

Statistical Methods III: Spring 2013

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Ordered means + inference

Outline

- 1 Introduction
 - Ordered means
 - Regression/ML/Trinity of tests

- 2 Review
 - Normal, χ^2 , t + 2-tail

- 3 Testing equality constraints

Polity scores and child mortality

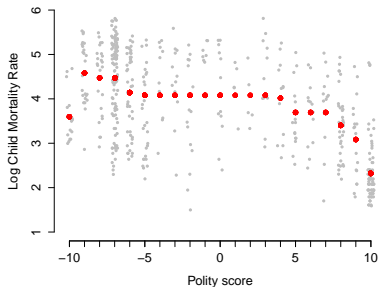
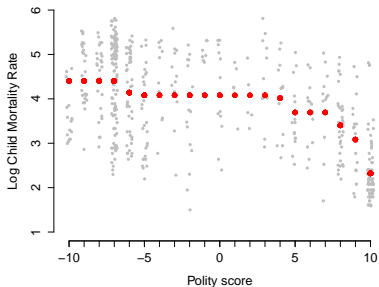
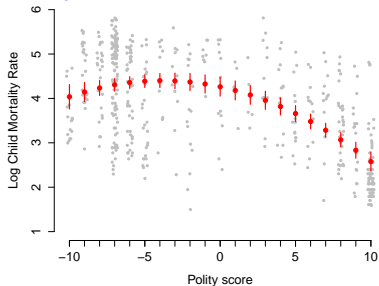
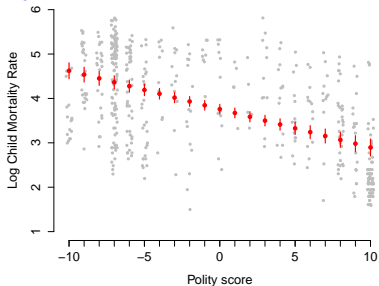


Table 1. Probit Parameter Estimates for Republican Vote Propensity, 1992

	Fully Informed Preferences	Uninformed Preferences	Information Effect (Difference)
Intercept	-1.542 (.766)	-.348 (1.112)	-1.194 (1.673)
Age (years)	-.0435 (.0278)	.0000 (.0389)	-.0436 (.0594)
Age squared (years)	.000429 (.000278)	-.000045 (.00384)	.000474 (.000590)
Education (years)	.0962 (.0337)	.0017 (.0536)	.0945 (.0779)
Income (percentile)	.399 (.329)	.828 (.563)	-.428 (.802)
Black	-1.063 (.319)	-2.285 (.479)	1.222 (.717)
Female	-.420 (.153)	.326 (.269)	-.746 (.381)

Normal Distribution

If $X \sim N(\mu, \sigma^2)$, the pdf is

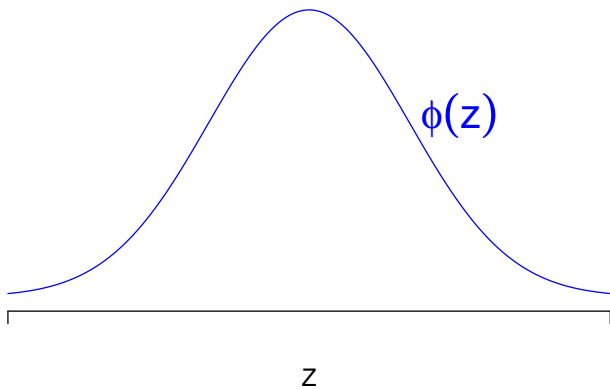
$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2}$$

