

Evaluating Regression Models Ahmed Eleish Data Analytics ITWS/MGMT-4600/6600, CSCI-4600, BCBP-4600

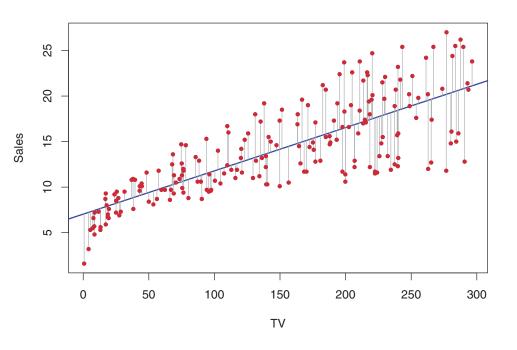
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Tetherless World Constellation Rensselaer Polytechnic Institute



Evaluating Linear Models

- Sales vs. TV ad spending
- Sales in 1000s of units
- TV ad spending in 1000s of \$





Evaluating Linear Models

	Coefficient	Std. error	t-statistic	p-value
Intercept	7.0325	0.4578	15.36	< 0.0001
TV	0.0475	0.0027	17.67	< 0.0001

Values of coefficients >> their Std. errors

High t-statistic

Very low p-value

$$t = \frac{\hat{\beta}_1 - 0}{\operatorname{SE}(\hat{\beta}_1)}$$





Residual Standard Error

• Mean sales $\approx 14,000$ units

R^2

- measures the proportion of the variability in Y that can be explained using X
- has a value between 0,1

Quantity	Value
Residual standard error	3.26
R^2	0.612
F-statistic	312.1

RSE =
$$\sqrt{\frac{1}{n-2}}$$
RSS = $\sqrt{\frac{1}{n-2}\sum_{i=1}^{n}(y_i - \hat{y}_i)^2}$

$$R^2 = \frac{TSS - RSS}{TSS} = 1 - \frac{RSS}{TSS}$$

$$TSS = \sum (y_i - \bar{y})^2$$





Residual Sum of Squares (RSS)

For given data $(x_1,y_1), ..., (x_n,y_n) \in \mathbb{R} \times \mathbb{R}$,

- Residual Sum of Squares (RSS), the ith residual $e_i = y_i - \hat{y}_i$

$$RSS = e_1^2 + e_2^2 + \dots + e_n^2$$





Mean Absolute Error

Mean(||Predicted value - Real value||)

MAE =
$$\frac{\sum_{i=1}^{n} |y_i - \hat{y}_i|}{n} = \frac{\sum_{i=1}^{n} |e_i|}{n}$$



Mean Squared Error

• Mean((Predicted value - Real value)²)

MSE =
$$\frac{1}{n} \sum_{i=1}^{n} (y_i - \hat{y}_i)^2$$





Root Mean Squared Error

SquareRoot(Mean((Predicted value - Real value)²))

$$RMSE = \sqrt{\frac{\sum_{i=1}^{n} (y_i - \overline{y_i})^2}{n}}$$



Thanks!



