

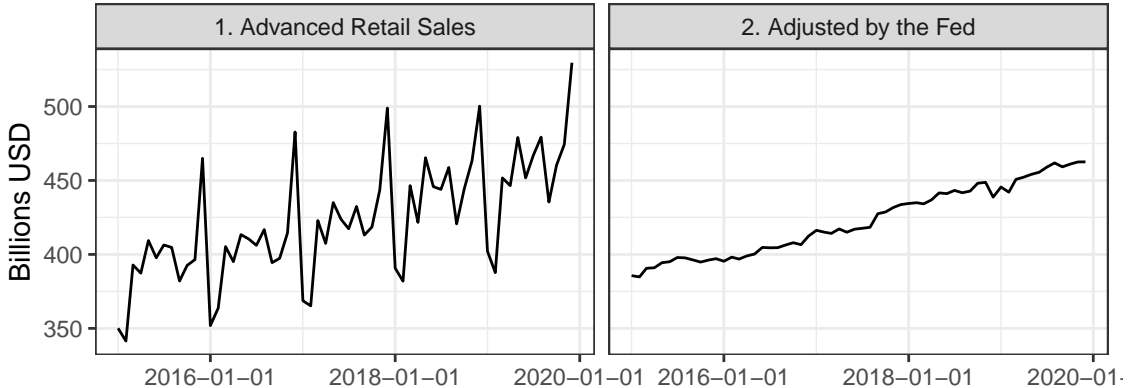
# Today's Agenda

Fitting and evaluating linear trend models with seasonality effects

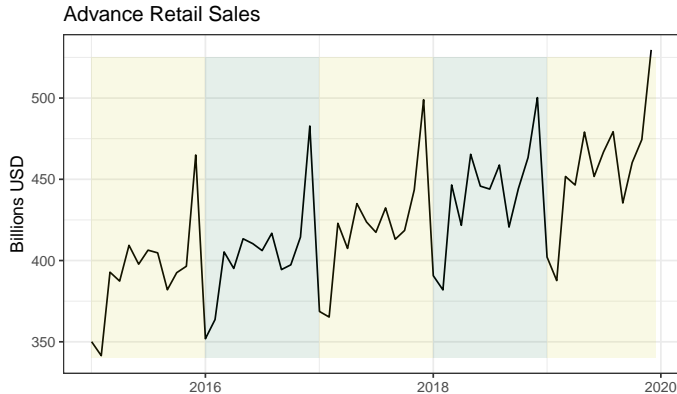
- Dataset: Advanced Retail Sales

Justin Leinaweaver (Spring 2022)

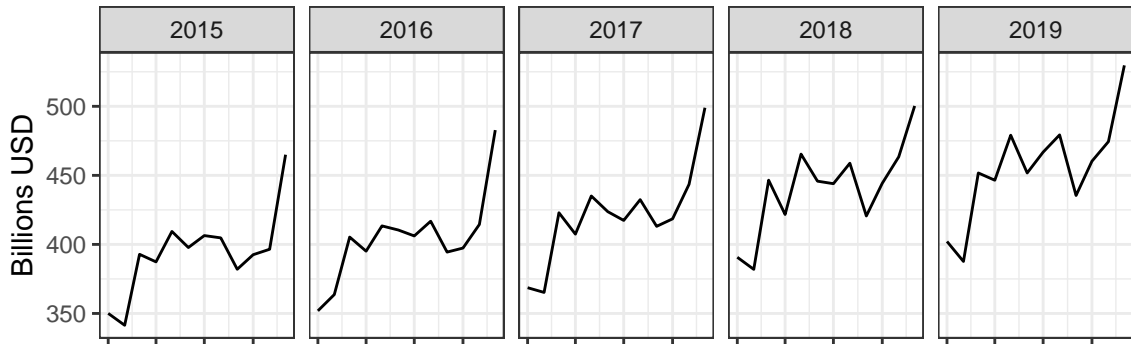
date	year	month	advance_retail_sales	advance_retail_sales_adj
16436	2015	1	350.067	385.672
16467	2015	2	341.459	384.783
16495	2015	3	392.848	390.642



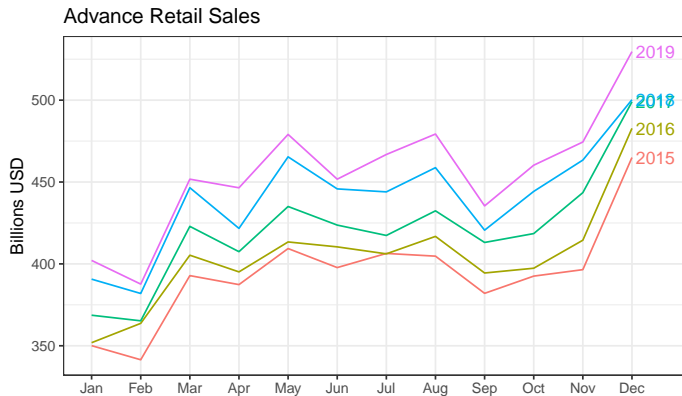
"Seasonality is a characteristic of a time series in which the data experiences regular and predictable changes that recur every calendar year" (Investopedia 2020).

