

Correlation between the Independent data

$$\text{Total Sqf} = F1 \text{ Sqf} + F2 \text{ Sqf} + \text{Underg Sqf}$$

Keep Total Sqf or keep 3 column
Multi Collinearity

Polynomial regression creates many new feature
 $X_1 \rightarrow X_1^2, X_1^3$

If duplicated or redundant, will py remove them

No, Not remove anything

Polynomial $\xrightarrow{\text{Multiplicate}}$

Problems of Multi Collinearity

1. unstable Coefficients
2. large swing in the Coefficient values
3. inflated variance

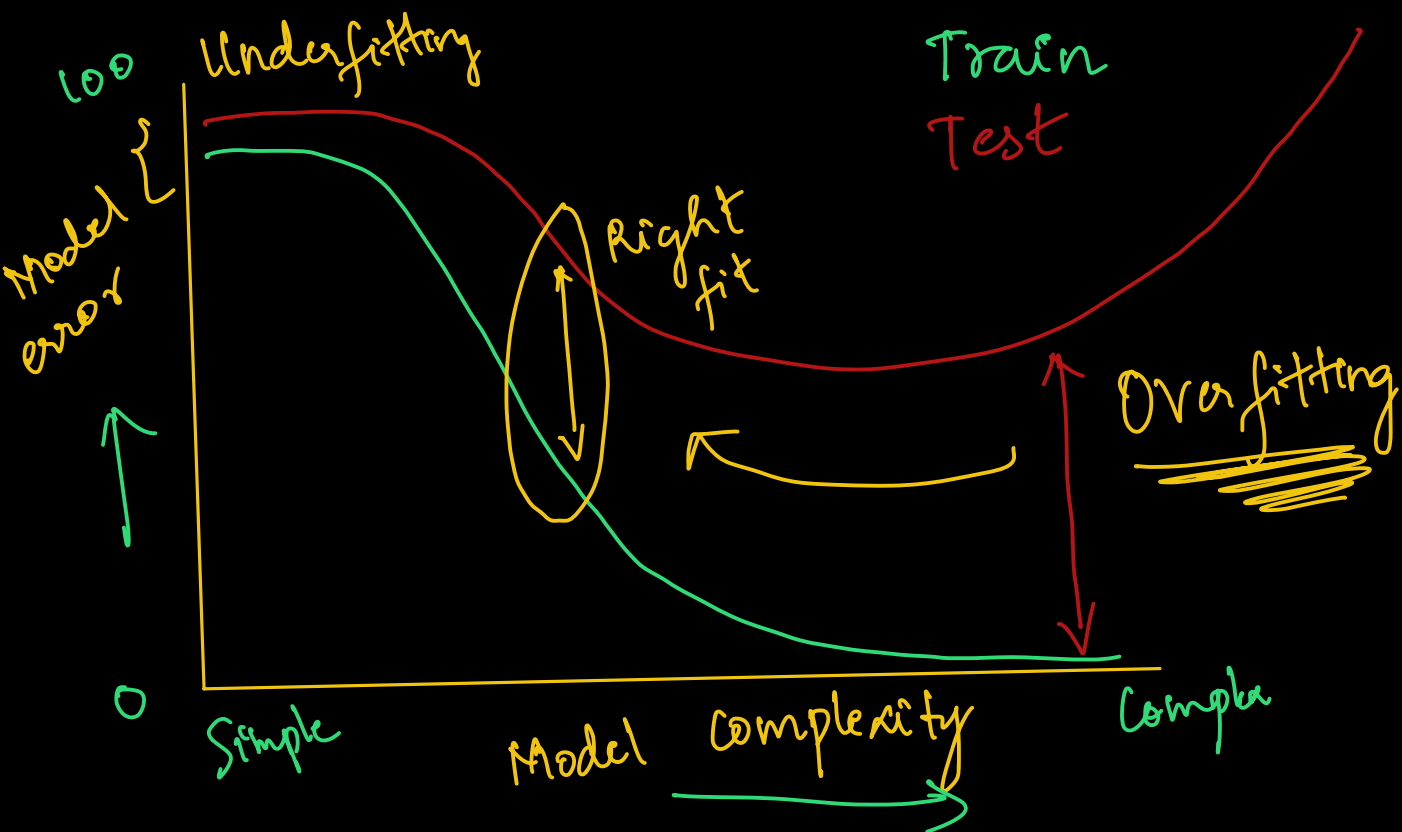
4. Poor model generalization

Linear Regression + Polynomial features

Overfit

avoid

Regularization!



Overfitting

Co-efficients

→ high ✓



Gradient descent

{ find the Co-efficient + intercept
Constraint → $\frac{\text{Cost function}}{\text{MSE}}$ ↓ minimal

linear regression

Regularization

{ find the Co-efficient + intercept
Constraint → $\frac{\text{Cost function}}{\text{MSE}}$ ↓ minimal

