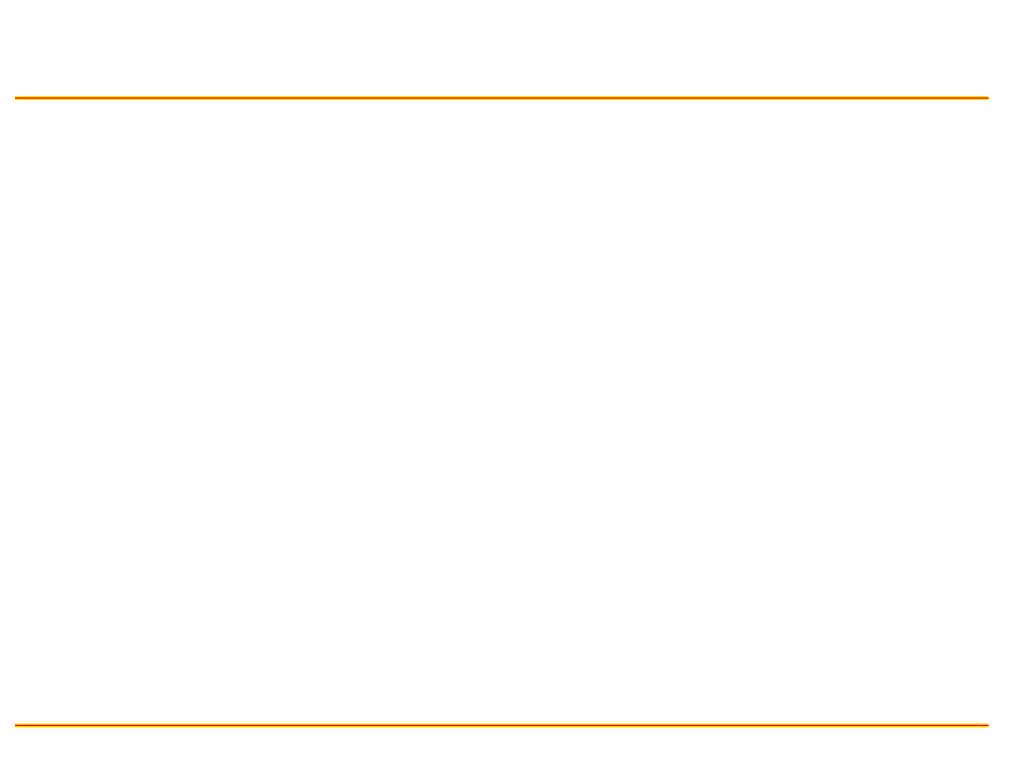
# **Building the Regression Model**

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## **Outline**





# **Data Collection and preparation**

- How the data are collected? (Design, Nature of the study)
- Data consistency. (Plots and numerical summaries, logical relations)
- Is this data analysis-ready? (Format checking, file conversion, etc.)
- GIGO (Garbage In; Garbage Out.)

# **Objectives**

- Reduction of explanatory or predictor variables Find parsimonious model with good explanatory/prediction power. Trade-off.
- Model refinement and selection Choosing from many "good" models, checking the adequacy of the models, sensitivity of the models, fixing the weak spots.
- Model Validation

### Selection-I.1

• 
$$\mathbf{R}_p^2 = 1 - \frac{SSE_p}{SSTO}$$
.

#### Selection-I.2

▶ AIC: Akaike's information criterion  $AIC_p = -2 \ln likelihood + 2p \propto n \ln SSE_p - n \ln n + 2p$ . ID models with smaller AIC.

#### **Selection-III**

**General Linear Test Approach** Given  $Y | X_1, \dots, X_2$ 

Should  $X_3, \dots, X_5$  be added?.

Testing  $H_0$ : Reduced vs  $H_1$ : Full

- Fit the *full* model (Y  $|X_1, \dots, X_5$ ) and get SSE(F),  $df_F$
- Fit the *reduced* model (Y  $|X_1, \dots, X_2|$ ) and get SSE(R),  $df_R$
- Calculate

$$\mathbf{F} = \frac{SSE(R)}{df_R} \frac{SSE(F)}{df_F} / \frac{SSE(F)}{df_F} \sim \mathbf{F}_{df_R} df_{F,df_F}$$

Then perform a level test.

