

Basic Skills Needed for Economic Forecasting

Mathematics

Statistics

Econometrics

Six rules of approximation for small numbers

I. Log $\ln(1+r) = r$

II. Product $(1+\pi)(1+g) \cong (1+\pi+g)$ $(1+g)^n \cong 1+ng$

III. Division $(1+r) = \frac{1+i}{1+\pi^e} \approx (1+i-\pi^e)$

IV. Growth of a product $Y = LW$ $g_Y = g_L + g_W$

V. Growth of a Ratio $y = \frac{Y}{L}$ $g_y = g_Y - g_L$

VI. Sum of a geometric Series $= 1 + X + X^2 + \dots + X^n = \frac{1 - X^{n+1}}{1 - X}$

Five Rules of log

I. Power Rule $Y = K^\alpha$ $\ln(Y) = \alpha \ln(K)$

II. Product Rule $R = PY$ $\ln(R) = \ln(Y) + \ln(P)$

III. Quotient Rule $y = \frac{Y}{L}$ $\ln(y) = \ln(Y) - \ln(L)$

IV. Log of Exponentials $Y_t = Y_0 e^{gt}$ $\ln(Y_t) = \ln(Y_0) + gt$

V. Differentiation of variable in log with respect to time:

$$\frac{d(\ln Y_t)}{dt} = \frac{dY_t}{Y_t} = g_y$$

Four Rules of Differentiation

I. Power Rule $Y = K^\alpha \quad \frac{\partial Y}{\partial K} = \alpha K^{\alpha-1}$

II. Product Rule $R = PY \quad dR = Y \times dP + P \times dY$

III. Quotient Rule $y = \frac{Y}{L} \quad dy = \frac{L \times dY - dL \times Y}{L^2} = \frac{dY}{L} - \frac{dL}{L} \frac{Y}{L}$

IV. Chain Rule $Y = L^{0.5} \quad L = 50 \left(\frac{w}{p} \right)^2 \quad Y = \left[50 \left(\frac{w}{p} \right)^2 \right]^{0.5}$

$$\frac{\partial Y}{\partial (w/p)} = \frac{\partial Y}{\partial L} \frac{\partial L}{\partial (w/p)} = 0.5 L^{-0.5} 50 \times 2 (w/p) = \left(\frac{50}{L} \right) (w/p)$$

Some Example of Matrix Operation

1. Find the determinant of the following matrix.

$$\text{a) } A = \begin{bmatrix} 2 & 1 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{bmatrix} \quad \text{b) } B = \begin{bmatrix} -7 & 0 & 3 \\ 9 & 1 & 4 \\ 0 & 6 & 5 \end{bmatrix}$$

(a)

$$|A| = \begin{vmatrix} 2 & 1 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{vmatrix} = (2 \cdot 5 \cdot 9) + (1 \cdot 6 \cdot 7) + (3 \cdot 8 \cdot 4) - (7 \cdot 5 \cdot 3) - (8 \cdot 6 \cdot 2) - (9 \cdot 4 \cdot 1) = 90 + 42 + 96 - 105 - 96 - 36 = 228 - 237 = -9$$

(b)

$$|B| = \begin{vmatrix} -7 & 0 & 3 \\ 9 & 1 & 4 \\ 0 & 6 & 5 \end{vmatrix} = (-7 \cdot 1 \cdot 5) + (0 \cdot 4 \cdot 0) + (3 \cdot 6 \cdot 9) - (0 \cdot 1 \cdot 3) - (6 \cdot 4 \cdot (-7)) - (5 \cdot 9 \cdot 0) = -35 + 0 + 162 - 0 + 168 - 0 = -35 + 330 = 295.$$

Sum and Multiplication of Vectors

$$\begin{aligned}
 e &= \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix} \quad (\text{T rows and 1 column}) \quad e'e = \begin{bmatrix} e_1 & e_2 & e_3 & e_4 & e_5 \end{bmatrix} \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix} = \sum_i^5 e_i^2; \\
 ee' &= \begin{bmatrix} e_1 \\ e_2 \\ e_3 \\ e_4 \\ e_5 \end{bmatrix} \begin{bmatrix} e_1 & e_2 & e_3 & e_4 & e_5 \end{bmatrix} = \begin{bmatrix} e_1^2 & e_1e_2 & e_1e_3 & e_1e_4 & e_1e_5 \\ e_2e_1 & e_2^2 & e_2e_3 & e_2e_4 & e_2e_5 \\ e_3e_1 & e_3e_2 & e_3^2 & e_3e_4 & e_3e_5 \\ e_4e_1 & e_4e_2 & e_4e_3 & e_4^2 & e_4e_5 \\ e_5e_1 & e_5e_2 & e_5e_3 & e_5e_4 & e_5^2 \end{bmatrix}
 \end{aligned}$$

