# 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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 for  $x \in \mathcal{L}_n := (nb + h\mathbb{Z})/\sqrt{n}$   
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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  $n \to \infty$ ,

$$\left|\sup_{x\in\mathcal{L}^n}|n^{1/2}/hp_n(x)-n(x)|\to 0.\right.$$

Proof idea: Use characteristic functions, reduce to periodic integral problem. Look up "Fourier series". Note that for Y supported on a + θZ, we have P(Y = x) = 1/(2π/θ) ∫<sup>π/θ</sup><sub>-π/θ</sub> e<sup>-itx</sup>φ<sub>Y</sub>(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 Z<sup>d</sup> 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).

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

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#### Expectation

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

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

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- ► This suggests  $\operatorname{Var}[X] \approx npq \approx \lambda$  (since  $np \approx \lambda$  and  $q = 1 p \approx 1$ ). Can we show directly that  $\operatorname{Var}[X] = \lambda$ ?

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• Then 
$$\operatorname{Var}[X] = E[X^2] - E[X]^2 = \lambda(\lambda + 1) - \lambda^2 = \lambda$$
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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  $\max_{1 \le m \le n} p_{n,m} \rightarrow 0$ . Then  $S_n = X_{n,1} + \ldots + X_{n,n} \implies Z$  were 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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Strong continuity theorem: If µ<sub>n</sub> ⇒ µ<sub>∞</sub> then φ<sub>n</sub>(t) → φ<sub>∞</sub>(t) for all t. Conversely, if φ<sub>n</sub>(t) converges to a limit that is continuous at 0, then the associated sequence of distributions µ<sub>n</sub> is tight and converges weakly to a measure µ with characteristic function φ. • Let X be a random variable.

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▶ Write 
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. Then  $L_X(0) = 0$  and  $L'_X(0) = -\phi'_X(0)/\phi_X(0) = 0$  and  $L''_X = -(\phi''_X(0)\phi_X(0) - \phi'_X(0)^2)/\phi_X(0)^2 = 1$ .

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- Write L<sub>X</sub> := -log φ<sub>X</sub>. Then L<sub>X</sub>(0) = 0 and L'<sub>X</sub>(0) = -φ'<sub>X</sub>(0)/φ<sub>X</sub>(0) = 0 and L''<sub>X</sub> = -(φ''<sub>X</sub>(0)φ<sub>X</sub>(0) - φ'<sub>X</sub>(0)<sup>2</sup>)/φ<sub>X</sub>(0)<sup>2</sup> = 1.
   If V<sub>n</sub> = n<sup>-1/2</sup> ∑<sup>n</sup><sub>i=1</sub> X<sub>i</sub> where X<sub>i</sub> are i.i.d. with law of X, then L<sub>Y</sub>(t) = nL<sub>X</sub>(n<sup>-1/2</sup>t).

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- If  $V_n = n^{-1/2} \sum_{i=1}^n X_i$  where  $X_i$  are i.i.d. with law of X, then  $L_{V_n}(t) = nL_X(n^{-1/2}t)$ .
- When we zoom in on a twice differentiable function near zero (scaling vertically by n and horizontally by √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<sub>n</sub> = n<sup>-a</sup> ∑<sup>n</sup><sub>i=1</sub> X<sub>i</sub> for some a. Could the law of these guys converge to something non-Gaussian?

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- Let's look up stable distributions.

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- More general constructions are possible via Lévy Khintchine representation.