#### Comments

- No easy, clear-cut way to ID the best model
- Usually, many "good" models rather than one best model
- Respect the hierarchy of models
  Higher order terms < lower order terms
  (X<sup>4</sup> < X<sup>1</sup>)
  Interaction terms < main effect terms
  (X<sub>1</sub>X<sub>2</sub> < X<sub>1</sub> or X<sub>2</sub>)
- Chapter 10 Variable Selection of Faraway, J. (2002).
   Also his Chapter 11 is highly recommended

# Improper functional form of a prediction

- Goal: Detect the suitable form of Y vs  $X_q$  while  $X_1, \dots, X_{q-1}$  in the model.
- Partial Regression Plots:

$$\begin{array}{l} \mathbf{e}(\mathbf{Y}\,|\mathbf{X}_1,\cdots,\mathbf{X}_{q-1}) \text{ vs. } \mathbf{e}(\mathbf{X}_q|\mathbf{X}_1,\cdots,\mathbf{X}_{q-1}). \\ \\ \mathbf{e}(\mathbf{Y}\,|\mathbf{X}_1,\cdots,\mathbf{X}_{q-1}) \text{: residual of } \mathbf{Y} \text{ regresses on } \\ \mathbf{X}_1,\cdots,\mathbf{X}_{q-1} \\ \\ \mathbf{e}(\mathbf{X}_q|\mathbf{X}_1,\cdots,\mathbf{X}_{q-1}) \text{: residual of } \mathbf{X}_q \text{ regresses on } \\ \mathbf{X}_1,\cdots,\mathbf{X}_{q-1} \end{array}$$

Why bother?

#### **Outliers-I**

The model (fitted) shouldn't be affect by just few points.

- LSE is EXTREMELY sensitive to outliers. ExampX0.
- Detection: Residual-based tests and pXots towards outlying Y. Why? What to expect?

Semistudentized residual: Same scale (Naive).

$$\mathbf{e}_i^* = rac{e_i}{\sqrt{MSE}}$$
 ,  $\mathbf{e}_i = \mathbf{Y}_i - \hat{\mathbf{Y}_i}$ 

Studentized residual: In the same scale (Refined).

$$\mathbf{r}_i = rac{e_i}{\sqrt{MSE(1-h_{ii})}}$$
 since

$$^{2}(\mathbf{e}) = ^{2}(\mathbf{I} - \mathbf{H}), \quad \mathbf{H} = \mathbf{X}(\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'.$$

Deleted Residuals: With or Without You. Outlying Y.

$$\mathbf{d}_i = \mathbf{Y}_i - \hat{\mathbf{Y}_{i(i)}}$$

#### **Outliers-II**

Studentized Deleted Residual:

$$\mathbf{t}_i = rac{d_i}{s(d_i)}$$
 where  $\mathbf{s}(\mathbf{d}_i) = \sqrt{\mathbf{MSE}_{(i)}(1 - \mathbf{h}_{ii})}$ 

■ Hat matrix Leverage values → Outlying X

$$0 \le \mathbf{h}_{ii} \le 1$$
,  $n \atop i=1$   $\mathbf{h}_{ii} = \mathbf{p}$ .

## Afluential obs

$$\textbf{DFFITS})_i = \frac{\widehat{Y_i} - \widehat{Y_{i(i)}}}{\sqrt{\text{MSE}_{(i)}h_{ii}}} \text{ Flag: If } |\text{DFFITS}| > 1 \text{ for small/merchium 13433]} > 1$$

# **Multicollinearity: VIF**

- Problems of MLCL: X, Extra SSR, s(^), nonsignificance
- Informal Diagnosis
  - Sensitive incl/exclud of X or data
  - Nonsignificance on important predictors
  - Wrong sign of estimated
  - Large coefficient in  $\mathbf{r}_{XX}$ , Large  $\mathbf{R}^2$  among  $\mathbf{X}$
  - Wide confidence intervals of
- Variation Inflation Factor (TL<sup>-1</sup>) V IF<sub>k</sub> diagonal entry of  $\mathbf{r}_{XX}$ .

$$(\mathbf{VIF})_k = (1 - \mathbf{R}_k^2)^{-1},$$

 $\mathbf{R}_k^2$ :  $\mathbf{R}^2$  of  $\mathbf{X}_k$  regressing on the other  $\mathbf{X}'\mathbf{s}$ .

Flag: Larger than 10 or  $\gg V \bar{I} F$ 

#### **Model Validation**

- Estimation/Fit the past; Predict the future
- Consistency with New Data
- Comparison with theoretical expectation, earlier empirical and simulation results
- Cross-Validation: Use of a holdout sample to check the model and predictive ability.

#### What's next?