Matrix Inversion

- If the determinant of a matrix is zero then that matrix is called a singular matrix. Singularity reflects linear dependence among explanatory variables.

1. Find the inverse of the following matrix

$$A = \begin{bmatrix} 4 & 1 & -1 \\ 0 & 3 & 2 \\ 3 & 0 & 7 \end{bmatrix}$$

Determinant of this matrix = $|A| = 99 \neq 0 \Rightarrow$ the inverse A^{-1} exists.

The cofactor matrix $C_{i,j} = (-1)^{i+j} |M_{i,j}|$ where $M_{i,j}$ are minors of each element.

$$C = \begin{bmatrix} \begin{vmatrix} 3 & 2 \\ 0 & 7 \end{vmatrix} & -\begin{vmatrix} 0 & 2 \\ 3 & 7 \end{vmatrix} & \begin{vmatrix} 0 & 3 \\ 3 & 0 \end{vmatrix} \\ -\begin{vmatrix} 1 & -1 \\ 0 & 7 \end{vmatrix} & \begin{vmatrix} 4 & -1 \\ 3 & 7 \end{vmatrix} & -\begin{vmatrix} 4 & 1 \\ 3 & 0 \end{vmatrix} \\ \begin{vmatrix} 1 & -1 \\ 3 & 2 \end{vmatrix} & -\begin{vmatrix} 4 & -1 \\ 0 & 2 \end{vmatrix} & \begin{vmatrix} 4 & 1 \\ 0 & 3 \end{vmatrix} \end{bmatrix} = \begin{bmatrix} 21 & 6 & -9 \\ -7 & 31 & 3 \\ 5 & -8 & 12 \end{bmatrix} \quad C' = Adj(A) = \begin{bmatrix} 21 & -7 & 5 \\ 6 & 31 & -8 \\ -9 & 3 & 12 \end{bmatrix}$$

$$\text{The desired inverse matrix is } A^{-1} = \frac{1}{|A|} Adj(A) = \frac{1}{99} \begin{bmatrix} 21 & -7 & 5 \\ 6 & 31 & -8 \\ -9 & 3 & 12 \end{bmatrix}$$

1. Solve following equations system using Cramer's rule

$$x_1 + 2x_2 + 2x_3 = 1 \quad (1)$$

$$2x_1 + 2x_2 + 3x_3 = 3 \quad (2)$$

$$x_1 - x_2 + 3x_3 = 5 \quad (3)$$

Answer: First trite this system of three equations and three unknowns in matrix format as following.

$$\begin{matrix} \begin{bmatrix} 1 & 2 & 2 \\ 2 & 2 & 3 \\ 1 & -1 & 3 \end{bmatrix} & \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} & = & \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix} & ; \\ \text{A} & \text{x} & & \text{b} & \end{matrix}$$

$$|A| = \begin{vmatrix} 1 & 2 & 2 \\ 2 & 2 & 3 \\ 1 & -1 & 3 \end{vmatrix} = 1 \cdot 2 \cdot 3 + 2 \cdot 3 \cdot 1 + 2 \cdot (-1) \cdot 2 - 2 \cdot 2 \cdot 1 - (-1) \cdot 3 \cdot 1 - 3 \cdot 2 \cdot 2 = 6 + 6 - 4 - 4 + 3 - 12 = 12 - 8 + 3 - 12 = -5$$

The solutions for x1, x2, and x3 is defined as following: $x_1 = \frac{\begin{vmatrix} 1 & 2 & 2 \\ 3 & 2 & 3 \\ 5 & -1 & 3 \end{vmatrix}}{|A|} = \frac{-5}{-5} = 1$

$$x_2 = \frac{\begin{vmatrix} 1 & 1 & 2 \\ 2 & 3 & 3 \\ 1 & 5 & 3 \end{vmatrix}}{|A|} = \frac{5}{-5} = -1 \quad x_3 = \frac{\begin{vmatrix} 1 & 2 & 1 \\ 2 & 2 & 3 \\ 1 & -1 & 5 \end{vmatrix}}{|A|} = \frac{-5}{5} = -1 \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ -1 \\ 1 \end{bmatrix}$$

