Retail Dataset Analysis

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I am an undergraduate Statistics student looking for opportunities in quantitative analysis, machine learning researches, Customer Relationship Management, actuary, and data science field with a background in actuarial, data analysis, and statistics.

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About the Data

This Online Retail II data set contains all the transactions occurring for a UK-based and registered, non-store online retail between 01/12/2009 and 09/12/2011.

InvoiceNo

Stock Code

Description

Quantity

Invoice Date

Unit Price

CustomerID

Country

The company mainly sells unique all-occasion gift-ware. Many customers of the company are wholesalers.

Source:

https://www.kaggle.com/data sets/lakshmi25npathi/onlineretail-dataset

Definition

Data Type

Invoice number	Nominal.
Product (item) code	Nominal
Product (item) name	Nominal
The quantities of each product (item) per transaction	Numeric
Invoice date and time	Numeric
Product price per unit in sterling (£).	Numeric
A 5-digit integral number uniquely assigned to each customer	Nominal
The name of the country where a customer resides	Nominal

O1 Sales Forecast

Tableau Method for Forecasting

Exponential Smoothing and Trend

- Exponential Smoothing Models iteratively predict the future value of a series of regular values from the weighted averages of the past values of the series.
- Exponential method because the value of each level is affected by each actual value of the previous level that decreases/increases exponentially—newer values are given greater weight

MAPE	Forecasting power
<10%	Highly accurate forecasting
10%~20%	Good forecasting
20%~50%	Reasonable forecasting
>50%	Weak and inaccurate forecasting

Source: Lewis (1982)

Forecast Revenue in 4 Quartal (Model 1)

Using ARIMA (Trend Multiplicative)

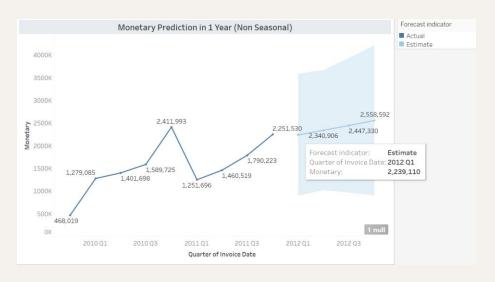
Estimate Revenue in:

Q1 2012 — 2.239.110

Q2 2012 — 2.340.906

Q3 2012 — 2.447.330

Q4 2012 — 2.558.592





9.585.938Total Revenue in 2012



Analysis Model 1

All forecasts were computed using exponential smoothing.

Sum of Monetary

M	Model			Qua	lity Me	trics			moothi efficie	•
Level T	rend	Season	RMSE	MAE	MASE	MAPE	AIC	Alpha	Beta	Gamma
Multiplica Mu	ultiplica tive	None	548,474	416,289	0.81	42,0%	248	0.211	0.500	0.000

Forecast Revenue in 4 Quartal (Model 2)

Using SARIMA (Seasonal Multiplicative)

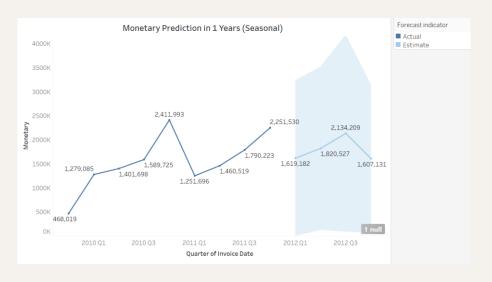
Estimate Revenue in:

Q1 2012 — 1.619.182

<u>O2 2012</u> — 1.820.527

Q3 2012 — 2.134.209

Q4 2012 — 1.607.133





7.181.049



Total Revenue in 2012

Analysis Model 2

All forecasts were computed using exponential smoothing.