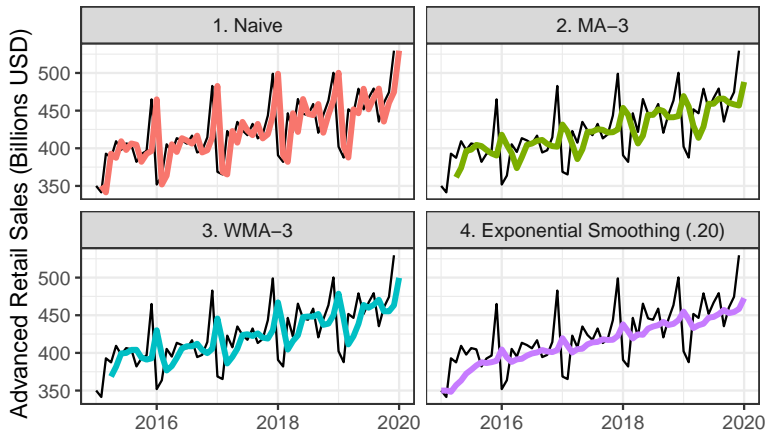


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Tools	MSE
Naive	1629
MA-3	1289
WMA-3	1310
ExpSmth (.2)	1075

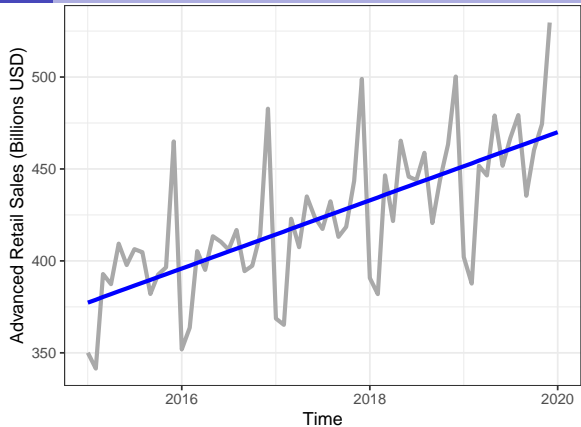
# Model 1

## **Regress advanced retail sales on time period**

- ① Fit the model (Time period = 1:60)
- ② Visualize the model (line plot)
- ③ Predict the next 12 months

	Retail Sales
Time	1.54* (0.22)
Constant	375.76* (7.62)
Observations	60
Adjusted R <sup>2</sup>	0.46
Residual Std. Error	29.13 (df = 58)
F Statistic	50.61* (df = 1; 58)

Note: \*p<0.05



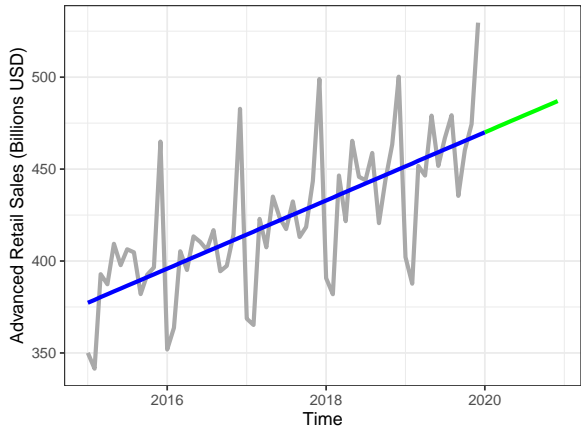
$$\text{Sales} = 375.76 + 1.54 \times \text{Time}$$

- Time = 61, 62, 63, 64, ...