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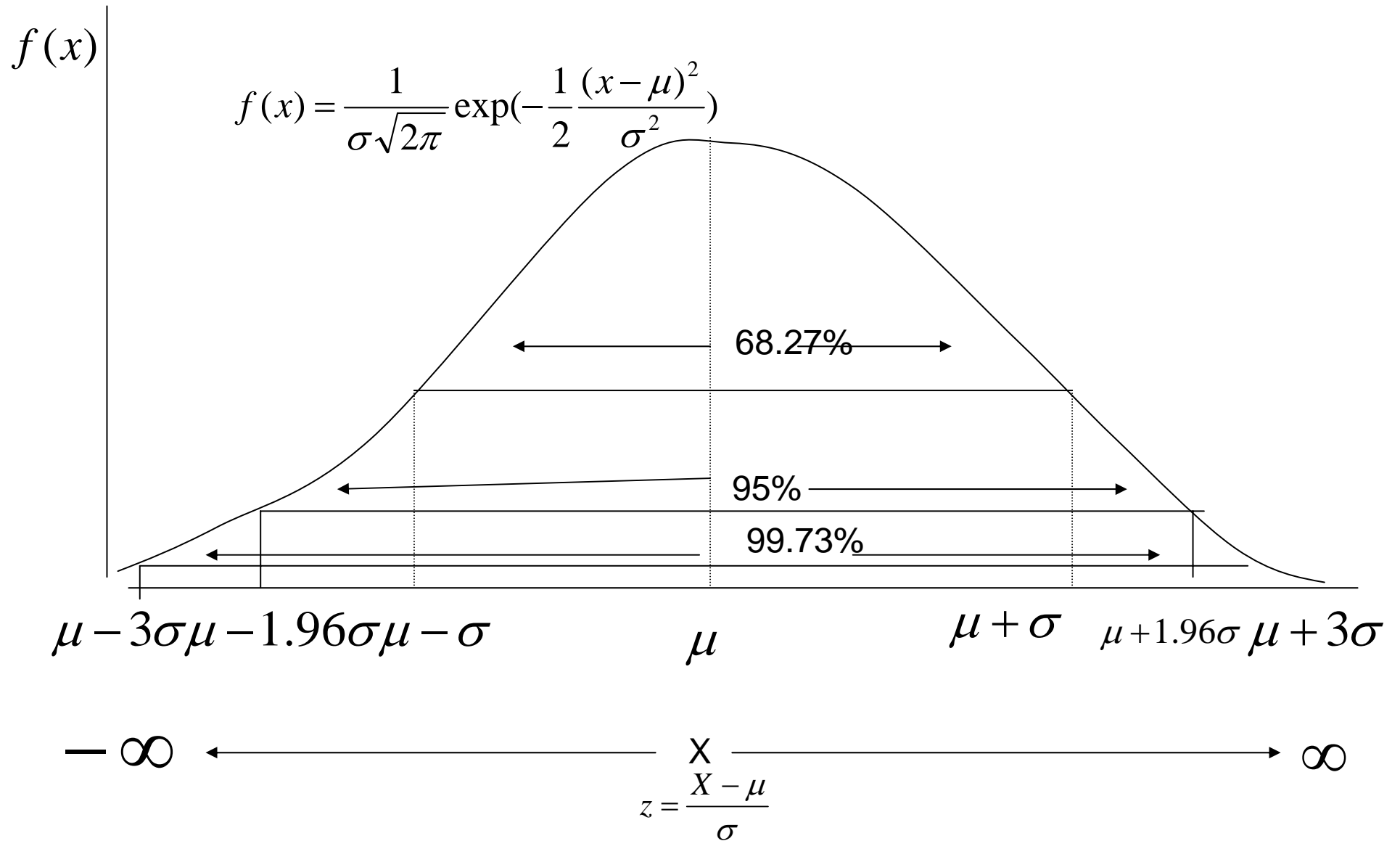
$$\begin{matrix} \begin{bmatrix} 1 & 2 & 2 \\ 2 & 2 & 3 \\ 1 & -1 & 3 \end{bmatrix} & \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} & = & \begin{bmatrix} 1 \\ 3 \\ 5 \end{bmatrix} & ; \\ \text{A} & \text{x} & & \text{b} & \end{matrix}$$

