$\text{where } i := \sqrt{-1} \Rightarrow i^2 = -1$$

$$i^3 = -\sqrt{-1} = -i, i^4 = 1$$

$$\text{Re}(z) := a, \text{Im}(z) := b$$

Define $L^1 := \left\{ f: \int_{\mathbb{R}} |f(x)| dx < \infty \right\}$ " L^1 integrable" or absolutely integrable

are all PDF's $\in L^1$? YES!

are all $f \in L^1$ PDF's? NO... But... they can be scaled to be

If $f \in L^1$ then \hat{f} known as the "Fourier transform of f " defined by

$$\hat{f}(t) = \int_{\mathbb{R}} e^{-2\pi i t x} f(x) dx$$

\Rightarrow (not guaranteed!)

If $\hat{f} \in L^1$, then the Fourier transform can be inverted back to f via:

$$f(x) = \int_{\mathbb{R}} e^{2\pi i t x} \hat{f}(t) dt$$

$f(x)$ is called the "time domain", $\hat{f}(t)$ is called the frequency domain

Why? $f(x)$ can be decomposed into a sum of sines and cosines called a "Fourier Series".

$\text{Re}[\hat{f}(t)]$ gives the amplitude of a frequency t and
 $\text{Arg}[\hat{f}(t)]$ " " " " phase shift of the " " " ".

$$\text{let } \phi(t) = \hat{f}\left(-\frac{t}{2\pi}\right) = \int_{\mathbb{R}} e^{itx} f(x) dx$$

Inversion ... if $\hat{\phi}(t) \in L^1$ then ...

$$\begin{aligned} f(x) &= \int_{\mathbb{R}} e^{2\pi i t x} \hat{f}(t) dt & \text{let } u = -2\pi t \Rightarrow \frac{du}{dt} = -2\pi \Rightarrow du = -\frac{1}{2\pi} du \\ & & \Rightarrow t = -\frac{u}{2\pi} \\ & & u = -\infty \\ & & u = \infty \\ &= \int_{-\infty}^{\infty} e^{2\pi i \left(-\frac{1}{2\pi} u\right) x} \underbrace{\hat{f}\left(-\frac{u}{2\pi}\right)}_{\phi(u)} du = - \int_{\mathbb{R}} e^{-i u x} \phi(u) \left(-\frac{1}{2\pi}\right) du = \frac{1}{2\pi} \int_{\mathbb{R}} e^{-i u x} \phi(u) du \end{aligned}$$

$$\phi_X(t) = E[e^{itX}] = \int_{\mathbb{R}} e^{itx} f(x) dx \text{ if } f \text{ is a PDF of r.v. } X$$

Also

$$\phi_X(t) = E[e^{itX}] = \sum_{x \in \mathcal{X}(X)} e^{itx} p(x) \text{ if } p \text{ is a PMF of r.v. } X$$

technically,
Lebesgue integral

$\phi_X(t)$ is called the characteristic function of r.v. X (Ch. f.)

Let's look properties of ch.f.'s ① $\phi(0) = 1$ since $E[e^{i(0)X}] = E(1) = 1$

(6)

Let say X_1, X_2 i.i.d. $Y = X_1 + X_2 \sim$ convolution

$$\textcircled{2} \phi_{X_1+X_2}(t) = E[e^{it(X_1+X_2)}] = E[e^{itX_1} e^{itX_2}] = E[e^{itX_1}] E[e^{itX_2}] \text{ by indep.} \\ = \phi_{X_1}(t) \phi_{X_2}(t)$$

Let $Y = aX + b$

$$\textcircled{3} \phi_Y(t) = E[e^{itY}] = E[e^{it(aX+b)}] = E[e^{i\overset{t'}{\uparrow} aX} e^{itb}] = e^{itb} \phi_X(at)$$

④ $\phi_X(t)$ is bounded by 1 and \Rightarrow it always exists

$$|\phi_X(t)| = |E[e^{itX}]] = \left| \int_{\mathbb{R}} e^{itx} f(x) dx \right| \leq \int_{\mathbb{R}} |e^{itx} f(x)| dx \leq \int_{\mathbb{R}} |e^{itx}| |f(x)| dx$$

$$= \int_{\mathbb{R}} |\cos(tx) + i \sin(tx)| |f(x)| dx$$

\Downarrow

$$= \int_{\mathbb{R}} (\underbrace{\cos^2(tx)}_{\leq 1} + \underbrace{\sin^2(tx)}_{\leq 1}) |f(x)| dx$$

since if $f(x)$ is PDF its always ≥ 0

$$= \int_{\mathbb{R}} |f(x)| dx = \int_{\mathbb{R}} f(x) dx = 1 \checkmark$$

Define $m_X(t) = \phi\left(\frac{t}{i}\right) = E[e^{tX}]$ the moment gen. function

is not guaranteed to exist! ch.f. is more powerful...

$$\textcircled{5} E(X^k) = \phi_X^{(k)}(0) \text{ (it can generate moments)}$$