

# $R^2$ Score

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- Shows how well our chosen variable predict the target variable

$$R^2 \text{ Score} = \frac{1 - SS_{\text{res}}}{SS_{\text{tot}}}$$



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*"How much the predicted values deviate from the actual values"*

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*"Measures the total variance in the dependent variable"*





How is  $R^2$  Score interpreted?

# $R^2$ Score

R2 Score of the model = **72%**



Model **explains 72% variability** in the dependent variable



Model fails to **explain 28% variability** in the dependent variable

**Note:** Closer the  $R^2$  Score is to 1, better the model explains the variability



# $R^2$ Score

## OVERFIT MODEL

$R^2$  Score



On addition of **irrelevant features** to the data

?





Adjusted  $R^2$  Score

# $R^2$ Score

A modified version of  $R^2$  Score that accounts for overfitting by penalizing excessive features in a model.

$$\text{Adjusted } R^2 \text{ Score} = 1 - \left[ \frac{(1 - R^2) \times (n-1)}{(n - k - 1)} \right]$$

**$R^2$** : The  $R^2$  of the model

**n**: The number of observations

**k**: The number of predictor variables

