### 18.440: Lecture 21

## More continuous random variables

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## Outline

## Gamma distribution

## Cauchy distribution

Beta distribution

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## Defining gamma function $\Gamma$

- Last time we found that if $X$ is geometric with rate 1 and $n \geq 0$ then $E\left[X^{n}\right]=\int_{0}^{\infty} x^{n} e^{-x} d x=n!$.
- This expectation $E\left[X^{n}\right]$ is actually well defined whenever $n>-1$. Set $\alpha=n+1$. The following quantity is well defined for any $\alpha>0$ :
$\Gamma(\alpha):=E\left[X^{\alpha-1}\right]=\int_{0}^{\infty} x^{\alpha-1} e^{-x} d x=(\alpha-1)!$.
- So $\Gamma(\alpha)$ extends the function $(\alpha-1)$ ! (as defined for strictly positive integers $\alpha$ ) to the positive reals.
- Vexing notational issue: why define $\Gamma$ so that $\Gamma(\alpha)=(\alpha-1)$ ! instead of $\Gamma(\alpha)=\alpha!?$
- At least it's kind of convenient that $\Gamma$ is defined on $(0, \infty)$ instead of $(-1, \infty)$.


## Recall: geometric and negative binomials

- The sum $X$ of $n$ independent geometric random variables of parameter $p$ is negative binomial with parameter $(n, p)$.
- Waiting for the $n$th heads. What is $P\{X=k\}$ ?
- Answer: $\binom{k-1}{n-1} p^{n-1}(1-p)^{k-n} p$.
- What's the continuous (Poisson point process) version of "waiting for the $n$th event"?


## Poisson point process limit

- Recall that we can approximate a Poisson process of rate $\lambda$ by tossing $N$ coins per time unit and taking $p=\lambda / N$.
- Let's fix a rational number $x$ and try to figure out the probability that that the $n$th coin toss happens at time $x$ (i.e., on exactly $x N$ th trials, assuming $x N$ is an integer).
- Write $p=\lambda / N$ and $k=x N$. (Note $p=\lambda x / k$.)
- For large $N,\binom{k-1}{n-1} p^{n-1}(1-p)^{k-n} p$ is

$$
\begin{aligned}
& \frac{(k-1)(k-2) \ldots(k-n+1)}{(n-1)!} p^{n-1}(1-p)^{k-n} p \\
& \approx \frac{k^{n-1}}{(n-1)!} p^{n-1} e^{-x \lambda} p=\frac{1}{N}\left(\frac{(\lambda x)^{(n-1)} e^{-\lambda x} \lambda}{(n-1)!}\right)
\end{aligned}
$$

## Defining $\Gamma$ distribution

- The probability from previous side, $\frac{1}{N}\left(\frac{(\lambda x)^{(n-1)} e^{-\lambda x} \lambda}{(n-1)!}\right)$ suggests the form for a continuum random variable.
- Replace $n$ (generally integer valued) with $\alpha$ (which we will eventually allow be to be any real number).
- Say that random variable $X$ has gamma distribution with parameters $(\alpha, \lambda)$ if $f_{X}(x)=\left\{\begin{array}{ll}\frac{(\lambda x)^{\alpha-1} e^{-\lambda x} \lambda}{\Gamma(\alpha)} & x \geq 0 \\ 0 & x<0\end{array}\right.$.
- Waiting time interpretation makes sense only for integer $\alpha$, but distribution is defined for general positive $\alpha$.


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- A standard Cauchy random variable is a random real number with probability density $f(x)=\frac{1}{\pi} \frac{1}{1+x^{2}}$.
- There is a "spinning flashlight" interpretation. Put a flashlight at $(0,1)$, spin it to a uniformly random angle in $[-\pi / 2, \pi / 2]$, and consider point $X$ where light beam hits the $x$-axis.
- $F_{X}(x)=P\{X \leq x\}=P\{\tan \theta \leq x\}=P\left\{\theta \leq \tan ^{-1} x\right\}=$ $\frac{1}{2}+\frac{1}{\pi} \tan ^{-1} x$.
- Find $f_{X}(x)=\frac{d}{d x} F(x)=\frac{1}{\pi} \frac{1}{1+x^{2}}$.