Important: If $X \sim N(\mu, \sigma^2)$, then

$$Z = \frac{X - \mu}{\sigma} \sim N(0, 1).$$

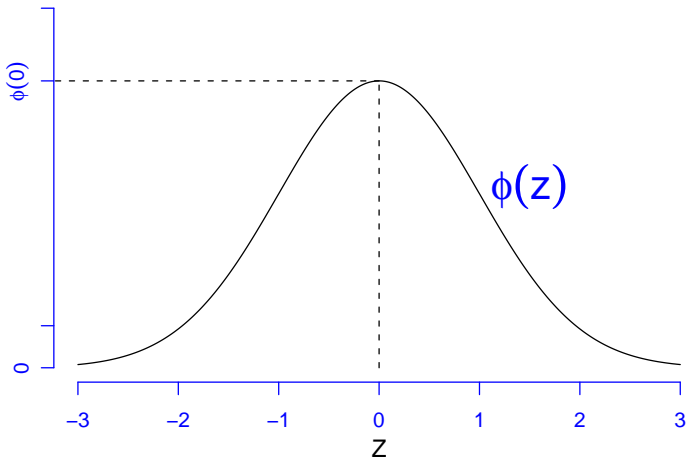
and conversely, we can start with $Z \sim N(0, 1)$,

$$X = Z\sigma + \mu \sim N(\mu, \sigma^2).$$



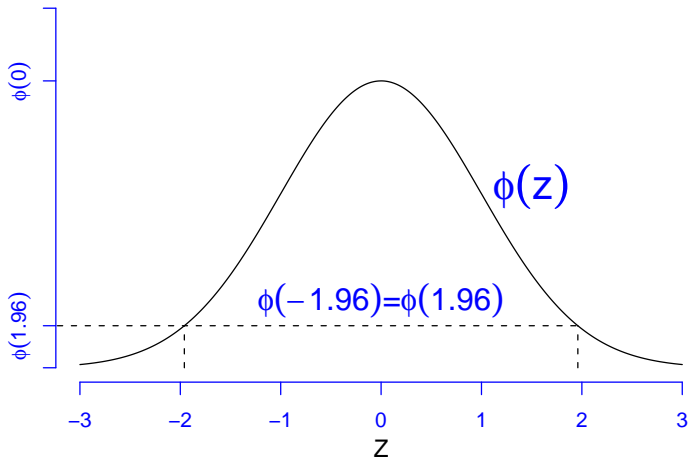
Let,

$$Z \sim N(0, 1)$$



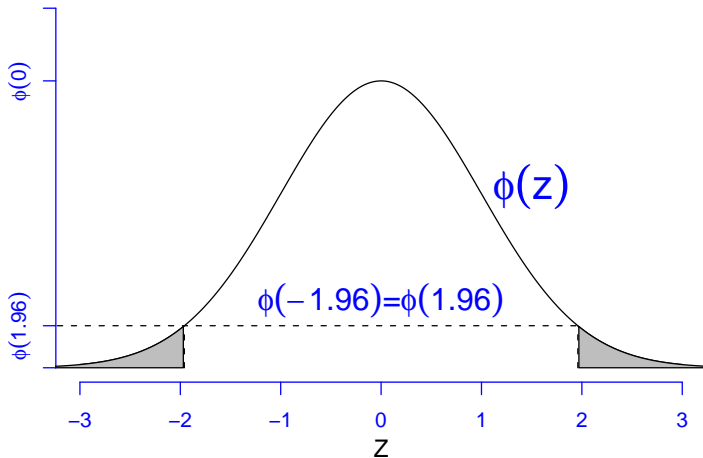
Let,

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Standard Normal Distribution

Memorize the following for quick calculations and seminars:

$$\begin{aligned}P(|Z| > 1) &= 2[1 - \Phi(1)] \approx 0.32 \\P(|Z| > 1.5) &= 2[1 - \Phi(1.5)] \approx 0.14 \\P(|Z| > 1.96) &= 2[1 - \Phi(1.96)] \approx 0.05 \\P(|Z| > 2) &= 2[1 - \Phi(2)] \approx 0.044 \\P(|Z| > 2.5) &= 2[1 - \Phi(2.5)] \approx 0.01 \\P(|Z| > 3) &= 2[1 - \Phi(3)] \approx 0.003\end{aligned}$$

noting that the representation in second term is by symmetry, if $z > 0$ then

$$\begin{aligned}P(|Z| > z) &= \Phi(-z) + (1 - \Phi(z)) \\&= (1 - \Phi(z)) + (1 - \Phi(z)) \\&= 2(1 - \Phi(z))\end{aligned}$$

