Intro Prob Lecture Notes

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Conditional Distributions

- Suppose X, Y are jointly continuous
- Define $f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)}$ assuming $f_X(x) > 0$.
 - Conditional pdf of Y given X = x
- Notice that

$$f(x,y) = f_{X|Y}(x|y)f_Y(y) = f_{Y|X}(y|x)f_X(x)$$

which will later allow us to construct joint pdfs via conditional pdfs

• Ex: Suppose X, Y joint pdf

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$$f(x,y) = \begin{cases} x+y & \text{for } 0 < x < 1, 0 < y < 1 \\ 0 \text{ otherwise} \end{cases}$$

- Suppose we are told $Y = \frac{3}{4}$. $P(X \le \frac{1}{2}|Y = \frac{3}{4}) = ?$
 - * Note: We can't do (*) $\frac{P(X \leq \frac{1}{2}|Y = \frac{3}{4})}{P(Y = \frac{3}{4})}$ because the denominator is 0.
- Compute

$$f_{X|Y}(x|\frac{3}{4}) = \frac{f(x,\frac{3}{4})}{f_Y(\frac{3}{4})}$$
$$= \frac{x + \frac{3}{4}}{\frac{1}{2} + \frac{3}{4}}$$
$$= \frac{x + \frac{3}{4}}{\frac{5}{4}}$$
$$= \frac{4}{5}x + \frac{3}{5}$$

for
$$0 < x < 1$$

• Ex:
$$f_Y(y) = \int_0^1 (x+y)dx = \frac{x^2}{2} + xy\Big|_0^1 = \frac{1}{2} + y$$
 for $0 < y < 1$

$$-P(X \le \frac{1}{2}|Y = \frac{3}{4}) = \int_{0}^{\frac{1}{2}} f_{X|Y}(x|\frac{3}{4})dx = \dots = \frac{4}{10}$$

* This technique needed when conditioning P=0. If asked $P(X \leq \frac{1}{2}|Y>\frac{3}{4})$, we can use above (*)

– Check if
$$P(X \leq \frac{1}{2}|Y > \frac{3}{4}) = \frac{P(X \leq \frac{1}{2}|Y > \frac{3}{4})}{P(Y > \frac{3}{4})}$$

* In general,
$$\neq \int_{\frac{3}{4}}^{1} \int_{0}^{\frac{1}{2}} f_{X|Y}(x|y) dx dy$$

· Try it!

- Ex: Compound Poisson. N claims made. N \sim Poisson(λ)
 - X_i payout for *i*th claim, i = 1, ..., n.

$$-T = \sum_{i=1}^{N} X_i, X_i \sim \text{independent } \exp(\frac{1}{\beta}), T | N = n \sim \text{Gamma}(n, \beta)$$

- * Incomplete discussion. Later.
- Ex: $X \sim \exp(\Lambda)$ and suppose Λ is another random variable

$$-\Lambda \sim \text{Gamma}(\alpha, \beta)$$

$$-(X|\Lambda=\lambda)\sim \exp(\lambda)$$

$$-f_{X|\Lambda}(x|\lambda) = \lambda e^{-\lambda x}$$
 for $x > 0$

$$- f_{\Lambda}(\lambda) = \frac{\lambda^{\alpha - 1} e^{-\frac{\lambda}{\beta}}}{\beta^{\alpha} \Gamma(\alpha)}$$

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$$\begin{split} f_{X,\Lambda}(x,\lambda) &= [f_{X|\Lambda}(x|\lambda)] \times [f_{\Lambda}(\lambda)] \\ &= \frac{\lambda^{\alpha} e^{-\lambda(x+\frac{1}{\beta})}}{\beta^{\alpha}\Gamma(\alpha)} \end{split}$$

for $x > 0, \lambda > 0$

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$$f_X(x) = \int_0^\infty \frac{\lambda^{\alpha} e^{-\lambda(x+\frac{1}{\beta})}}{\beta^{\alpha} \Gamma(\alpha)} d\lambda$$
$$= \frac{1}{\beta^{\alpha} \Gamma(\alpha)} \int_0^\infty x^{(\alpha+1)-1} e^{-\lambda/(x+\frac{1}{\beta})^{-1}} d\lambda$$

Integral is a gamma distribution

$$= \frac{\alpha}{\beta^{\alpha}(x + \frac{1}{\beta})^{\alpha+1}} \text{ for } x > 0 \sim \text{ Pareto Distribution}$$
$$= \frac{\alpha\beta}{(1 + \beta x)^{\alpha+1}}$$

- Ex: Bayes' Statistics
 - $-X|p \sim \text{Binomial}(n,p)$
 - $p \sim$ Uniform (0, 1) \sim "prior distribution"

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$$P_{X,p}(x,p) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} \cdot 1 \text{ for } x = 0, 1, 2, \dots, n, 0$$

- Exercises: (x = 0, 1, 2, ... n)
 - * $P_X(x) \sim \text{discrete uniform} = \frac{1}{n+1} x = 0, 1, \dots n+1$
 - * $P|X \sim \mathrm{Beta}(x+1,n-x+1) \sim$ "Posterior distribution"