# Math 340: Lec 16 Big Ideas Journal (Random walks continued; central limit theorem)

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## Random walks

**Theorem 1** (Ballot theorem). Consider  $N_n^+(0,b)$ : paths from  $S_0=0$  to  $S_n=b$  for which  $S_k>0 \forall k\in\{1,\ldots,n\}$ . We can think of  $S_k$  as how many votes ahead candidate X is over candidate Y on election night.

$$\forall b \neq 0, N_n^+(0, b) = \frac{|b|}{n} N_n(0, b)$$

|b|/n is the fraction of paths that don't touch the x-axis (never go negative).

## Central limit theorem

#### CLT applied to random walks

**Example** (CLT random walks). Consider the following probability measure (bakes in equal parity)

$$\mathbb{P}_n^{\ 0}(S_{2n} = 2k) = \binom{2n}{n+k} 2^{-(2n)}$$

Let  $X_k$  be a random variable denoting how much we move on step k.

$$\mathbb{E}[S_n] = \mathbb{E}[\alpha + \sum_{k=1}^n X_k(\omega) = \sum_{k=1}^n \mathbb{E}[X_k] = 0$$

$$\sum_{k=1}^n X_k(X_k) = \sum_{k=1}^n \mathbb{E}[X_k] = 0$$

$$Var(S_n) = \sum_{k=1}^n Var(X_k) = \sum_{k=1}^n \mathbb{E}(|X_k - \mu|^2) = \sum_{k=1}^n \mathbb{E}(|X_k - 0|^2) = \sum_{k=1}^n 1 = n$$

$$SD(S_n) = \sqrt{n}$$

**Remark.** The CLT suggests us that we shouldn't be surprised if  $S_n \approx O(\sqrt{n})$ .

**Theorem 2** (Central Limit Theorem for random walks). For any  $\alpha, \beta$ 

$$\lim_{n \to \infty} \mathbb{P}_n^{\alpha} \left( \alpha \le \frac{S_n}{\sqrt{n}} \le \beta \right) = \int_{\alpha}^{\beta} \frac{e^{-\frac{y^2}{2}}}{\sqrt{2\pi}} dy$$

Remark. The CLT lets us bound how far away we expect a random walker to wander from the mean of the walk.

#### General CLT

**Theorem 3** (Central Limit Theorem). Suppose  $X_1, \ldots, X_n$  is a sequence of i.i.d random variables with  $\mu = \mathbb{E}[X_i], \sigma^2 = \text{Var}(X_i)$  and  $\mathbb{E}[X_i^4] < \infty$ . For any  $\alpha, \beta$ 

$$\lim_{n \to \infty} \mathbb{P}\left(\alpha \le \frac{(X_1 + \dots + X_n) - \mu n}{\sqrt{n\sigma^2}} \le \beta\right) = \int_{\alpha}^{\beta} \frac{e^{-\frac{y^2}{2}}}{\sqrt{2\pi}} dy = \Phi(\beta) - \Phi(\alpha)$$

**Remark.**  $\sqrt{n\sigma^2}$  is the standard deviation for a sum of iid  $X_1, \ldots, X_n$ . If  $X_1 + \ldots + X_n$  were replaced by some other random variable that is a sum of components (e.g. a Poisson arrival process variable), the denominator of the CLT would reflect the SD of that random variable.

**Remark.** Note that the CLT for random walks is a particular case of the general CLT where  $S_n$  is a sum of random step variables with mean zero and variance 1.

**Example** (Using CLT to bound p-coin head count). Let  $Z_n$  be the nubmer of heads we see in n tosses. The marignal distribution of  $Z_n$  is given by the binomial distribution. As we know,  $Z_n = X_1 + \ldots + X_n$  where  $X_j$  is a bernoulli random variable.  $\mathbb{E}[X_j] = p$  and  $\text{Var}(X_j) = p(1-p)$ . Per the CLT:

$$\mathbb{P}\left(\alpha \le \frac{(X_1 + \dots + X_n) - np}{\sqrt{np(1-p)}} \le b\right) = \int_{\alpha}^{\beta} \frac{e^{-\frac{y^2}{2}}}{\sqrt{2\pi}} dy$$

Also, note that the CLT expression gives us

$$\mathbb{P}\left(\alpha\sqrt{(np(1-p))} \le (X_1 + \ldots + X_n) - np \le \beta\sqrt{np(1-p)}\right)$$
$$= \mathbb{P}(Z_n \in (np + \alpha\sqrt{np(1-p)}, np + \beta\sqrt{np(1-p)})$$

i.e. the probability that Z is within  $\alpha$  and  $\beta$  standard deviations of its mean.