Useful facts: χ^2

CB Lemma 5.4.1 (simplified)

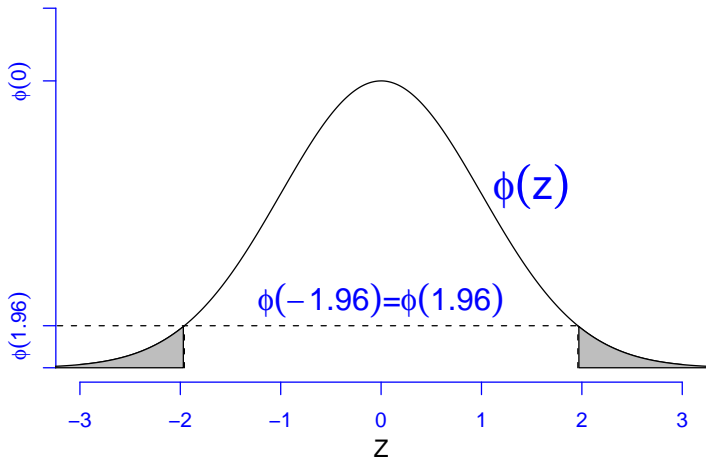
- If $Z \sim N(0, 1)$ then $Z^2 \sim \chi_1^2$
- If Z_1, \dots, Z_n are iid, and $Z_i \sim \chi_1^2$ then $\sum Z_i \sim \chi_n^2$

Properties restated

- Sums of squared normals are distrib χ^2
- If $X \sim \chi_{k_1}^2$ and $Y \sim \chi_{k_2}^2$ then $X + Y \sim \chi_{k_1+k_2}^2$

