# 18.175: Lecture 17 

# Poisson random variables 

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## Outline

More on random walks and local CLT

Poisson random variable convergence

Extend CLT idea to stable random variables

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## Recall local CLT for walks on $\mathbb{Z}$

- Suppose $X \in b+h \mathbb{Z}$ a.s. for some fixed constants $b$ and $h$.
- Observe that if $\phi_{X}(\lambda)=1$ for some $\lambda \neq 0$ then $X$ is supported on (some translation of) $(2 \pi / \lambda) \mathbb{Z}$. If this holds for all $\lambda$, then $X$ is a.s. some constant. When the former holds but not the latter (i.e., $\phi_{X}$ is periodic but not identically 1 ) we call $X$ a lattice random variable.
- Write $p_{n}(x)=P\left(S_{n} / \sqrt{n}=x\right)$ for $x \in \mathcal{L}_{n}:=(n b+h \mathbb{Z}) / \sqrt{n}$ and $n(x)=\left(2 \pi \sigma^{2}\right)^{-1 / 2} \exp \left(-x^{2} / 2 \sigma^{2}\right)$.
- Assume $X_{i}$ are i.i.d. lattice with $E X_{i}=0$ and $E X_{i}^{2}=\sigma^{2} \in(0, \infty)$. Theorem: As $n \rightarrow \infty$,

$$
\left|\sup _{x \in \mathcal{L}^{n}}\right| n^{1 / 2} / h p_{n}(x)-n(x) \mid \rightarrow 0 .
$$

## Recall local CLT for walks on $\mathbb{Z}$

- Proof idea: Use characteristic functions, reduce to periodic integral problem. Look up "Fourier series". Note that for $Y$ supported on $a+\theta \mathbb{Z}$, we have $P(Y=x)=\frac{1}{2 \pi / \theta} \int_{-\pi / \theta}^{\pi / \theta} e^{-i t x} \phi_{Y}(t) d t$.


## Extending this idea to higher dimensions

- Example: suppose we have random walk on $\mathbb{Z}$ that at each step tosses fair 4 -sided coin to decide whether to go 1 unit left, 1 unit right, 2 units left, or 2 units right?
- What is the probability that the walk is back at the origin after one step? Two steps? Three steps?
- Let's compute this in Mathematica by writing out the characteristic function $\phi_{X}$ for one-step increment $X$ and calculating $\int_{0}^{2 \pi} \phi_{X}^{k}(t) d t / 2 \pi$.
- How about a random walk on $\mathbb{Z}^{2}$ ?
- Can one use this to establish when a random walk on $\mathbb{Z}^{d}$ is recurrent versus transient?


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## Poisson random variables: motivating questions

- How many raindrops hit a given square inch of sidewalk during a ten minute period?
- How many people fall down the stairs in a major city on a given day?
- How many plane crashes in a given year?
- How many radioactive particles emitted during a time period in which the expected number emitted is 5 ?
- How many calls to call center during a given minute?
- How many goals scored during a 90 minute soccer game?
- How many notable gaffes during 90 minute debate?
- Key idea for all these examples: Divide time into large number of small increments. Assume that during each increment, there is some small probability of thing happening (independently of other increments).


## Bernoulli random variable with $n$ large and $n p=\lambda$

- Let $\lambda$ be some moderate-sized number. Say $\lambda=2$ or $\lambda=3$. Let $n$ be a huge number, say $n=10^{6}$.
- Suppose I have a coin that comes up heads with probability $\lambda / n$ and I toss it $n$ times.
- How many heads do I expect to see?
- Answer: $n p=\lambda$.
- Let $k$ be some moderate sized number (say $k=4$ ). What is the probability that I see exactly $k$ heads?
- Binomial formula:

$$
\binom{n}{k} p^{k}(1-p)^{n-k}=\frac{n(n-1)(n-2) \ldots(n-k+1)}{k!} p^{k}(1-p)^{n-k} .
$$

- This is approximately $\frac{\lambda^{k}}{k!}(1-p)^{n-k} \approx \frac{\lambda^{k}}{k!} e^{-\lambda}$.
- A Poisson random variable $X$ with parameter $\lambda$ satisfies $P\{X=k\}=\frac{\lambda^{k}}{k!} e^{-\lambda}$ for integer $k \geq 0$.


## Probabilities sum to one

- A Poisson random variable $X$ with parameter $\lambda$ satisfies $p(k)=P\{X=k\}=\frac{\lambda^{k}}{k!} e^{-\lambda}$ for integer $k \geq 0$.
- How can we show that $\sum_{k=0}^{\infty} p(k)=1$ ?
- Use Taylor expansion $e^{\lambda}=\sum_{k=0}^{\infty} \frac{\lambda^{k}}{k!}$.


## Expectation

- A Poisson random variable $X$ with parameter $\lambda$ satisfies $P\{X=k\}=\frac{\lambda^{k}}{k!} e^{-\lambda}$ for integer $k \geq 0$.
- What is $E[X]$ ?
- We think of a Poisson random variable as being (roughly) a Bernoulli ( $n, p$ ) random variable with $n$ very large and $p=\lambda / n$.
- This would suggest $E[X]=\lambda$. Can we show this directly from the formula for $P\{X=k\}$ ?
- By definition of expectation

$$
E[X]=\sum_{k=0}^{\infty} P\{X=k\} k=\sum_{k=0}^{\infty} k \frac{\lambda^{k}}{k!} e^{-\lambda}=\sum_{k=1}^{\infty} \frac{\lambda^{k}}{(k-1)!} e^{-\lambda} .
$$

- Setting $j=k-1$, this is $\lambda \sum_{j=0}^{\infty} \frac{\lambda_{j}^{j}}{j!} e^{-\lambda}=\lambda$.


