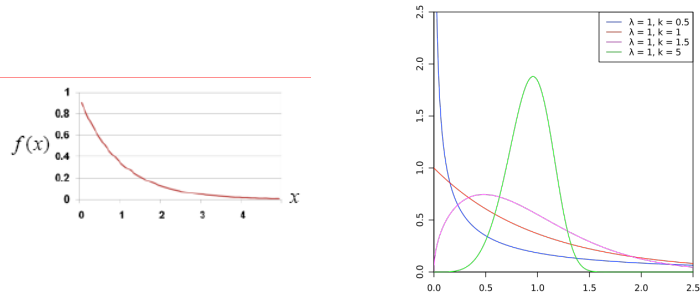
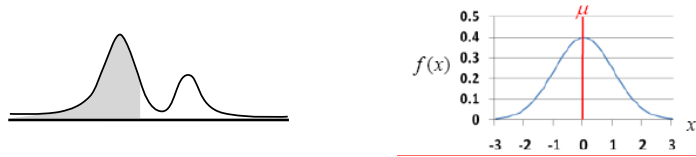


limit theorems



Limit Theorems

Consider i.i.d. (independent, identically distributed) random vars X_1, X_2, X_3, \dots

X_i has $\mu = E[X_i]$ and $\sigma^2 = \text{Var}[X_i]$

Consider random variables

$$X_1 + X_2 + \dots + X_n$$

and

$$\frac{1}{n} \sum_{i=1}^n X_i$$

Law of Large Numbers

If we observe a random variable X many times (independently) and take the average, this average will converge to a real number which is $E(X)$.

Formally, let X_1, \dots, X_n be independent, identically distributed random variables with mean μ .

Define $A_n = \frac{1}{n} \sum_{i=1}^n X_i$ Then for any $\alpha > 0$ we have

$$\Pr(|A_n - \mu| > \alpha) \rightarrow 0 \quad \text{as } n \rightarrow \infty$$

Proof: Use Chebychev's inequality.

the central limit theorem (CLT)

Consider i.i.d. (independent, identically distributed) random vars X_1, X_2, X_3, \dots