Sum of Monetary

Model				Qua	lity Me	trics			mootni pefficie	9
Level	Trend	Season	RMSE	MAE	MASE	MAPE	AIC	Alpha	Beta	Gamma
Multiplica tive	None	Multiplica tive	593,008	442,668	0.93	34,2%	253	0.274	0.000	0.000

Forecast Revenue in 8 Quartal (Model 3)

Using ARIMA (Trend Multiplicative)

Estimate Revenue in:

Q1 2012 — 2.239.110

Q2 2012 — 2.340.906

Q3 2012 — 2.447.330

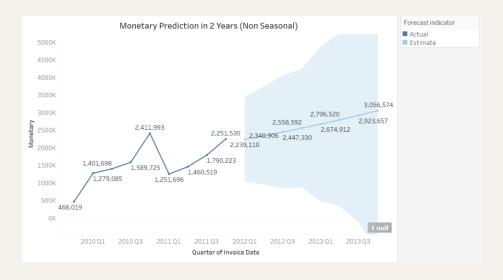
Q4 2012 — 2.558.592

O1 2013 — 2.674.912

Q2 2013 — 2.796.520

Q3 2013 — 2.923.657

Q4 2013 **—** 3.056.574



9.585.938

Total Revenue in 2012



11.451.663
Total Revenue in 2013

Analysis Model 3

All forecasts were computed using exponential smoothing.

Sum of Monetary

Model			Quality Metrics					Coefficients		
Level	Trend	Season	RMSE	MAE	MASE	MAPE	AIC	Alpha	Beta	Gamma
Multiplic ative	Multiplic ative	None	548,474	416,289	0.81	42,0%	248	0.211	0.500	0.000

Smoothing

Forecast Revenue in 8 Quartal (Model 4)

Using SARIMA (Seasonal Multiplicative)

Estimate Revenue in:

Q1 2012 — 1.619.182

Q2 2012 — 1.820.527

Q3 2012 — 2.134.209

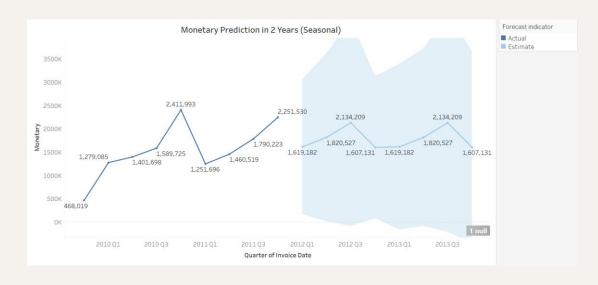
Q4 2012 — 1.607.131

O1 2013 — 1.619.182

Q2 2013 — 1.820.527

Q3 2013 — 2.134.209

Q4 2013 — 1.607.131



1.607.131

Total Revenue in 2012



1.607.131
Total Revenue in 2013

Analysis Model 4

All forecasts were computed using exponential smoothing.

Sum of Monetary

Model				Qua	lity Me	trics			pefficie	9
Level	Trend	Season	RMSE	MAE	MASE	MAPE	AIC	Alpha	Beta	Gamma
Multiplic ative	None	Multiplic ative	593,008	442,668	0.93	34,2%	253	0.274	0.000	0.000

Smoothing

Model Selection for Forecast

MAPE

Model 1 (Forecast Revenue in 4 Quartal –
Non Seasonal)

Model 2 (Forecast Revenue in 4 Quartal – Seasonal)

Model 3 (Forecast Revenue in 8 Quartal – Non Seasonal)

Model 4 (Forecast Revenue in 8 Quartal – Seasonal)

42%
34,2%
42%
34,2%

We choose Model 2 dan Model 4 as a Correct Model

O2 Regression Model

OLS Regression

2010 Monetary, Recency, and avg_order_cost are significant variable for predict Monetary in 2011

OLS Regression Results									
Dep. Variable: Model: Method: Date: Time: No. Observations Df Residuals: Df Model: Covariance Type:	Wed, 2	y OLS ast Squares 23 Mar 2022 14:12:53 1797 1793 4 nonrobust	Adj. R-squ F-statisti	atistic):	,	0.853 0.853 2602. 0.00 -16805. 3.362e+04			
============	coef	std err	t	P> t	[0.025	0.975]			
2010 Monetary Recency Frequency avg_order_cost		0.012 0.691 12.046 0.140	75.010 3.326 -1.413 -1.908	0.000 0.001 0.158 0.057		0.905 3.653 6.608 0.007			
Omnibus: Prob(Omnibus): Skew: Kurtosis:		3103.582 0.000 11.095 341.497	Durbin-Wat Jarque-Ber Prob(JB): Cond. No.		861606	2.024 5.462 0.00 2e+03			

Warnings:

- [1] Standard Errors assume that the covariance matrix of the errors is correctly specified.
- [2] The condition number is large, 1.42e+03. This might indicate that there are strong multicollinearity or other numerical problems.