	Retail Sales
Time	1.54* (0.22)
Constant	375.76* (7.62)
Observations	60
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Residual Std. Error	29.13 (df = 58)
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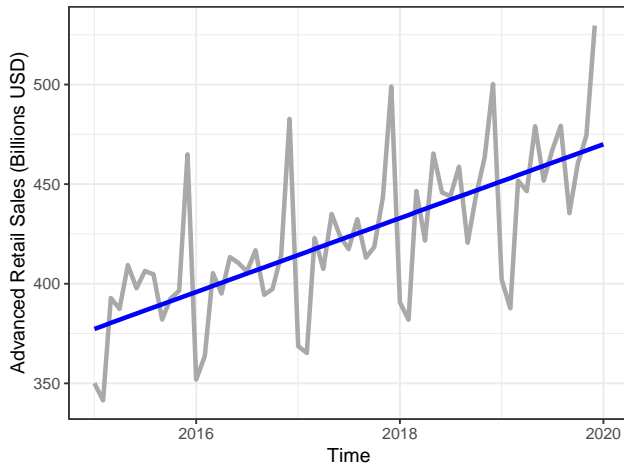
Note: \*p<0.05



1 2 3 4 5 6 7 8 9 10 11 12  
 470 472 473 475 476 478 479 481 482 484 485 487

# Fitting Linear Trend Models with OLS

Tools	MSE
Naive	1629
WMA-3	1310
MA-3	1289
ExpSmth (.2)	1075
OLS Time	820



## Model 2: Regress advanced retail sales on time period and season dummies

- 1 Fit the model
  - Time period = 1:60
  - Spring = '1' if Apr, May, Jun
  - Summer = '1' if Jul, Aug, Sep
  - Fall = '1' if Oct, Nov, Dec
- 2 Visualize the model (line plot)
- 3 Predict the next 12 months

# Fitting Linear Trend Models with OLS

	A	B	C	D	E	F	G	H	
1	date	year	month	advance_retail_sales	Time	Spring	Summer	Fall	
2	2015-01-01	2015	1	350.067	1				
3	2015-02-01	2015	2	341.459	2				
4	2015-03-01	2015	3	392.848	3				
5	2015-04-01	2015	4	387.352	4				
6	2015-05-01	2015	5	409.376	5				
7	2015-06-01	2015	6	397.752	6				
8	2015-07-01	2015	7	406.393	7				
9	2015-08-01	2015	8	404.729	8				
10	2015-09-01	2015	9	382.02	9				
11	2015-10-01	2015	10	392.545	10				
12	2015-11-01	2015	11	396.49	11				
13	2015-12-01	2015	12	464.962	12				

# Fitting Linear Trend Models with OLS

	date	year	month	advance_retail_sales	Time	Spring	Summer	Fall	
1	2015-01-01	2015	1	350.067	1	0			
2	2015-02-01	2015	2	341.459	2	0			
3	2015-03-01	2015	3	392.848	3	0			
4	2015-04-01	2015	4	387.352	4	1			
5	2015-05-01	2015	5	409.376	5	1			
6	2015-06-01	2015	6	397.752	6	1			
7	2015-07-01	2015	7	406.393	7	0			
8	2015-08-01	2015	8	404.729	8	0			
9	2015-09-01	2015	9	382.02	9	0			
10	2015-10-01	2015	10	392.545	10	0			
11	2015-11-01	2015	11	396.49	11	0			
12	2015-12-01	2015	12	464.962	12	0			
13	2016-01-01	2016	1	351.80	13				

# Fitting Linear Trend Models with OLS

	A	B	C	D	E	F	G	H	
1	date	year	month	advance_retail_sales	Time	Spring	Summer	Fall	
2	2015-01-01	2015	1	350.067	1	0	0		
3	2015-02-01	2015	2	341.459	2	0	0		
4	2015-03-01	2015	3	392.848	3	0	0		
5	2015-04-01	2015	4	387.352	4	1	0		
6	2015-05-01	2015	5	409.376	5	1	0		
7	2015-06-01	2015	6	397.752	6	1	0		
8	2015-07-01	2015	7	406.393	7	0	1		
9	2015-08-01	2015	8	404.729	8	0	1		
10	2015-09-01	2015	9	382.02	9	0	1		
11	2015-10-01	2015	10	392.545	10	0	0		
12	2015-11-01	2015	11	396.49	11	0	0		
13	2015-12-01	2015	12	464.962	12	0	0		
14	2016-01-01	2016	1	351.80	13				

# Fitting Linear Trend Models with OLS

	A	B	C	D	E	F	G	H
1	date	year	month	advance_retail_sales	Time	Spring	Summer	Fall
2	2015-01-01	2015	1	350.067	1	0	0	0
3	2015-02-01	2015	2	341.459	2	0	0	0
4	2015-03-01	2015	3	392.848	3	0	0	0
5	2015-04-01	2015	4	387.352	4	1	0	0
6	2015-05-01	2015	5	409.376	5	1	0	0
7	2015-06-01	2015	6	397.752	6	1	0	0
8	2015-07-01	2015	7	406.393	7	0	1	0
9	2015-08-01	2015	8	404.729	8	0	1	0
10	2015-09-01	2015	9	382.02	9	0	1	0
11	2015-10-01	2015	10	392.545	10	0	0	1
12	2015-11-01	2015	11	396.49	11	0	0	1
13	2015-12-01	2015	12	464.962	12	0	0	1
14	2016-01-01	2016	1	351.80	13			

