

Simple Linear Regression

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Why Regression?

- Regression relates an **explanatory variable (X)** to a **response variable (Y)**.
- Example: Are changes in BMI associated with changes in systolic blood pressure?

Simple Linear Regression Model

$$Y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$$

- Y_i : Systolic blood pressure (response).
- x_i : BMI (predictor).
- β_0 : Intercept.
- β_1 : Slope – mean change in BP for a 1-unit increase in BMI.
- ε_i : Random error term.

Interpretation of Coefficients

$$\mu_j = 95 + 1.5x_j$$

- **Intercept (95):** Mean BP if BMI = 0 (not realistic).
- **Slope (1.5):** Each 1-unit increase in BMI is associated with a 1.5 mmHg increase in mean systolic BP.

True line:

$$\mu_i = 95 + 1.5x_i$$

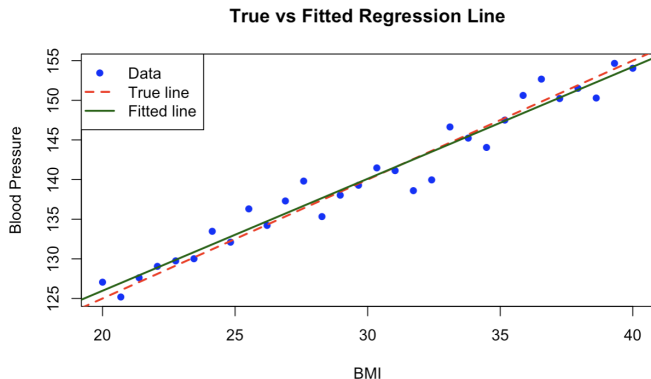
Estimated/Fitted line (from simulated sample):

$$\hat{\mu}_i = 97.72 + 1.41x_i$$

Simulated Dataset (First 10 Patients)

| BMI (kg/m^2) | BP (mmHg) |
|--------------------------------|-----------|
| 20.0 | 129.97 |
| 20.7 | 124.65 |
| 21.4 | 133.55 |
| 22.1 | 143.33 |
| 22.8 | 126.80 |
| 23.4 | 127.83 |
| 24.1 | 146.99 |
| 24.8 | 139.92 |
| 25.5 | 128.58 |
| 26.2 | 139.74 |

True vs Fitted Regression Line



- Red dashed line: true regression line.
- Green solid line: fitted regression line.
- Blue points: simulated data.

- OLS estimates minimize the sum of squared errors.

$$\hat{\beta}_1 = \frac{\sum(x_i - \bar{x})(y_i - \bar{y})}{\sum(x_i - \bar{x})^2}$$

$$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

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- $\hat{\beta}_1$: estimated slope
- $\hat{\beta}_0$: estimated intercept

Properties of OLS Estimator

- **Unbiasedness:** $E[\tilde{\beta}_0] = \beta_0$, $E[\tilde{\beta}_1] = \beta_1$.
- **Efficiency:** OLS gives the *Best Linear Unbiased Estimator (BLUE)* where "best" implies minimum variance.

Summary & Questions

- Simple linear regression links a predictor to an outcome.
- Interpretation: intercept, slope (mean change).
- OLS formulas for $\hat{\beta}_0$ and $\hat{\beta}_1$.
- Properties $\tilde{\beta}_0, \tilde{\beta}_1$: unbiased, efficient.

Questions?