## Variance

- Given $P\{X=k\}=\frac{\lambda^{k}}{k!} e^{-\lambda}$ for integer $k \geq 0$, what is $\operatorname{Var}[X]$ ?
- Think of $X$ as (roughly) a Bernoulli ( $n, p$ ) random variable with $n$ very large and $p=\lambda / n$.
- This suggests $\operatorname{Var}[X] \approx n p q \approx \lambda$ (since $n p \approx \lambda$ and $q=1-p \approx 1$ ). Can we show directly that $\operatorname{Var}[X]=\lambda$ ?
- Compute

$$
E\left[X^{2}\right]=\sum_{k=0}^{\infty} P\{X=k\} k^{2}=\sum_{k=0}^{\infty} k^{2} \frac{\lambda^{k}}{k!} e^{-\lambda}=\lambda \sum_{k=1}^{\infty} k \frac{\lambda^{k-1}}{(k-1)!} e^{-\lambda} .
$$

- Setting $j=k-1$, this is

$$
\lambda\left(\sum_{j=0}^{\infty}(j+1) \frac{\lambda^{j}}{j!} e^{-\lambda}\right)=\lambda E[X+1]=\lambda(\lambda+1)
$$

- Then $\operatorname{Var}[X]=E\left[X^{2}\right]-E[X]^{2}=\lambda(\lambda+1)-\lambda^{2}=\lambda$.


## Poisson convergence

- Idea: if we have lots of independent random events, each with very small probability to occur, and expected number to occur is $\lambda$, then total number that occur is roughly Poisson $\lambda$.
- Theorem: Let $X_{n, m}$ be independent $\{0,1\}$-valued random variables with $P\left(X_{n, m}=1\right)=p_{n, m}$. Suppose $\sum_{m=1}^{n} p_{n, m} \rightarrow \lambda$ and $\max _{1 \leq m \leq n} p_{n, m} \rightarrow 0$. Then $S_{n}=X_{n, 1}+\ldots+X_{n, n} \Longrightarrow Z$ were $Z$ is $\operatorname{Poisson}(\lambda)$.
- Proof idea: Just write down the log characteristic functions for Bernoulli and Poisson random variables. Check the conditions of the continuity theorem.


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## Recall continuity theorem

- Strong continuity theorem: If $\mu_{n} \Longrightarrow \mu_{\infty}$ then $\phi_{n}(t) \rightarrow \phi_{\infty}(t)$ for all $t$. Conversely, if $\phi_{n}(t)$ converges to a limit that is continuous at 0 , then the associated sequence of distributions $\mu_{n}$ is tight and converges weakly to a measure $\mu$ with characteristic function $\phi$.


## Recall CLT idea

- Let $X$ be a random variable.
- The characteristic function of $X$ is defined by $\phi(t)=\phi_{X}(t):=E\left[e^{i t X}\right]$.
- And if $X$ has an $m$ th moment then $E\left[X^{m}\right]=i^{m} \phi_{X}^{(m)}(0)$.
- In particular, if $E[X]=0$ and $E\left[X^{2}\right]=1$ then $\phi_{X}(0)=1$ and $\phi_{X}^{\prime}(0)=0$ and $\phi_{X}^{\prime \prime}(0)=-1$.
- Write $L_{X}:=-\log \phi_{X}$. Then $L_{X}(0)=0$ and
$L_{X}^{\prime}(0)=-\phi_{x}^{\prime}(0) / \phi_{x}(0)=0$ and
$L_{X}^{\prime \prime}=-\left(\phi_{X}^{\prime \prime}(0) \phi_{X}(0)-\phi_{X}^{\prime}(0)^{2}\right) / \phi_{X}(0)^{2}=1$.
- If $V_{n}=n^{-1 / 2} \sum_{i=1}^{n} X_{i}$ where $X_{i}$ are i.i.d. with law of $X$, then $L_{V_{n}}(t)=n L_{X}\left(n^{-1 / 2} t\right)$.
- When we zoom in on a twice differentiable function near zero (scaling vertically by $n$ and horizontally by $\sqrt{n}$ ) the picture looks increasingly like a parabola.


## Stable laws

- Question? Is it possible for something like a CLT to hold if $X$ has infinite variance? Say we write $V_{n}=n^{-a} \sum_{i=1}^{n} X_{i}$ for some a. Could the law of these guys converge to something non-Gaussian?
- What if the $L_{V_{n}}$ converge to something else as we increase $n$, maybe to some other power of $|t|$ instead of $|t|^{2}$ ?
- The the appropriately normalized sum should be converge in law to something with characteristic function $e^{-|t|^{\alpha}}$ instead of $e^{-|t|^{2}}$.
- We already saw that this should work for Cauchy random variables. What's the characteristic function in that case?
- Let's look up stable distributions.


## Infinitely divisible laws

- Say a random variable $X$ is infinitely divisible, for each $n$, there is a random variable $Y$ such that $X$ has the same law as the sum of $n$ i.i.d. copies of $Y$.
- What random variables are infinitely divisible?
- Poisson, Cauchy, normal, stable, etc.
- Let's look at the characteristic functions of these objects. What about compound Poisson random variables (linear combinations of Poisson random variables)? What are their characteristic functions like?
- More general constructions are possible via Lévy Khintchine representation.

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