## Model 2: Regress advanced retail sales on time period and season dummies

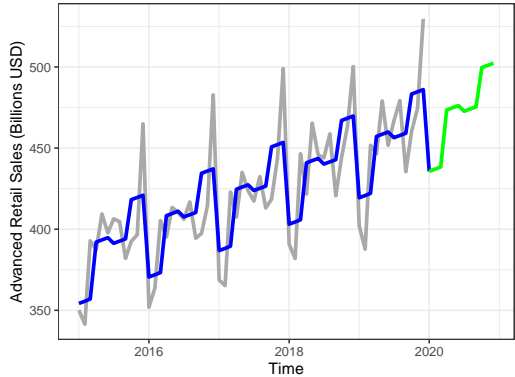
- 1 Fit the model
  - Time period = 1:60
  - Spring = '1' if Apr, May, Jun
  - Summer = '1' if Jul, Aug, Sep
  - Fall = '1' if Oct, Nov, Dec
- 2 Visualize the model (line plot)
- 3 Predict the next 12 months



	Retail Sales
Time	1.36* (0.17)
Spring	33.73* (8.43)
Summer	28.89* (8.47)
Fall	51.73* (8.56)
Constant	352.90* (7.49)

Observations	60
Adjusted R <sup>2</sup>	0.66
Residual Std. Error	23.03 (df = 55)
F Statistic	29.70* (df = 4; 55)

Note: \*p<0.05

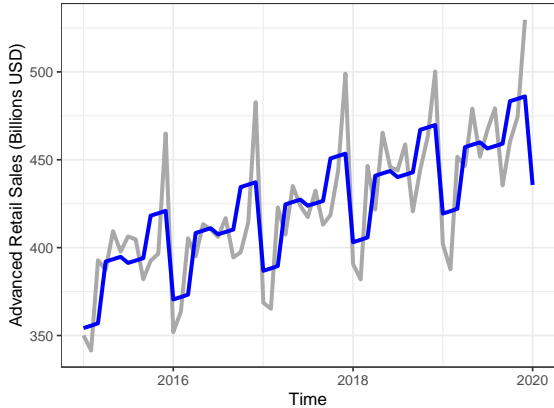


## Predictions

c(436, 437, 438, 473, 475, 476, 473, 474,  
475, 500, 501, 502)

# Fitting Linear Trend Models with OLS

Tools	MSE
Naive	1629
WMA-3	1310
MA-3	1289
ExpSmth (.2)	1075
OLS Time	820
OLS Time and Season	486



## Model 3: Regress advanced retail sales on time period and monthly dummies

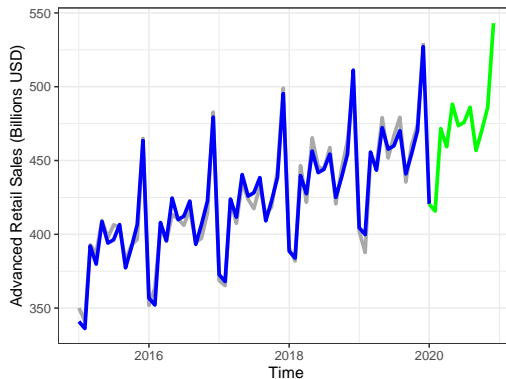
- 1 Fit the model
  - Time period = 1:60
  - Month dummies (x 11)
- 2 Visualize the model (line plot)
- 3 Predict the next 12 months

# Fitting Linear Trend Models with OLS

	A	B	C	D	E	F	G	H	I
1	date	year	month	advance_retail_sales	Time	Feb	Mar	Apr	May
2	2015-01-01	2015	1	350.067	1	0	0	0	0
3	2015-02-01	2015	2	341.459	2	1	0	0	0
4	2015-03-01	2015	3	392.848	3	0	1	0	0
5	2015-04-01	2015	4	387.352	4	0	0	1	0
6	2015-05-01	2015	5	409.376	5	0	0	0	1
7	2015-06-01	2015	6	397.752	6	0	0	0	0
8	2015-07-01	2015	7	406.393	7	0	0	0	0
9	2015-08-01	2015	8	404.729	8	0	0	0	0
10	2015-09-01	2015	9	382.02	9	0	0	0	0
11	2015-10-01	2015	10	392.545	10	0	0	0	0
12	2015-11-01	2015	11	396.49	11	0	0	0	0
13	2015-12-01	2015	12	464.962	12	0	0	0	0
14	2016-01-01	2016	1	351.88	13				