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Normal Distribution of X: Bell Shaped Distribution



Mean

$$\bar{X} = \frac{\sum X_i}{N}$$

Variance

$$\text{var}(X) = \frac{\sum (X_i - \bar{X})^2}{N - 1}$$

Standard Dev

$$s = \sqrt{\frac{\sum (X_i - \bar{X})^2}{N - 1}}$$

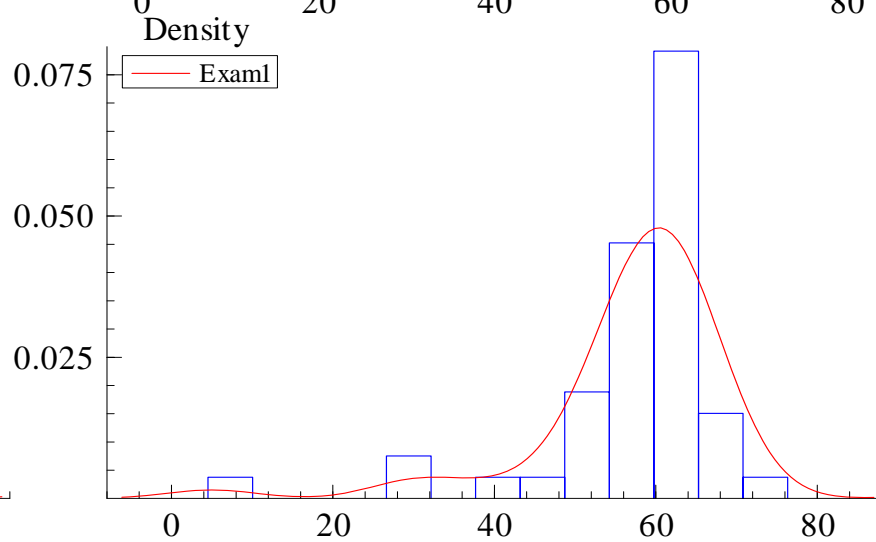
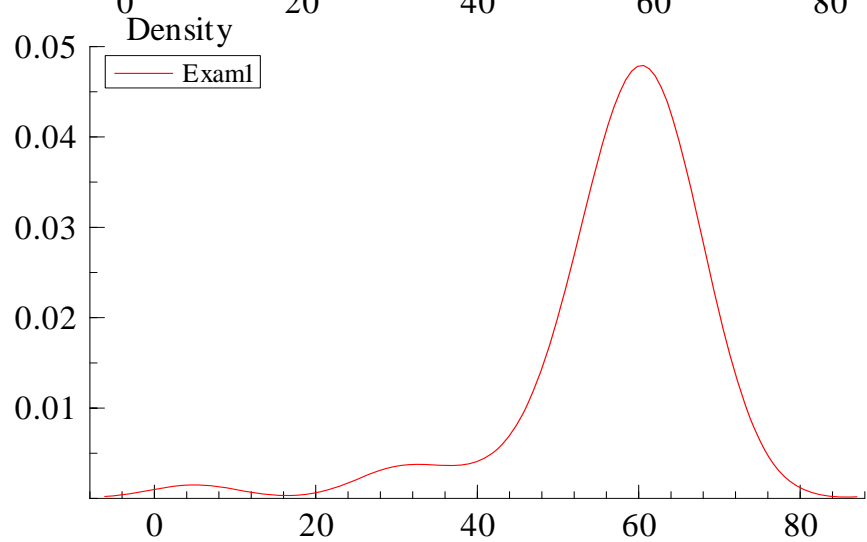
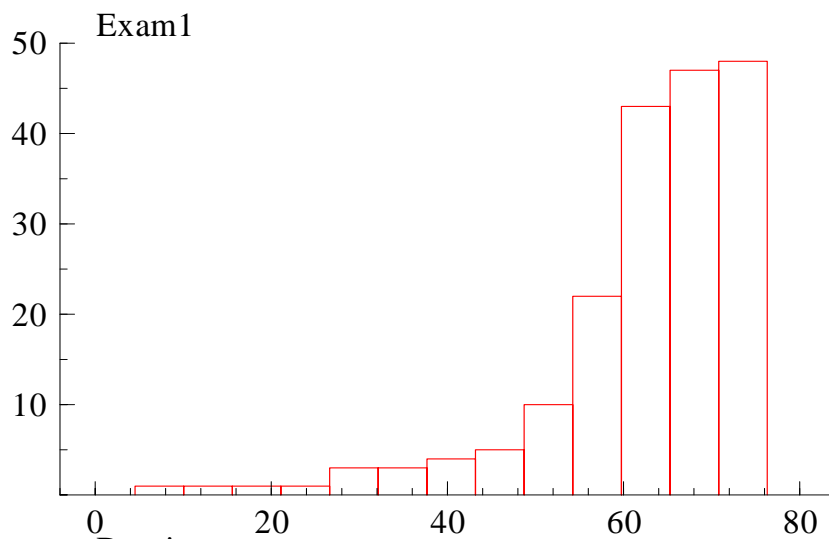
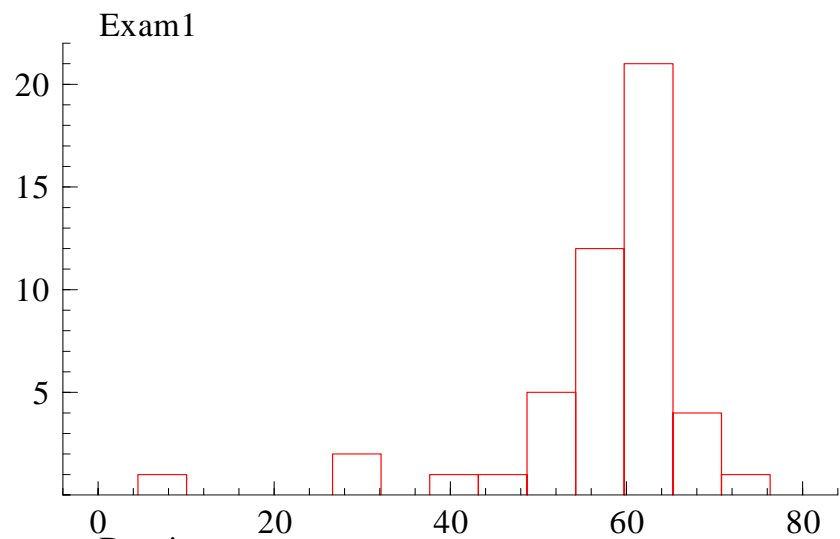
Covariance Correlation and Regression