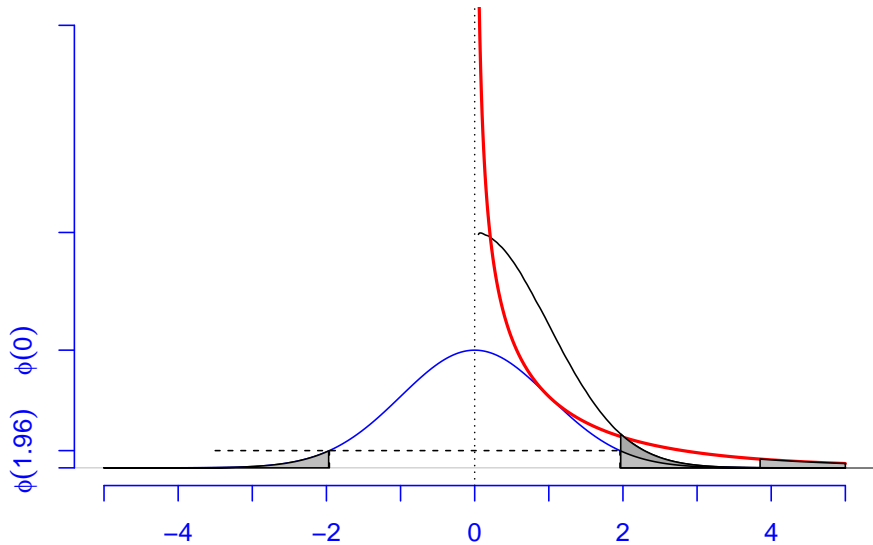
Expectations

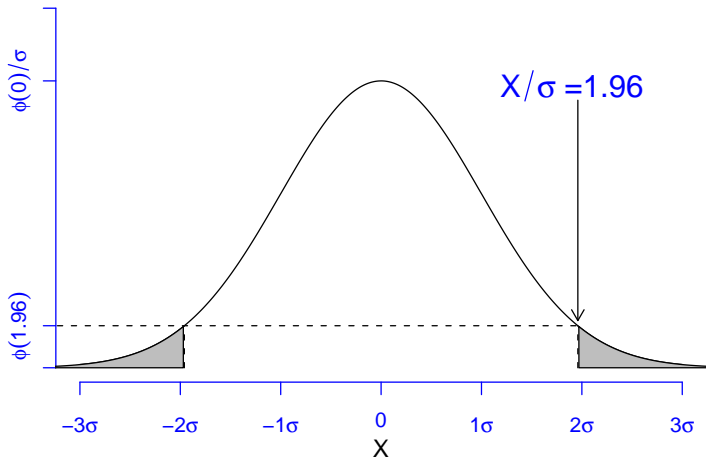
- $E\chi_k^2 = k$
- $V\chi_k^2 = 2k$



Let,

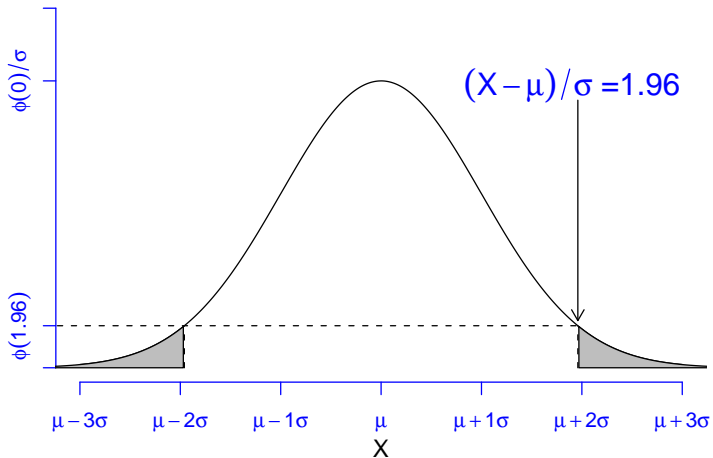
$$Z \sim N(0, 1)$$

χ_1^2 



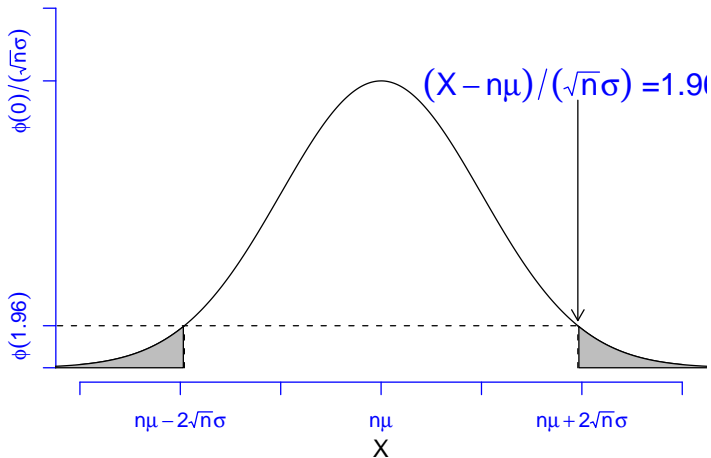
Let,

$$X = Z\sigma \quad \text{thus} \quad X \sim N(0, \sigma^2)$$



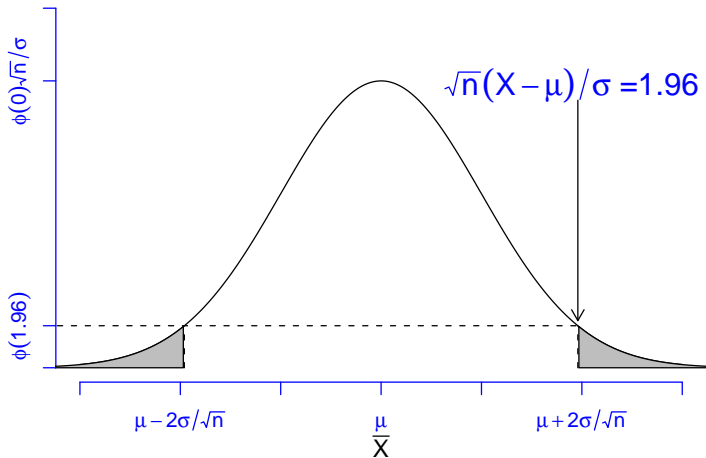
Let,

$$X = \mu + Z\sigma \quad \text{thus} \quad X \sim N(\mu, \sigma^2)$$



Let,

$$X = \sum_{i=1}^n (\mu + Z_i\sigma) \quad \text{thus} \quad X \sim N(n\mu, n\sigma^2)$$