	Retail Sales
Time	1.33* (0.05)
Constant	339.55* (3.31)
Observations	60
Adjusted R <sup>2</sup>	0.97
Residual Std. Error	6.81 (df = 47)
F Statistic	161.59* (df = 12; 47)

Note: \*p<0.05

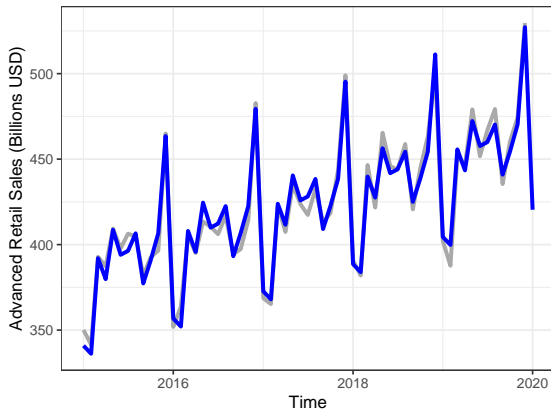


Predictions = 420, 416, 472, 459, 488, 474, 476, 486, 457, 470, 486, 543

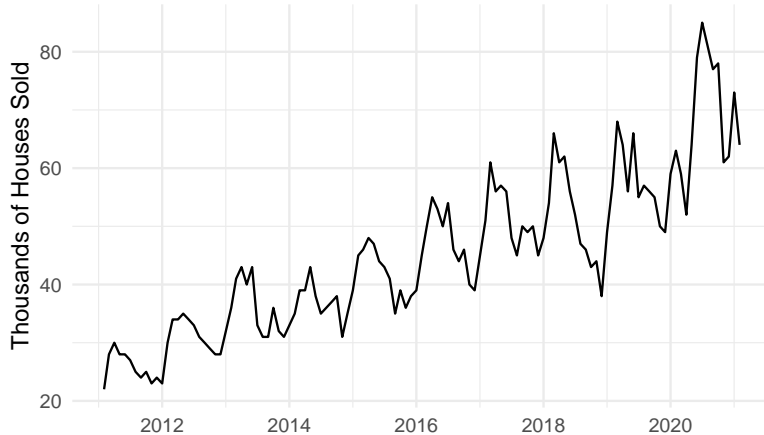
- Month coefficients omitted from the table.

# Fitting Linear Trend Models with OLS

Tools	MSE
Naive	1629
WMA-3	1310
MA-3	1289
ExpSmth (.2)	1075
OLS Time	820
OLS Time and Season	486
OLS Time and Month	36



# Predict the next 12 months of new home sales

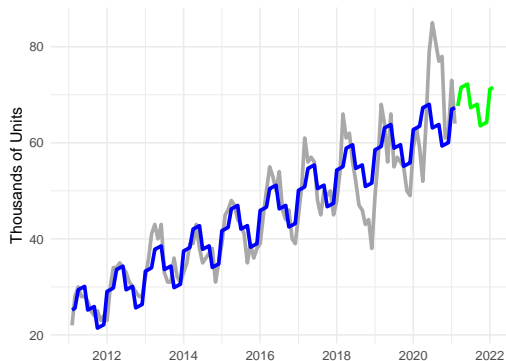


**Regress new home sales on time and season dummies**

## New Home Sales

Time	0.35* (0.01)
Spring	3.50* (1.44)
Summer	-1.76 (1.44)
Fall	-6.58* (1.44)
Constant	24.84* (1.35)
Observations	121
Adjusted R <sup>2</sup>	0.83
Residual Std. Error	5.62 (df = 116)
F Statistic	151.41* (df = 4; 116)

Note: \*p<0.05





New Home Sales	
Time	0.35* (0.01)
Spring	3.50* (1.44)
Summer	-1.76 (1.44)
Fall	-6.58* (1.44)
Constant	24.84* (1.35)
Observations	121
Adjusted R <sup>2</sup>	0.83
Residual Std. Error	5.62 (df = 116)
F Statistic	151.41* (df = 4; 116)

Note: \*p<0.05

Time	Date	Predictions
122	2021-03-01	68
123	2021-04-01	72
124	2021-05-01	72
125	2021-06-01	72
126	2021-07-01	67
127	2021-08-01	68
128	2021-09-01	68
129	2021-10-01	64
130	2021-11-01	64
131	2021-12-01	64
132	2022-01-01	71
133	2022-02-01	72