

# Business and Data Understanding

## What decisions need to be made?

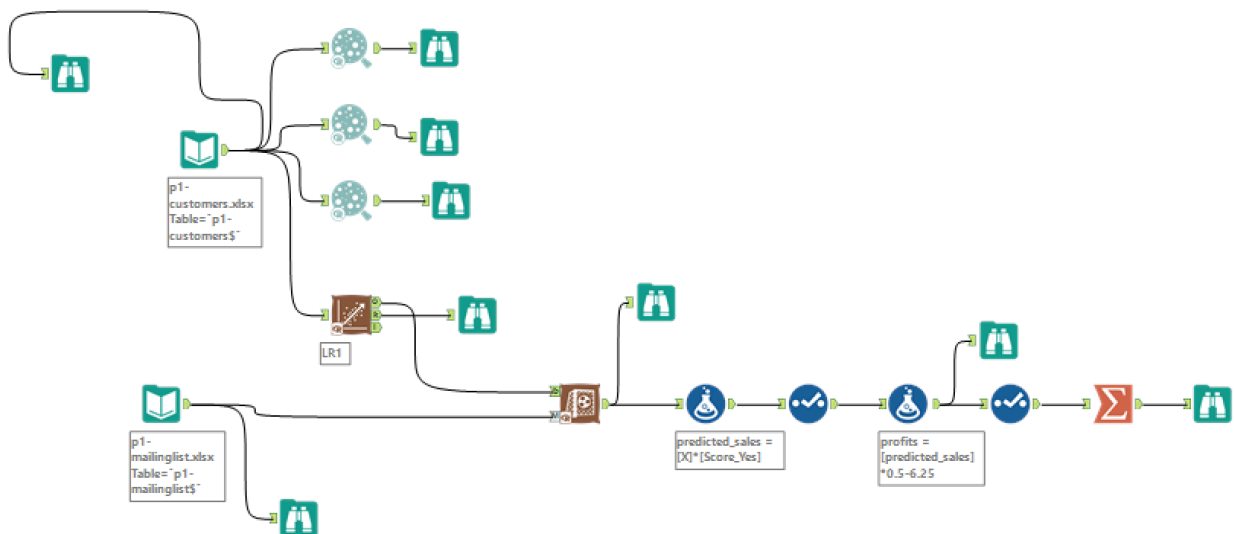
公司想依据现有的数据，预测一下如果给250位顾客派发新的产品册能否带来一定收入，从而决定是否进行派发。

Based on the existing data, the company would like to predict whether the distribution of new product catalog to 250 customers will bring certain profits, so as to decide whether to distribute them.

## What data is needed to inform those decisions?

- 已有的顾客数据，来建立预测模型
- 通过线性回归模型，预测这250个顾客的可能花费，求出总和即可
- Existing customer data, to establish a prediction model
- Through the linear regression model, we can predict the possible costs of these 250 customers and calculate the sum

## Analysis, Modeling, and Validation



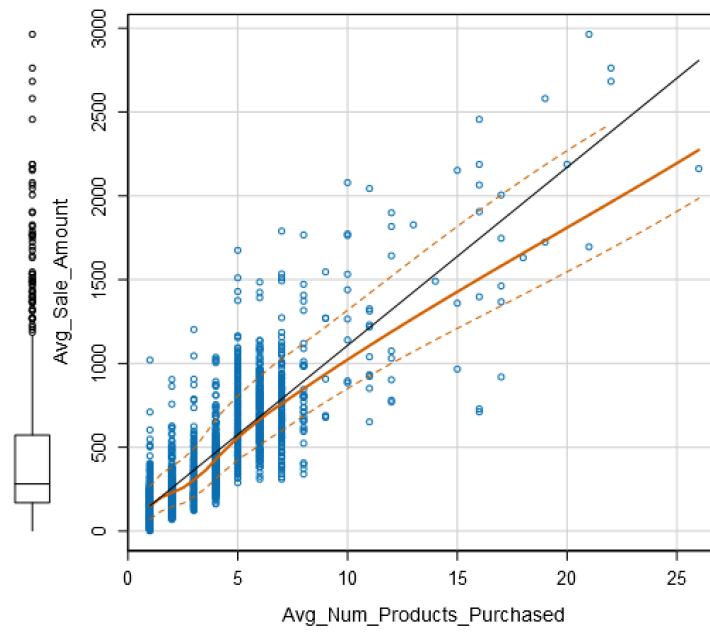
## How and why did you select the predictor variables in your model?

Name	Customer_Se	Customer_ID	Address	City	State	ZIP	Avg_Sale_Am	Store_Numbe	Responded_t	Avg_Num_Pro	#_Years_as_Customer
Pamela Wrig	Store Maili	2	376 S Jasmii	Denver	CO	80224	227.9	100	No	1	6
Danell Vald	Store Maili	7	12066 E Lak	Greenwood V	CO	80111	55	105	Yes	1	6
Jessica Rin	Store Maili	8	7225 S Gayl	Centennial	CO	80122	212.57	101	No	1	3
Nancy Clark	Store Maili	9	4497 Cornis	Denver	CO	80239	195.31	105	Yes	1	6
Andrea Brun	Store Maili	10	2316 E 5th	Denver	CO	80206	110.55	100	Yes	1	2
Danica Pont	Store Maili	11	3882 Quitma	Denver	CO	80212	149.01	106	No	1	8