Penalty factor

MSE + ✓

Co-efficient ↓ minimal

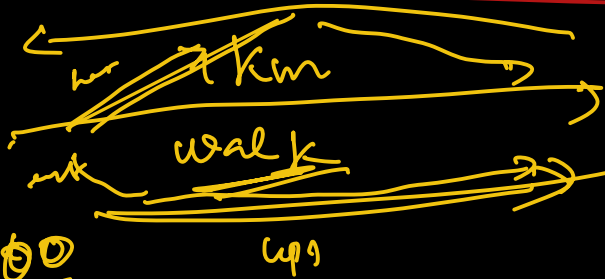
Shoppery

\$ 500

Shop

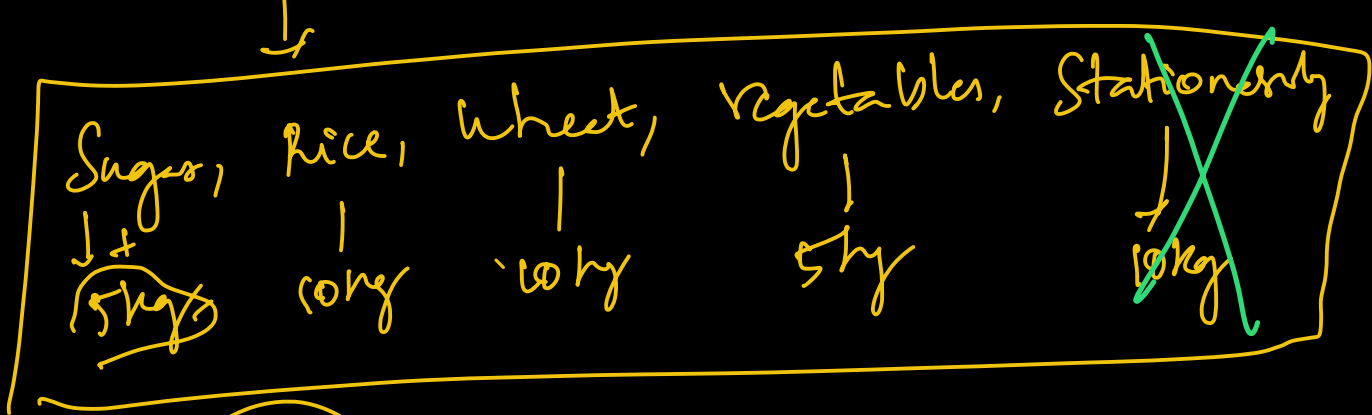
→ monthly

Multiple Products



Home





Sticky
 ↓
Penalize = carry and walk Penalty

↓

Sugar	+	Rice	+	Wheat		Vegetables	Stationery
1 kg		2 kg		1 kg		1 kg	2 kg

1 + 2 + 1 + 1 + 2

7 kg

Car → driving

↓

150 W/m

for → 300 hr

h → 200 km/hr

Face see 200 km/hr

Penalty \rightarrow
 so \rightarrow 120 km/h \rightarrow \$10000

$$\text{Cost function} = \frac{\sum (y - (\beta_0 + \beta_1 x))^2}{n} \quad | \quad \text{LR}$$

Regularization

$$\text{Cost function} = \frac{\sum (y - (\overset{\uparrow \lambda}{\beta_0} + \overset{\uparrow \lambda}{\beta_1} x))^2}{n} + \underbrace{\lambda \sum_{i=0}^n (\overset{\uparrow \lambda}{\beta_i})^2}_{\substack{\lambda = 0.1 \\ \lambda = 2 \\ \text{penalty factor}}}$$

1. Ridge Regression
2. Lasso Regression
3. Elastic Net Regression

1. Ridge Regression L_2

$$\text{Cost function} = \frac{\sum (y - (\beta_0 + \beta_1 x))^2}{n} + \lambda \sum_{i=0}^n (\beta_i^2)$$

\Downarrow
 Shrink all the coeff
 don't make anything

2. Lasso Regression L1 MSE

$$\text{Cost function} = \frac{\sum (y - (\beta_0 + \beta_1 x))^2}{n}$$

Feature Selection

$$+ \lambda \sum_{i=0}^n |\beta_i|$$

shrink all the coeffs
+
makes unimportant
coeff zero

3. Elastic Net

$$\text{Cost func} = \frac{\sum (y - (\beta_0 + \beta_1 x))^2}{n} +$$

$$\lambda \left[(\alpha) \sum |\beta| + (1-\alpha) \sum (\beta)^2 \right]$$

Gradient Descent

Cost function = MSE

find m and b

MSE ↓ minimal

Derivative

↓ $y = 2x^2 + 2x + 5$ ✓

↓
What is value of x
where y is minimum

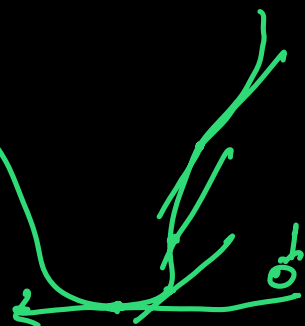
↳ Derivative ✓

$$y = 2x^2 + 2x + 5 + [1x + 3]$$

✓

$$= 2x^2 + 3x + 8$$

↘ w



$$MSE = \frac{\sum (y - \hat{y})^2}{n} \quad \checkmark$$

$$\hat{y} = \beta_0 + \beta_1 x$$

Partial
Derivate

$$MSE = \frac{\sum (y - \hat{y})^2}{n} + \sum (\beta_0^2 + \beta_1^2) \quad \checkmark$$

$$\beta_0 = \frac{100}{1}$$

$$\beta_1 = \frac{50}{1}$$

$$+ 10,000 + 2,500$$

$$12,500$$

$$12,500 + 150$$