## Cauchy distribution: Brownian motion interpretation

- The light beam travels in (randomly directed) straight line. There's a windier random path called Brownian motion.
- If you do a simple random walk on a grid and take the grid size to zero, then you get Brownian motion as a limit.
- We will not give a complete mathematical description of Brownian motion here, just one nice fact.
- FACT: start Brownian motion at point $(x, y)$ in the upper half plane. Probability it hits negative $x$-axis before positive $x$-axis is $\frac{1}{2}+\frac{1}{\pi} \tan ^{-1} \frac{y}{x}$. Linear function of angle between positive $x$-axis and line through $(0,0)$ and $(x, y)$.
- Start Brownian motion at $(0,1)$ and let $X$ be the location of the first point on the $x$-axis it hits. What's $P\{X<a\}$ ?
- Applying FACT, translation invariance, reflection symmetry: $P\{X<x\}=P\{X>-x\}=\frac{1}{2}+\frac{1}{\pi} \tan ^{-1} \frac{1}{x}$.
- So $X$ is a standard Cauchy random variable.


## Question: what if we start at $(0,2)$ ?

- Start at $(0,2)$. Let $Y$ be first point on $x$-axis hit by Brownian motion. Again, same probability distribution as point hit by flashlight trajectory.
- Flashlight point of view: $Y$ has the same law as $2 X$ where $X$ is standard Cauchy.
- Brownian point of view: $Y$ has same law as $X_{1}+X_{2}$ where $X_{1}$ and $X_{2}$ are standard Cauchy.
- But wait a minute. $\operatorname{Var}(Y)=4 \operatorname{Var}(X)$ and by independence $\operatorname{Var}\left(X_{1}+X_{2}\right)=\operatorname{Var}\left(X_{1}\right)+\operatorname{Var}\left(X_{2}\right)=2 \operatorname{Var}\left(X_{2}\right)$. Can this be right?
- Cauchy distribution doesn't have finite variance or mean.
- Some standard facts we'll learn later in the course (central limit theorem, law of large numbers) don't apply to it.


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## Beta distribution: Alice and Bob revisited

- Suppose I have a coin with a heads probability $p$ that I don't know much about.
- What do I mean by not knowing anything? Let's say that I think $p$ is equally likely to be any of the numbers $\{0, .1, .2, .3, .4, \ldots, .9,1\}$.
- Now imagine a multi-stage experiment where I first choose $p$ and then I toss $n$ coins.
- Given that number $h$ of heads is $a-1$, and $b-1$ tails, what's conditional probability $p$ was a certain value $x$ ?
- $P(p=x \mid h=(a-1))=\frac{\frac{1}{11}\binom{n}{a-1} x^{a-1}(1-x)^{b-1}}{P\{h=(a-1)\}}$ which is $x^{a-1}(1-x)^{b-1}$ times a constant that doesn't depend on $x$.


## Beta distribution

- Suppose I have a coin with a heads probability $p$ that I really don't know anything about. Let's say $p$ is uniform on $[0,1]$.
- Now imagine a multi-stage experiment where I first choose $p$ uniformly from $[0,1]$ and then I toss $n$ coins.
- If I get, say, $a-1$ heads and $b-1$ tails, then what is the conditional probability density for $p$ ?
- Turns out to be a constant (that doesn't depend on $x$ ) times $x^{a-1}(1-x)^{b-1}$.
- $\frac{1}{B(a, b)} x^{a-1}(1-x)^{b-1}$ on $[0,1]$, where $B(a, b)$ is constant chosen to make integral one. Can be shown that $B(a, b)=\frac{\Gamma(a) \Gamma(b)}{\Gamma(a+b)}$.
- What is $E[X]$ ?
- Answer: $\frac{a}{a+b}$.

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