#### What is Simple Linear Regression?

Simple Linear Regression is a method to model the relationship between two continuous variables - one independent variable (X) and one dependent variable (Y). The model is expressed as: Y = mX + c + e.

## What are the key assumptions of Simple Linear Regression?

- 1. Linearity
- 2. Independence
- 3. Homoscedasticity
- 4. Normality
- 5. No multicollinearity

## What does the coefficient m represent in the equation Y = mX + c?

m (the slope) indicates the change in Y for a one-unit change in X.

# What does the intercept c represent in the equation Y = mX + c?

c is the predicted value of Y when X is 0.

#### How do we calculate the slope m in Simple Linear Regression?

$$m = ((Xi - X)(Yi - )) / ((Xi - X)^2)$$

#### What is the purpose of the least squares method in Simple Linear Regression?

It minimizes the sum of squared differences between observed and predicted values.

# How is the coefficient of determination (R^2) interpreted in Simple Linear Regression?

 $R^2$  indicates the proportion of variance in Y explained by X.  $R^2 = 1$  means perfect fit,  $R^2 = 0$  means no explanatory power.

#### What is Multiple Linear Regression?

It uses two or more independent variables to predict a dependent variable. Y = b0 + b1X1 + b2X2 + ... + bnXn + e

# What is the main difference between Simple and Multiple Linear Regression?

Simple uses 1 independent variable, Multiple uses 2 or more.

# What are the key assumptions of Multiple Linear Regression?

- 1. Linearity
- 2. Independence
- 3. Homoscedasticity
- 4. Normality
- 5. No multicollinearity

# What is heteroscedasticity, and how does it affect the results?

It means non-constant variance of residuals, which can lead to inefficient estimates.

## How can you improve a model with high multicollinearity?

- 1. Remove correlated predictors
- 2. PCA
- 3. Ridge/Lasso regression
- 4. Combine variables

# Techniques for transforming categorical variables?

- 1. One-hot encoding
- 2. Label encoding
- 3. Binary encoding
- 4. Ordinal encoding

#### What is the role of interaction terms?

They capture the combined effect of two variables, not captured individually.

## Interpretation of intercept in Simple vs. Multiple Linear Regression?

In Simple LR: Y when X = 0. In Multiple LR: Y when all predictors = 0.

## Significance of the slope?

It shows the effect of a 1-unit change in X on Y.

#### How does the intercept provide context?

It gives the baseline value of Y when predictors are 0.

#### Limitations of R^2?

- 1. Doesn't account for overfitting
- 2. Doesn't imply causation
- 3. Ignores complexity
- 4. Misleading in non-linear models

## Interpret a large standard error for a coefficient?

Indicates high variability and low confidence in the estimate.

#### Identifying heteroscedasticity in residual plots?

Look for fanning or narrowing of residuals. It's important to address for valid inferences.

## High R^2 but low adjusted R^2?

Indicates overfitting due to non-informative predictors.

## Why scale variables?

For faster convergence, unbiased coefficient interpretation, and essential for regularization.

## What is polynomial regression?

Regression using polynomial terms to fit non-linear data.

## How does it differ from linear regression?

Linear fits a straight line, polynomial fits a curve.

#### When is it used?

When the relationship is non-linear but can be modeled with a polynomial.

## **General equation?**

$$Y = b0 + b1X + b2X^2 + ... + bnX^n + e$$

# Can it be applied to multiple variables?

Yes, called multivariate polynomial regression.

# Limitations of polynomial regression?

- 1. Overfitting
- 2. Poor extrapolation
- 3. Computationally expensive
- 4. Collinearity

# Evaluating model fit for polynomial degree?

- 1. Cross-validation
- 2. Adjusted R^2

- 3. AIC/BIC
- 4. Residual plots

# Why is visualization important?

To detect overfitting/underfitting, understand model behavior, and validate curve fit.

# How is it implemented in Python?

Use sklearn's PolynomialFeatures with LinearRegression. See code sample in full answer.