Best Model from OLS Regression with p-value $< \alpha = 0.90$

$$\hat{y}_i = \beta_0 + \beta_1 x_{2010 Monetary} + \beta_2 x_{recency} + \beta_3 x_{avg_order_cost}$$

$$\hat{y}_i = 0.8821 x_{2010 \, Monetary} + 2.2981 x_{recency} \pm 0.266 x_{avg_order_cost}$$

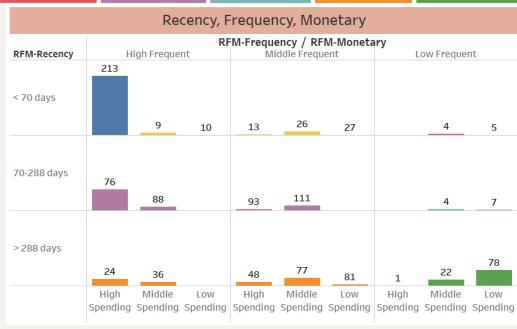
- Where the F-statistic model = $0.00 < \alpha = 0.90$ so the model is quite useful for predicting monetary in 2011
- The value of $R^2_{adj}=0.853$ implies that 85,3% variation of the data can be explained by the model

O3 RFM Analysis



Customer Segmentation via RFM Analysis

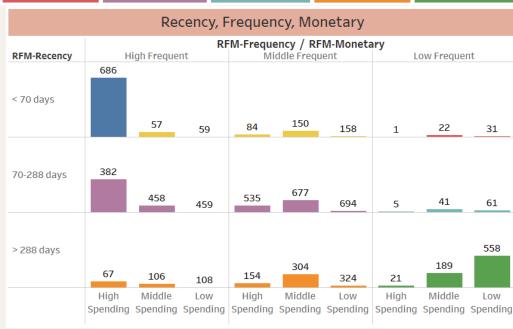
Case in 2009





Customer Segmentation via RFM Analysis

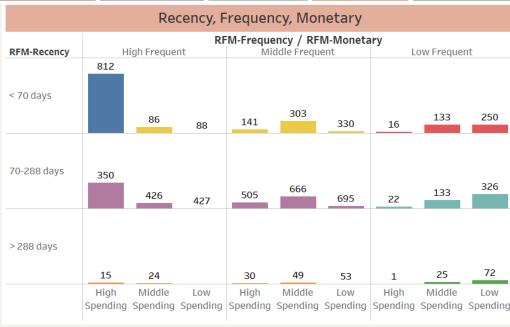
Case in 2010





Customer Segmentation via RFM Analysis

Case in 2011



Detail Customer ID in each Segment

	Detail Customer Segment Best									
RFM Segment	Customer ID	Frequency	Monetary	Recency						
Best	12347	8	5,476	23	^					
	12352	6	2,034	57						
	12358	5	3,887	22						
	12362	11	4,943	24						
	12388	7	2,760	36						
	12395	11	3,898	40						
	12408	7	4,341	53						
	12415	17	102,874	45						
	12417	15	4,790	24						
	12423	8	2,096	21	~					

	Detail Customer Segment Valuable										
RFM Segment	Customer ID	Frequency	Monetary	Recency							
Valuable	12381	5	629	25 ^							
	12427	3	826	42							
	12428	4	6,541	46							
	12444	4	4,248	42							
	12464	4	867	31							
	12504	4	779	39							
	12517	4	2,250	49							
	12541	3	981	50							
	12597	4	3,727	40							
	12610	4	1,530	43 ×							

Detail Customer Segment New Entry										
RFM Segment	Customer ID	Frequency	Monetary	Recency						
New Entry	12357	2	18,288	54						
	12367	1	169	25						
	12384	2	585	49						
	12397	2	1,365	56						
	12442	1	172	24						
	12462	2	1,190	39						
	12478	1	681	24						
	12491	1	460	59						
	12508	2	398	47						
	12526	2	797	21						