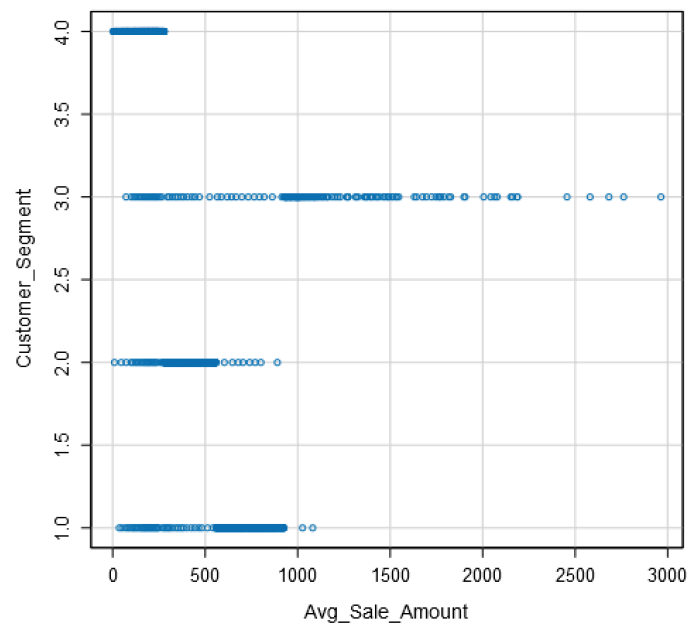
The general content of the dataset used for training is shown in the figure above. Useless information can be ignored first, like *Name*, *ID*, *ZIP*, etc.

Because *Avg\_sale\_amount* is our forecast variable, we can explore the relationship between characteristic parameters that may be relevant.

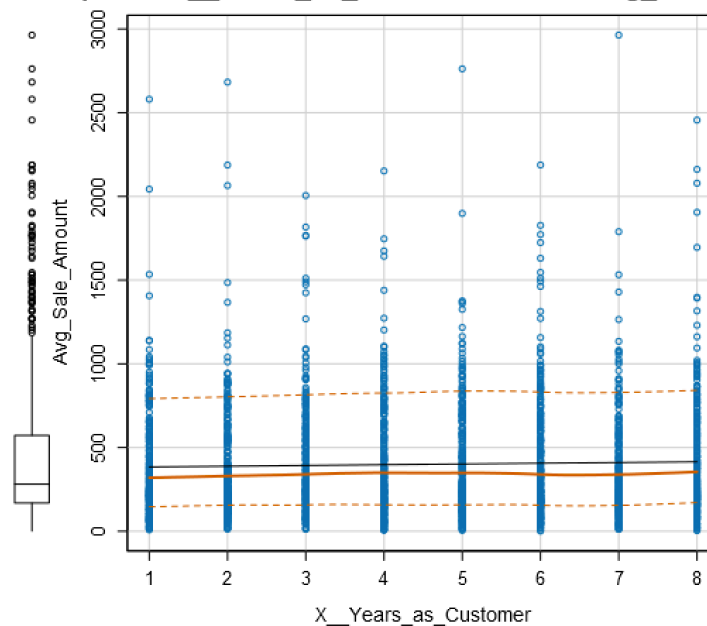
Scatterplot of Avg\_Num\_Products\_Purchased versus Avg\_Sale\_Amount



Scatterplot of Avg\_Sale\_Amount versus Customer\_Segment



atterplot of X\_Years\_as\_Customer versus Avg\_Sale\_An



From the scatter diagram of the above three possible variables and *Avg\_sales\_amount*, it's obvious that *Avg\_num\_products\_purchased* has strong linear relationship with *Avg\_sales\_amount*, and different *Customer\_segment* have different *Avg\_sales\_amount*. While *Years\_as\_customer*'s influence on *Avg\_sales\_amount* is not clear. (In fact, when using it as a parameter of linear regression, the corresponding p-value is relatively large)

**Explain why you believe your linear model is a good model.**

### Report for Linear Model LR1

#### Basic Summary

Call:

`lm(formula = Avg_Sale_Amount ~ Customer_Segment + Avg_Num_Products_Purchased, data = inputs$the.data)`

Residuals:

Min	1Q	Median	3Q	Max
-663.8	-67.3	-1.9	70.7	971.7

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	303.46	10.576	28.69	< 2.2e-16 ***
Customer_SegmentLoyalty Club Only	-149.36	8.973	-16.65	< 2.2e-16 ***
Customer_SegmentLoyalty Club and Credit Card	281.84	11.910	23.66	< 2.2e-16 ***
Customer_SegmentStore Mailing List	-245.42	9.768	-25.13	< 2.2e-16 ***
Avg_Num_Products_Purchased	66.98	1.515	44.21	< 2.2e-16 ***

Significance codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 137.48 on 2370 degrees of freedom

Multiple R-squared: 0.8369, Adjusted R-Squared: 0.8366

F-statistic: 3040 on 4 and 2370 DF, p-value: < 2.2e-16

Report for this linear model is shown in figure above. For two variables we select, their p-value is far less than 0.05 which means their relationship with *Avg\_sales\_amount* is strong. R-squared and adjusted R-squared are also good enough which indicating the linear model we built is reliable.

#### Linear equation:

$$\text{Avg\_sales\_amount} = 303.46 - 149.36(\text{Customer\_SegmentLoyalty Club Only}) + 281.34(\text{Customer\_SegmentLoyalty Club and Credit Card}) - 245.42(\text{Customer\_SegmentStore Mailing List}) + 66.98 \cdot \text{Avg\_Num\_Products\_Purchased}$$

## Conclusion

**What is your recommendation? Should the company send the catalog to these 250 customers?**

affirmative. The final calculated profit is **\$22050**. The company is interested in it when projected profits is above \$20000.