$$\text{Cov}(X, Y) = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{N}$$

$$\rho_{x,y} = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sqrt{\text{var}(X) \text{Var}(Y)}}$$

$$-1 \leq \rho_{x,y} \leq 1$$

$$Y_i = \beta_1 + \beta_2 K_i + \beta_3 L_i + e_i$$



Properties of a Discrete Random Variable

$$P(x_i) \geq 0$$

$$\sum_{i=1}^n P(x_i) = 1$$

$$E[X_i] = \sum_i X_i P(x_i)$$

$$\text{var}[X_i] = \sum_i (X_i - E(X_i))^2 P(x_i)$$

Continuous Random variable: IQ scores in a certain Exam

Profits of firms

Age of individuals in a country

Income of individuals and households

Consumer's surplus for a given product

Distance travelled by cars in between two destinations

Amount of blood in ones body

Amount of water in a bath tub

$$f(x) = P[X = x_1] \geq 0$$

$$\int_{-\infty}^{\infty} f(x).dx = 1$$

$$\int_a^b f(x).dx = P(a \leq x \leq b)$$

Conditional Distributions

Mutually Exclusive and Exhaustive events vs. Dependent Events

$$f(x, y)$$

$$f(x) = \sum_y f(x, y)$$

$$f(y) = \sum_x f(x, y)$$

$$f(x/y) = \frac{f(x, y)}{f(y)}$$

$$f(y/x) = \frac{f(x, y)}{f(x)}$$

Type II Error: Accepting a Null Hypothesis When it is False

$f(\hat{\beta}_1)$

Type I error is more serious than type II error
But when type I error is minimised type II error raises

