

# IMETRICS Cheat-Sheet

A.S. - First version October 24, 2019

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## Statistics and Probability

### Basics

$$\Pr(x, y) = \Pr(X = x, Y = y) \text{ [joint]}$$

$$\Pr(y) = \sum_i \Pr(X = x_i, Y = y) \text{ [marginal]}$$

$$\Pr(y|x) = \frac{\Pr(X = x, Y = y)}{\sum_j \Pr(X = x, Y = y_j)} \text{ [conditional]}$$

### Law of iterated expectations

$$E(Y) = \sum_i E(Y|X = x_i) \Pr(X = x_i) = E[E(Y|X)]$$

### Mean square error (MSE)

Let  $\theta$  be an unknown parameter and  $T_n(\theta)$  an estimator for it.

$$\text{MSE}(T_n(\theta)) = E[(T_n(\theta) - \theta)^2]$$

### Sums of random variables

$$E(a + bX + cY) = a + b\mu_X + c\mu_Y$$

$$\text{var}(a + bX) = b^2\sigma_X^2$$

$$\text{var}(aX + bY) = a^2\sigma_X^2 + 2ab\sigma_{XY} + b^2\sigma_Y^2$$

$$\text{var}(Y_1 + \dots + Y_n) = \sum_i \text{var}(Y_i) + \sum_i \sum_{j \neq i} \text{cov}(Y_i, Y_j)$$

$$E(Y^2) = \sigma_Y^2 + \mu_Y^2 \quad E(XY) = \sigma_{XY} + \mu_X\mu_Y$$

$$\text{cov}(a + bX + cR, Y) = b\sigma_{XY} + c\sigma_{RY}$$

### Mean and Variance estimators

Y is a r.v. with mean and variance  $\mu_Y$  and  $\sigma_Y^2$

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^n Y_i \quad E(\bar{Y}) = \mu_Y \quad \text{var}(\bar{Y}) = \frac{\sigma_Y^2}{n}$$

$$\bar{\sigma}_Y^2 = \frac{1}{n} \sum_{i=1}^n (Y_i - \bar{Y})^2 \quad E(\bar{\sigma}_Y^2) = \frac{n-1}{n} \sigma_Y^2$$

$$S_Y^2 = \frac{1}{n-1} \sum_{i=1}^n (Y_i - \bar{Y})^2 \quad E(S_Y^2) = \sigma_Y^2$$

### P-value

Y is a r.v. with mean and variance  $\mu_Y$  and  $\sigma_Y^2$

$$H_0 : E(Y) = \mu_0 \quad H_1 : E(Y) \neq \mu_0$$

$$\text{p-value} = \Pr_{H_0} [|\bar{Y} - \mu_0| > |\bar{Y}^{\text{act}} - \mu_0|]$$

## Econometrics

### OLS estimators

$$\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{X}$$

$$\hat{\beta}_1 = \frac{\sum_i (X_i - \bar{X})(Y_i - \bar{Y})}{\sum_i (X_i - \bar{X})^2} = \frac{\text{COV}(X, Y)}{\text{VAR}(X)}$$

### Variance of OLS estimators

$$\sigma_{\hat{\beta}_0}^2 = \frac{1}{n} \frac{\text{var}(H_i u_i)}{[E(H_i^2)]^2} \quad \text{with } H_i = 1 - \frac{\mu_X}{E[X_i^2]} X_i,$$

$$\sigma_{\hat{\beta}_1}^2 = \frac{1}{n} \frac{\text{var}[(X_i - \mu_X)u_i]}{[\text{var}(X_i)]^2}.$$

### Standard errors of OLS estimators

$$\text{SE}(\hat{\beta}_0) = \sqrt{\frac{1}{n} \frac{\frac{1}{n-2} \sum_i \hat{h}_i^2 \hat{u}_i^2}{\frac{1}{n} [\sum_i \hat{h}_i^2]^2}} \quad \text{with } \hat{h}_i = 1 - \left( \frac{\bar{X}}{\frac{1}{n} \sum_i X_i^2} \right) X_i$$

$$\text{SE}(\hat{\beta}_1) = \sqrt{\frac{\frac{1}{n-2} \sum_i (X_i - \bar{X})^2 \hat{u}_i^2}{\left[ \frac{1}{n} \sum_i (X_i - \bar{X})^2 \right]^2}}$$