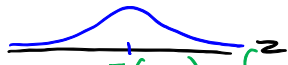


GAUSSIAN RVs

$f_z(z) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2}$
 $E(z) = \int_{\text{odd}} z f_z(z) dz = 0$ $E(z^2) = \int_{\text{even}} z^2 f_z(z) dz = \sigma_z^2 = 1 \rightarrow \sigma_z = 1$



$F_z(z) = \int_{-\infty}^z f_z(z) dz$ $P_r(z \leq \alpha) = \Phi(\alpha)$ $\Phi(-\alpha) = 1 - \Phi(\alpha)$
 $P_r(z > \alpha) = 1 - \Phi(\alpha)$
 $P(-\alpha \leq z \leq \alpha) = 2\Phi(\alpha) - 1$

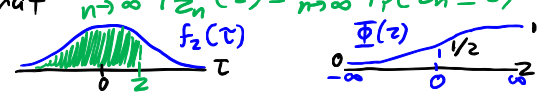
$P(|z| \leq 1) = 2\Phi(1) - 1 = 2 \cdot 0.8413 - 1 = 0.6827$ one standard deviation
 $P(|z| \leq 2) = 2\Phi(2) - 1 = 2 \cdot 0.9772 - 1 = 0.9545$ two standard deviations
 $P(|z| \leq 3) = 2\Phi(3) - 1 = 2 \cdot 0.9987 - 1 = 0.9974$ three standard deviations

STANDARDIZING A RV

$X, E(X), \sigma_X^2$ define intermediate RV $Y = X - E(X)$ $E(Y) = 0$
 variance hasn't changed (scalar shift) $\sigma_Y^2 = \sigma_X^2$
 $\sigma_Y = \sigma_X$, which we can use to get a unit-variance RV. $Z = Y/\sigma_Y$
 $\sigma_Z^2 = (\frac{1}{\sigma_Y})^2 \sigma_Y^2 = 1$ now, Z has $E(Z) = 0$ and $\sigma_Z^2 = 1$, regardless of X !
 $Z = (X - E(X)) / \sigma_X$ will always convert $X \sim N(\mu, \sigma^2)$ to $Z \sim N(0, 1)$
 $X = \sigma Z + \mu$ this operation is reversible $E(X) = \sigma E(Z) + E(\mu) = \mu$
 $\sigma_X^2 = \sigma^2 \text{var}(Z) = \sigma^2$

CENTRAL LIMIT THEOREM

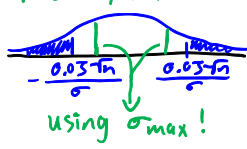
Suppose X_1, X_2, \dots, X_n are IID with $\mu = E(X)$ and $\sigma^2 = \sigma_X^2$.
 Sample mean $M_n = \frac{X_1 + X_2 + \dots + X_n}{n}$ where $E(M_n) = \mu$ and $\sigma_{M_n}^2 = \sigma^2/n$.
 Let $Z_n = \frac{M_n - \mu}{\sigma/\sqrt{n}}$. The CLT says that $\lim_{n \rightarrow \infty} F_{Z_n}(z) = \lim_{n \rightarrow \infty} P_r(Z_n \leq z) = \Phi(z)$
 But what does $\Phi(z)$ look like?



POLLSTER PROBLEM REVISITED

$\frac{\mu}{1-\mu} \frac{X_i=1}{X_i=0}$ i^{th} voter believes in legalization. Estimate $\mu \rightarrow$ how many people should be polled so $P_r(\mu - \epsilon \leq M_n \leq \mu + \epsilon) \geq 0.95$?

$P_r(-\epsilon \leq M_n - \mu \leq \epsilon) \rightarrow P_r(|M_n - \mu| \leq \epsilon) \geq 0.95 \rightarrow P(|M_n - \mu| > \epsilon) \leq 0.05$
 Let $\epsilon = 0.03$. Standardize M_n : $Z_n = \frac{M_n - E(M_n)}{\sigma_{M_n}} = \frac{M_n - \mu}{\sigma/\sqrt{n}} = \frac{M_n - \mu}{\sigma/\sqrt{n}}$
 $P(\frac{|M_n - \mu|}{\sigma/\sqrt{n}} > \frac{0.03}{\sigma/\sqrt{n}}) \leq 0.05 \rightarrow P(|Z_n| > \frac{0.03\sqrt{n}}{\sigma}) \leq 0.05$ CLT!



$\sigma^2 = \mu(1-\mu) \rightarrow \sigma_{\text{max}}^2 = 0.25 \rightarrow \sigma_{\text{max}} = 1/2$
 $P(|Z_n| > \frac{0.03\sqrt{n}}{\sigma_{\text{max}}}) = 2(1 - \Phi(\frac{0.06\sqrt{n}}{\sigma_{\text{max}}})) \leq 0.05$
 $1.96 \rightarrow 1.96 \leq 0.06\sqrt{n} \rightarrow \boxed{n \geq 1068}$