

# 18.175: Lecture 17

## Poisson random variables

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More on random walks and local CLT

Poisson random variable convergence

Extend CLT idea to stable random variables

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- ▶ Write  $p_n(x) = P(S_n/\sqrt{n} = x)$  for  $x \in \mathcal{L}_n := (nb + h\mathbb{Z})/\sqrt{n}$  and  $n(x) = (2\pi\sigma^2)^{-1/2} \exp(-x^2/2\sigma^2)$ .

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- ▶ Assume  $X_i$  are i.i.d. lattice with  $EX_i = 0$  and  $EX_i^2 = \sigma^2 \in (0, \infty)$ . **Theorem:** As  $n \rightarrow \infty$ ,

$$\sup_{x \in \mathcal{L}_n} \left| \frac{n^{1/2}}{h} p_n(x) - n(x) \right| \rightarrow 0.$$

- ▶ **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

$$P(Y = x) = \frac{1}{2\pi/\theta} \int_{-\pi/\theta}^{\pi/\theta} e^{-itx} \phi_Y(t) dt.$$

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- ▶ 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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- ▶ **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 $np = \lambda$

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- ▶ Use Taylor expansion  $e^{\lambda} = \sum_{k=0}^{\infty} \frac{\lambda^k}{k!}$ .

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

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

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- ▶ **Theorem:** Let  $X_{n,m}$  be independent  $\{0, 1\}$ -valued random variables with  $P(X_{n,m} = 1) = p_{n,m}$ . Suppose  $\sum_{m=1}^n p_{n,m} \rightarrow \lambda$  and  $\max_{1 \leq m \leq n} p_{n,m} \rightarrow 0$ . Then  $S_n = X_{n,1} + \dots + X_{n,n} \implies Z$  where  $Z$  is Poisson( $\lambda$ ).

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- ▶ **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  $\mu_n \implies \mu_\infty$  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  $\phi$ .

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- ▶ In particular, if  $E[X] = 0$  and  $E[X^2] = 1$  then  $\phi_X(0) = 1$  and  $\phi_X'(0) = 0$  and  $\phi_X''(0) = -1$ .

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- ▶ Write  $L_X := -\log \phi_X$ . Then  $L_X(0) = 0$  and  $L_X'(0) = -\phi_X'(0)/\phi_X(0) = 0$  and  $L_X''(0) = -(\phi_X''(0)\phi_X(0) - \phi_X'(0)^2)/\phi_X(0)^2 = 1$ .

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- ▶ The **characteristic function** of  $X$  is defined by  $\phi(t) = \phi_X(t) := E[e^{itX}]$ .
- ▶ And if  $X$  has an  $m$ th moment then  $E[X^m] = i^m \phi_X^{(m)}(0)$ .
- ▶ In particular, if  $E[X] = 0$  and  $E[X^2] = 1$  then  $\phi_X(0) = 1$  and  $\phi_X'(0) = 0$  and  $\phi_X''(0) = -1$ .
- ▶ Write  $L_X := -\log \phi_X$ . Then  $L_X(0) = 0$  and  $L_X'(0) = -\phi_X'(0)/\phi_X(0) = 0$  and  $L_X''(0) = -(\phi_X''(0)\phi_X(0) - \phi_X'(0)^2)/\phi_X(0)^2 = 1$ .
- ▶ If  $V_n = n^{-1/2} \sum_{i=1}^n X_i$  where  $X_i$  are i.i.d. with law of  $X$ , then  $L_{V_n}(t) = nL_X(n^{-1/2}t)$ .

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- ▶ 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.

- ▶ 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?

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- ▶ 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  $Y$ .

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- ▶ More general constructions are possible via Lévy Khintchine representation.