Multiple Linear Regression

- Simple Linear Regression predict Response for a single predictor X, for example TV
- Multiple Linear Regression deals with multiple predictors, even fitting separate Simple regression to each predictor X this will make us miss some key correlations and associations between predictors and Response

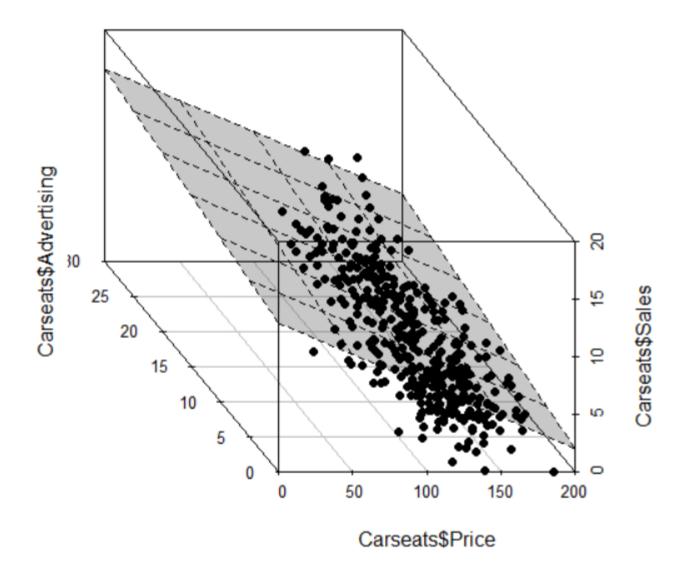
$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \dots + \beta_p X_p + \varepsilon$$

For example :

$$sales = \beta_0 + \beta_1 TV + \beta_2 Radio + \beta_3 Newspaper$$

Geometric Interpretation (Regression Plane)

Regression Plane



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• Unlike <u>Simple Linear Regression</u> Multiple Linear regression have multiple predictors which is draw as a hyperplan

Estimating The Regression Coefficients:

- The Coefficients in the multiple regression are unknown $eta_0,eta_1\dots,eta_p$
- So we estimate them as as in the <u>Simple Linear Regression</u>

$$\hat{y}=\hat{eta_0}+\hat{eta_1}x_1+\hat{eta_2}x_2+\cdots+\hat{eta}_px_p$$

Multiple Linear Regression Matrix Form:

• The multiple Regression formula can be written in a matrix form making it better to work with and derive the Coefficients

Y=Xeta+arepsilon

With:

• $Y \rightarrow$ Vector of dependent variables Response

• $X o \mathsf{Matrix}$ of n*p dimensions + **intercept** β_0

• $\beta \rightarrow$ Vector of Coefficients (to be estimated)

• $\varepsilon o extstyle extsty$

$$Y = egin{bmatrix} Y_1 \ Y_2 \ dots \ Y_n \end{bmatrix} = egin{bmatrix} 1 & X_{11} & X_{21} & \dots & X_{p1} \ 1 & X_{12} & X_{22} & \dots & X_{p2} \ dots & dots & dots & \ddots & dots \ 1 & X_{1n} & X_{2n} & \dots & X_{pn} \end{bmatrix} egin{bmatrix} eta_0 \ eta_1 \ eta_2 \ dots \ eta_n \end{bmatrix} + egin{bmatrix} arepsilon_1 \ eta_2 \ dots \ eta_n \end{bmatrix}$$

ullet E(arepsilon)=0

 $ullet \ var(arepsilon) = \sigma^2 I_{n*n}$

• $\beta's$ are called partial regression coefficients cause β_1 is the expected change in Y per Unite change in X_1 , While holding other X's constant