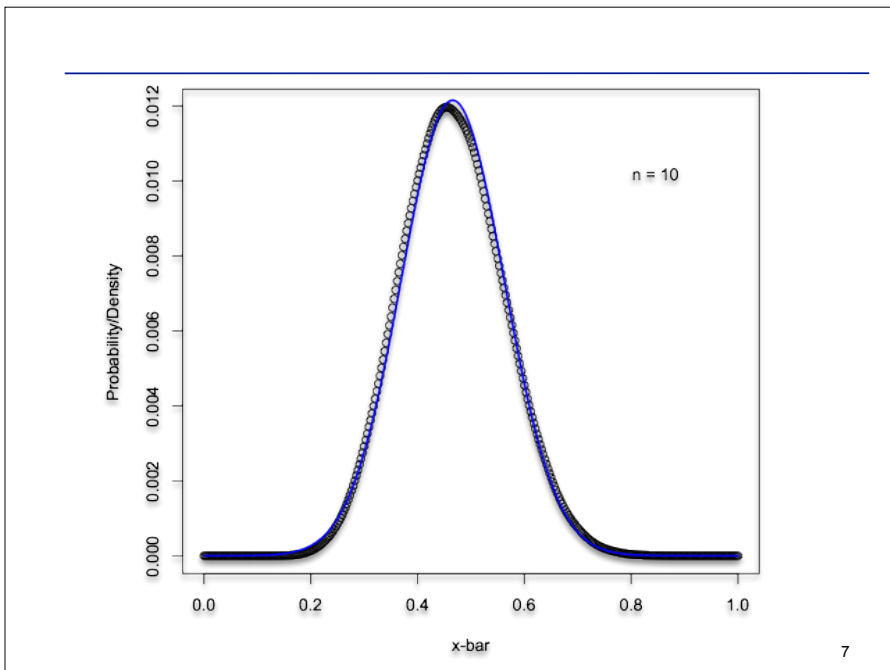
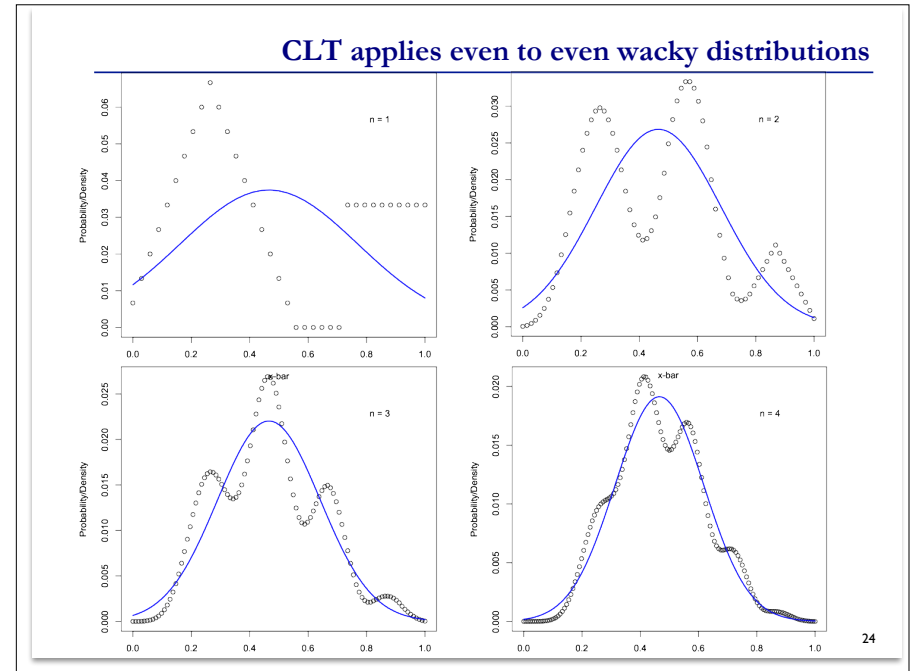
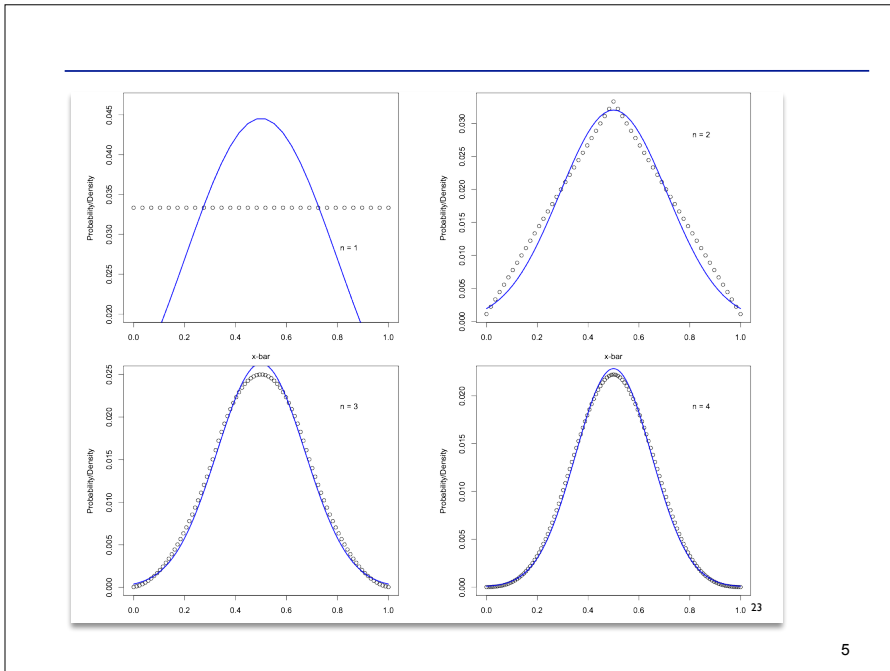
X_i has $\mu = E[X_i]$ and $\sigma^2 = \text{Var}[X_i]$

As $n \rightarrow \infty$,

$$\frac{X_1 + X_2 + \dots + X_n - n\mu}{\sigma\sqrt{n}} \rightarrow N(0, 1)$$

Restated: As $n \rightarrow \infty$,

$$M_n = \frac{1}{n} \sum_{i=1}^n X_i \rightarrow N\left(\mu, \frac{\sigma^2}{n}\right)$$



CLT in the real world

CLT is the reason many things appear normally distributed
 Many quantities = sums of (roughly) independent random vars

Exam scores: sums of individual problems
People's heights: sum of many genetic & environmental factors
Measurements: sums of various small instrument errors