Let,

$$\bar{X} = \frac{1}{n} \left(\sum_{i=1}^n (\mu + Z_i \sigma) \right) \quad \text{thus} \quad \bar{X} \sim N(\mu, \sigma^2/\sqrt{n})$$

Normal Distribution: critical values

Given $X_j \sim N(\mu, \sigma^2)$, and chosen n and α , and

$$H_0 : \mu = \mu_0 \quad \text{versus} \quad H_A : \mu \neq \mu_0$$

We can often state test statistic in different, equivalent ways—choose to make your life easy:

$$\begin{aligned}\alpha &= P(|\bar{X}_n - \mu_0| > c) \\ &= P(|\bar{X}_n - \mu_0| / (\sigma / \sqrt{n}) > c / (\sigma / \sqrt{n})) \\ &= P(|Z| > z) = \alpha\end{aligned}$$

where $Z \sim N(0, 1)$; e.g., if $\alpha = .0455$, then $z = 2$

Normal Distribution: critical values (cont'd)

For $P(|Z| > z) = \alpha$,

- In terms of normed value, $C = \{(-\infty, -z) \cup (z, \infty)\}$
- If we observe $Z \in C$, or equivalently

$$|\bar{X} - \mu_0| / (\sigma / \sqrt{n}) > z$$

we would reject H_0 .

For $P(|\bar{X}_n - \mu_0| > c) = \alpha$

- could also solve for $c = z\sigma / \sqrt{n}$
- In terms of deviation from mean

$$C = (-\infty, -z\sigma / \sqrt{n}) \cup (z\sigma / \sqrt{n}, \infty)$$

- given $\alpha = .0455$

$$C = (-\infty, -2\sigma / \sqrt{n}) \cup (2\sigma / \sqrt{n}, \infty)$$

Normal Distribution: critical values (example)

Let $X_i \sim N(\mu, \sigma^2)$, and $\sigma^2 = 9$

With a sample size $n = 16$, you observe $\bar{X} = 12$

Question: What decision would we make about $H_0 : \mu_0 = 10$?

$$T = \frac{\bar{X} - \mu_0}{\sigma/\sqrt{n}} = \frac{12 - 10}{3/4} = 3/2$$

The observed test statistic is not in critical region region, $T \notin C$ where

$$C = (-\infty, -2) \cup (2, \infty)\}$$

so would fail to reject H_0 (or loosely, accept H_0)

Normal Distribution: critical values (cont'd)

Connection to confidence intervals

With respect to $P(|\bar{X}_n - \mu| > c) = \alpha$, we just solved for $c = z\sigma/\sqrt{n}$ given α

Using this c , we can construct an interval centered around \bar{X} ,

$$(\bar{X}_n - c, \bar{X}_n + c) = (\bar{X}_n - z\sigma/\sqrt{n}, \bar{X}_n + z\sigma/\sqrt{n})$$

for specified values of z , σ and n

- this interval is referred to as a $(1 - \alpha) \times 100$ percent confidence interval
- $(1 - \alpha) \times 100$ percent of the time, this interval will include the true μ .

Normal Distribution: p-value

Let $X_i \sim N(\mu, \sigma^2)$, and $\sigma^2 = 9$

With a sample size $n = 16$, you observe $\bar{X} = 12$

Question: How confident are we that this sample could be produced with $\mu_0 = 10$?