Detail Customer Segment Valued Churn					
RFM Segment	Customer ID	Frequency	Monetary	Recency	
Valued Churn	12359	6	6,711	78	^
	12380	6	5,892	77	
	12393	6	2,347	93	
	12399	6	1,497	140	
	12407	5	1,708	70	
	12409	6	19,008	99	
	12413	4	999	87	
	12414	3	682	238	
	12422	11	4,415	116	
	12450	3	434	177	~

Detail Customer ID in each Segment

Detail Customer Segment Trialist						
RFM Segment	Customer ID	Frequency	Monetary	Recency		
Trialist	12348	2	1,260	269	^	
	12363	2	552	130		
	12371	2	2,457	80		
	12375	1	190	119		
	12378	2	5,374	150		
	12394	2	1,272	84		
	12420	1	600	84		
	12461	2	275	115		
	12519	2	727	84		
	12534	1	1,050	151	~	

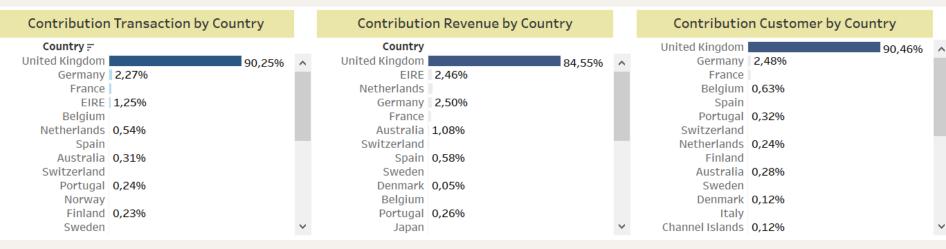
Detail Customer Segment Churn						
RFM Segment	Customer ID	Frequency	Monetary	Recency		
Churn	12623	5	2,533	297	^	
	12651	3	246	355		
	12755	9	5,227	301		
	12866	3	1,212	304		
	12933	5	1,172	304		
	12956	4	488	327		
	13093	36	43,148	296		
	13382	5	1,317	301		
	13497	4	867	303		
	13584	3	392	327	~	

Detail Customer Segment Uncertain					
RFM Segment	Customer ID	Frequency	Monetary	Recency	
Uncertain	12350	1	239	331	^
	12373	2	803	332	
	12386	2	402	358	
	12401	1	84	324	
	12410	2	1,014	329	
	12489	1	335	357	
	12501	1	2,169	357	
	12559	2	562	331	
	12561	1	239	323	
	12735	2	780	336	V

O4 Retail Data Report

Contribution by Country

Transaction, Customer, and Revenue in 2011



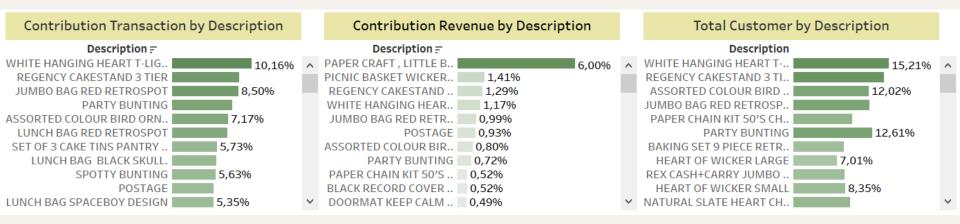
About **90,25%** total transactions come from **United Kingdom**. Next up is **Germany** with **2,27%** of total transactions

About **84,55%** total revenue come from **United Kingdom**. Next up is **Ireland** with **2,46%** of total revenue

About **90,46%** total customer come from **United Kingdom**. Next up is **Germany** with **2,48%** of total customer

Contribution by Description

Total Transaction, Customer, and Revenue in 2011



White hanging heart T-light Holder 587 accounted for 10,16% of Total Transaction in Description

Paper Craft, Little Birdy accounted for **6%** of Total Revenue in Description

White hanging heart T-light Holder 587 accounted for 15,21% of Total Customer in Description