

INDR 450/550

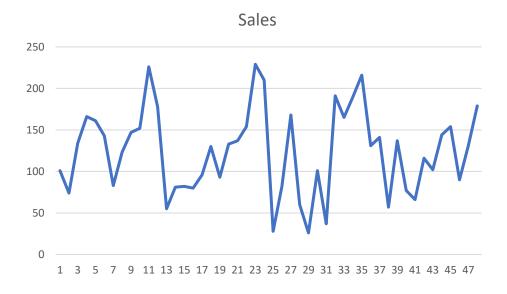
Spring 2022

Lecture 24: Mini Case (Prescriptive Analytics)

May 18, 2022

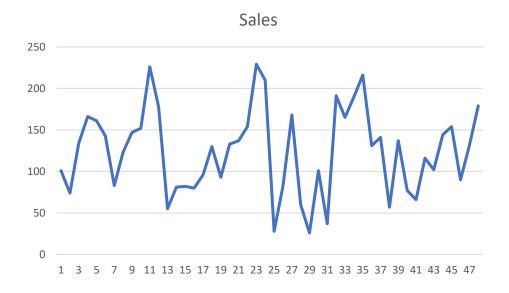
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- Here's a short prescriptive analytics case study.
- Implementation by Bijan Bibak (many thanks)
- We'll consider the monthly sales of Minis in Turkey



- Data from 2016 to beginning of 2021.
- A few complications: no data recorded for 2018 and 2019, we'll skip those two years.

• We'll consider the monthly inventory ordering problem: set the target inventory each month, assuming c_u =10, c_o = 2.



- There's no major visible trend or seasonality (except for the month of December).
- Let us assume that the sample corresponds to i.i.d. Observations.
- We can then use Sample Average Approximation.
- We'll solve the corresponding stochastic optimization problem.
- As usual, we randomly separate the data to a training set and a test set.

We solve the below LP using Gurobi Solver.

$$\min_{Q} R(Q) = \frac{1}{n} \sum_{i=1}^{n} c_{u} (d_{i} - Q)^{+} + c_{o}(Q - d_{i})^{+}$$

$$\equiv \min_{Q} \frac{1}{n} \sum_{i=1}^{n} c_{u} z_{i}^{+} + c_{o} z_{i}^{-}$$
s.t.
$$z_{i}^{+} \geq d_{i} - Q \quad i = 1, 2, ..., n$$

$$z_{i}^{-} \geq Q - d_{i} \quad i = 1, 2, ..., n$$

$$z_{i}^{+}, z_{i}^{-} \geq 0 \quad i = 1, 2, ..., n$$

Results of Sample Average Approximation:

```
Empirical Risk Minimization (train) ------
Objective function is: 152.667
Optimal Order is: 165.0
Empirical Risk Minimization (test) -----
Average cost is: 194.727
```

- Next, we try a joint estimation and optimization formulation using the following predictors:
- t: to capture the effect of trend (if any)
- Dec: a binary variable for the month of December
- Lag 1, 2, 3: The change in sales at lags 1, 2 and 3.

	Α	В	С	D	Е	F	G	Н
1	Month	Sales	t	Dec	Lag1	Lag2	Lag3	
2	1/1/2016	101	1	0				
3	2/1/2016	74	2	0				
4	3/1/2016	134	3	0	-27			
5	4/1/2016	166	4	0	60	-27		
6	5/1/2016	161	5	0	32	60	-27	
7	6/1/2016	143	6	0	-5	32	60	
8	7/1/2016	83	7	0	-18	-5	32	
9	8/1/2016	123	8	0	-60	-18	-5	
10	9/1/2016	147	9	0	40	-60	-18	
11	10/1/2016	152	10	0	24	40	-60	
12	11/1/2016	226	11	0	5	24	40	
13	12/1/2016	178	12	1	74	5	24	
14	1/1/2017	55	13	0	-48	74	5	
15	2/1/2017	81	14	0	-123	-48	74	
4.0	2/4/2247			_				

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• We then solve the following optimization problem:

$$\min_{Q(\cdot)} R(Q(\cdot), S_n) = \frac{1}{n} \sum_{i=1}^n c_u (d_i - Q(x_i))^+ + c_o (Q(x_i) - d_i)^+
\equiv \min_{Q = (q^1, q^2, \dots q^p)} \frac{1}{n} \sum_{i=1}^n c_u z_i^+ + c_o z_i^-
s.t.$$

$$z_i^+ \geq d_i - \left(q^0 + \sum_{j=1}^p q^j x_i^j\right) \quad i = 1, 2, \dots, n$$

$$z_i^- \geq q^0 + \sum_{j=1}^p q^j x_i^j - d_i \quad i = 1, 2, \dots, n$$

$$z_i^+, z_i^- \geq 0 \quad i = 1, 2, \dots, n$$

$$Q(x) = Q((x^1, x^2, \dots, x^p)) = q^0 + \sum_{j=1}^p q^j x^j$$

Here are the results when we use all five predictors:

```
Feature Based Newsvendor (train) ------ Objective function is: 108.27052529494726

Average Order Quantity is: 160.84910722851487

Feature Based Newsvendor (test) ------

Average cost is: 344.65686264558195

Average Order Quantity is: 131.0496792694867
```

$$Q(\mathbf{x}) = Q((x^1, x^2, \dots, x^p)) = q^0 + \sum_{j=1}^p q^j x^j$$

 q^0 : 195.319, q^1 : -1.562, q^2 : -14.894, q^3 : 0.245 q^4 : 0.574, q^5 : 0.535

It appears that there is significant overfitting here!

Here are the results when we use only the dummy for December

```
Feature Based Newsvendor (train) ------
Objective function is: 149.273
Average Order is: 162.030
Feature Based Newsvendor (test) ------
Average cost is: 173.273
Average Order is: 164.09
        Q(x) = Q(x^1) = q^0 + q^1 x^1
    q^0: 161, q^1: 17,
```

Very simple ordering rule: order 178 in December and 161 in all other months!