# College Study

#### Mia Yu

### **Data Overview**

We explore the application of nonlinear models to analyze the "College" dataset, comprising statistics from 565 US colleges as reported in a past issue of US News and World Report. The predictors are

predictors	Explanation					
Apps	umber of applications received					
Accept	Number of applications accepted					
Enroll	Number of new students enrolled					
Top10perc	Pct. new students from top 10% of H.S. class					
Top25perc	Pct. new students from top 25% of H.S. class					
F.Undergrad	Number of fulltime undergraduates					
P.Undergrad	Number of parttime undergraduates					
Room.Board	Room and board costs					
Books	Estimated book costs					
Personal	Estimated personal spending					
PhD	Pct. of faculty with Ph.D.'s					
Terminal	Pct. of faculty with terminal degree					
S.F.Ratio	Student/faculty ratio					
perc.alumni	Pct. alumni who donate					
Expend	Instructional expenditure per student					
Grad.Rate	Graduation rate					

For more information, you can go to U.S.News.

data(College)
skimr::skim(College)

Table 2: Data summary

Name	College
Number of rows	777
Number of columns	18
Column type frequency:	
factor	1
numeric	17
Group variables	None

Variable type: factor

skim_variable	n_missing	complete_rate	ordered	n_unique	top_counts
Private	0	1	FALSE	2	Yes: 565, No: 212

#### Variable type: numeric

skim_variablen_	_missing compl	ete_rate	e mean	sd	p0	p25	p50	p75	p100	hist
Apps	0	1	3001.64	3870.20	81.0	776.0	1558.0	3624.0	48094.0	
Accept	0	1	2018.80	2451.11	72.0	604.0	1110.0	2424.0	26330.0	
Enroll	0	1	779.97	929.18	35.0	242.0	434.0	902.0	6392.0	
Top10perc	0	1	27.56	17.64	1.0	15.0	23.0	35.0	96.0	
Top25perc	0	1	55.80	19.80	9.0	41.0	54.0	69.0	100.0	
F.Undergrad	0	1	3699.91	4850.42	139.0	992.0	1707.0	4005.0	31643.0	
P.Undergrad	0	1	855.30	1522.43	1.0	95.0	353.0	967.0	21836.0	
Outstate	0	1	10440.67	4023.02	2340.0	7320.0	9990.0	12925.0	21700.0	
Room.Board	0	1	4357.53	1096.70	1780.0	3597.0	4200.0	5050.0	8124.0	
Books	0	1	549.38	165.11	96.0	470.0	500.0	600.0	2340.0	
Personal	0	1	1340.64	677.07	250.0	850.0	1200.0	1700.0	6800.0	
PhD	0	1	72.66	16.33	8.0	62.0	75.0	85.0	103.0	
Terminal	0	1	79.70	14.72	24.0	71.0	82.0	92.0	100.0	
S.F.Ratio	0	1	14.09	3.96	2.5	11.5	13.6	16.5	39.8	
perc.alumni	0	1	22.74	12.39	0.0	13.0	21.0	31.0	64.0	
Expend	0	1	9660.17	5221.77	3186.0	6751.0	8377.0	10830.0	56233.0	
Grad.Rate	0	1	65.46	17.18	10.0	53.0	65.0	78.0	118.0	

Partition the dataset into two parts: training data (80%) and test data (20%).

```
set.seed(123)
total_rows <- nrow(College)
train_indices <- sample(1:total_rows, size = 0.8 * total_rows)
train_data <- College[train_indices, ]
test_data <- College[-train_indices, ]</pre>
```

# **Model Fitting**

Fit smoothing spline models to predict out-of-state tuition (Outstate) using the percentage of alumni who donate (perc.alumni) as the only predictor, across a range of degrees of freedom.

```
df_results <- data.frame()

for(df in seq(2, 10, by=1)){
   fit.ss <- smooth.spline(train_data$perc.alumni, train_data$Outstate, df=df)

   pred.ss <- predict(fit.ss, x = test_data$perc.alumni)

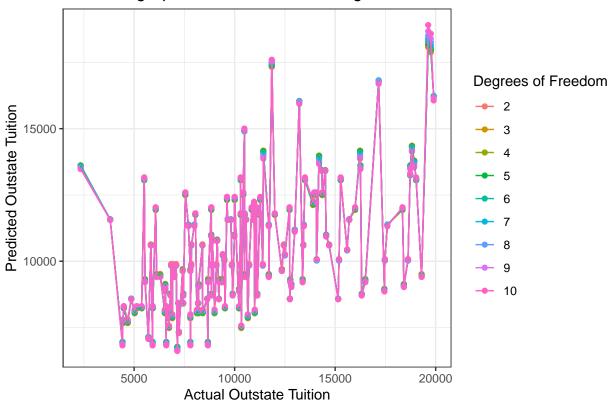
   pred.ss.df <- data.frame(real = test_data$Outstate, pred = pred.ss$y, df = df)
   df_results <- rbind(df_results, pred.ss.df)
}</pre>
```

Plot the model fits for each degree of freedom.

```
ggplot(df_results, aes(x = real, y = pred, colour = factor(df))) +
  geom_point() +
```

```
geom_line(aes(group = df)) +
theme_bw() +
labs(title = "Smoothing Spline Fits for Different Degrees of Freedom",
    x = "Actual Outstate Tuition",
    y = "Predicted Outstate Tuition",
    colour = "Degrees of Freedom")
```

## Smoothing Spline Fits for Different Degrees of Freedom



The obtained result of 2.000214 suggests that the optimal degree of freedom is approximately 2, indicating that the model is relatively linear. As the degree of freedom increases, the model transitions from linear to more curved, but too high a degree of freedom may lead to overfitting, meaning the model complexity is too high, reflecting the random noise in the data rather than the true trend. This degree of freedom is recommended by generalized cross-validation (GCV) and is automatically chosen by the smooth.spline() function, to demonstrate the relationship between the percentage of alumni who donate and the out-of-state tuition.