Set $c = \bar{X} - \mu_0 = 12 - 10 = 2$

Significance level interpretation of α :

$$\begin{aligned}\alpha &= P(|\bar{X}_n - \mu_0| > c) \\ &= P(|\bar{X}_n - \mu_0| / (\sigma / \sqrt{n}) > c / (\sigma / \sqrt{n})) \\ &= P(|\bar{X}_n - \mu_0| / (\sigma / \sqrt{n}) > 2 / (3/4)) \\ &= P(|Z| > 3/2) \approx .14\end{aligned}$$

Note: referred to as p-value

Useful facts: χ^2

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- If $Z \sim N(0, 1)$ then $Z^2 \sim \chi_1^2$
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Expectations

- $E\chi_k^2 = k$
- $V\chi_k^2 = 2k$

Hypothesis tests of $r = \text{Rank}(R)$ restrictions

- Equality restriction versus unconstrained

$$H_0 : R\theta = 0 \text{ versus } H_A : R\theta \neq 0$$

Inference with two means

Let

$$\hat{\Delta} = (\hat{\Delta}_1, \hat{\Delta}_2) \sim N(\Delta, I_2)$$

where, possibly,

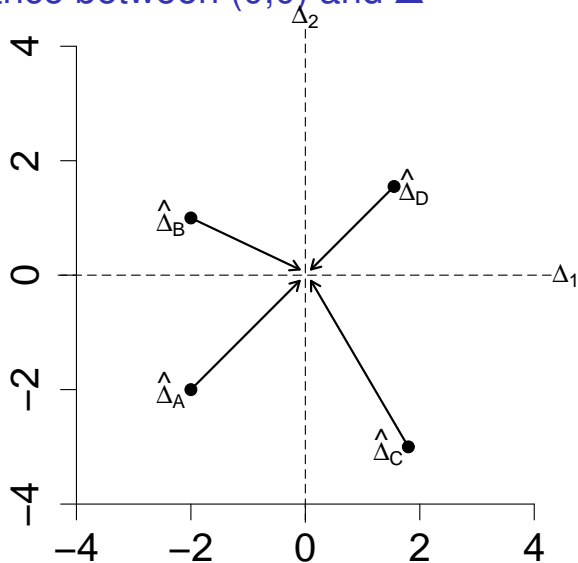
$$\hat{\Delta}_j = \hat{\mu}_j - \hat{\mu}_{j-1} \quad j \in \{1, 2\}$$

and I_k is a $k \times k$ identity matrix.

Hypotheses/comparisons

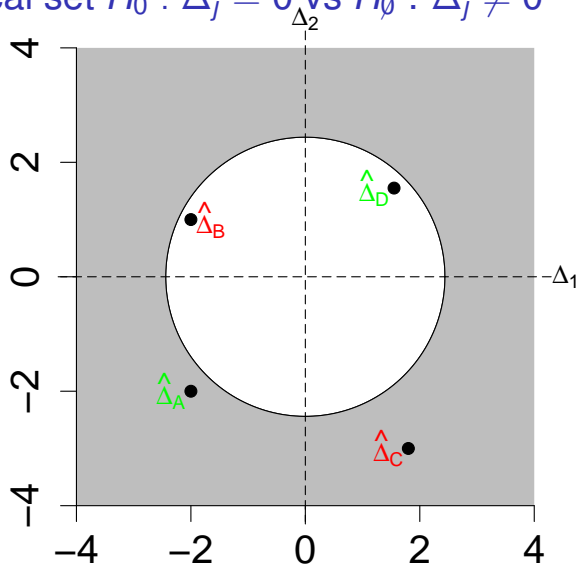
- $H_0 : \Delta = 0$ vs $H_\emptyset : \Delta \neq 0$
- what does this imply about R, r ?
- alternatives?

Distance between $(0,0)$ and $\hat{\Delta}$



$$\Delta_1^2 + \Delta_2^2 \sim \chi_2^2$$

Critical set $H_0 : \Delta_j = 0$ vs $H_\emptyset : \Delta_j \neq 0$



Critical value $\chi_{2,\alpha=.05}^2 = 6$, so $\sqrt{\Delta_1^2 + \Delta_2^2} = \sqrt{6} = \text{radius}$