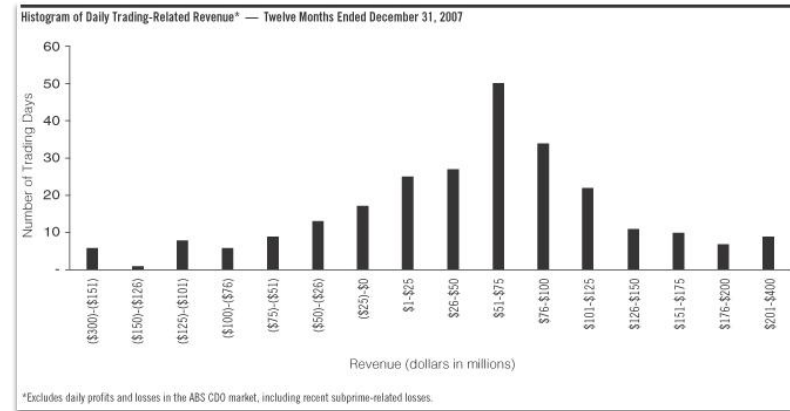
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in the real world...

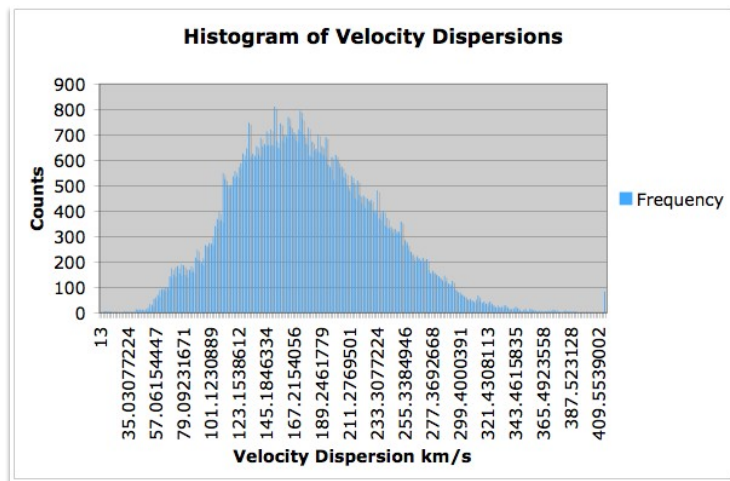


in the real world...



*Excludes daily profits and losses in the ABS CDO market, including recent subprime-related losses.

in the real world...



the central limit theorem (CLT)

Consider i.i.d. (independent, identically distributed) random vars X_1, X_2, X_3, \dots

X_i has $\mu = E[X_i]$ and $\sigma^2 = \text{Var}[X_i]$

As $n \rightarrow \infty$,

$$\frac{X_1 + X_2 + \dots + X_n - n\mu}{\sigma\sqrt{n}} \longrightarrow N(0, 1)$$

Restated: As $n \rightarrow \infty$,

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Example 1

Number of students who enroll in a class is Poisson (100).
Professor will teach course in two sections if more than 120 students enroll. What is the probability of two sections?

When applying approximation, use **continuity correction**:
Think of $\Pr(X=i) = \Pr(i-0.5 < X < i + 0.5)$

Recall Poisson (100) is sum of 100 independent Poisson(1) random variables, so we can apply CLT.

$$\begin{aligned} \Pr(X > 119.5) &= \Pr\left(\frac{X - 100}{\sqrt{100}} \geq \frac{119.5 - 100}{\sqrt{100}}\right) \\ &\approx 1 - \Phi(1.95) \\ &\approx 0.0256. \end{aligned}$$

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Example 2

If 10 fair die are rolled, find the approximate probability that the sum obtained is between 30 and 40, inclusive, using CLT.

X_i is the value of die # i.

$$E(X_i) = 7/2 \text{ and } \text{Var}(X_i) = 35/12$$

$$X = X_1 + X_2 + \dots + X_{10}$$

$$E(X) = 35$$

$$\text{Var}(X) = 350/12$$

$$\begin{aligned} \Pr(29.5 \leq X \leq 40.5) &= \Pr\left(\frac{29.5 - 35}{\sqrt{350/12}} \leq \frac{X - 35}{\sqrt{350/12}} \leq \frac{40.5 - 35}{\sqrt{350/12}}\right) \\ &\approx 2 \cdot \Phi(1.0184) - 1 \\ &\approx 0.692. \end{aligned}$$

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