

# Customer Life time value Descriptive

based on cohort

- CLV\_Churn
- CLV

```
In [ ]: import pandas as pd
import plotly.express as px
```

```
In [ ]: df=pd.read_csv("./Dataset/Cleaned_Online_retail_dec2010-dec2011.csv",dtype_backend="pyarrow")
df.head()
```

Out [ ]:

	Invoice	StockCode	Description	Quantity	InvoiceDate	Price	Customer ID	Country	Revenue
0	536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	2010-12-01 08:26:00	2.55	17850	United Kingdom	15.30
1	536365	71053	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	2010-12-01 08:26:00	2.75	17850	United Kingdom	22.00
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34

```
In [ ]: df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 397885 entries, 0 to 397884
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype  
---  -
0   Invoice          397885 non-null int64[pyarrow]
1   StockCode       397885 non-null string[pyarrow]
2   Description     397885 non-null string[pyarrow]
3   Quantity        397885 non-null int64[pyarrow]
4   InvoiceDate     397885 non-null string[pyarrow]
5   Price           397885 non-null double[pyarrow]
6   Customer ID    397885 non-null int64[pyarrow]
7   Country         397885 non-null string[pyarrow]
8   Revenue         397885 non-null double[pyarrow]
dtypes: double[pyarrow](2), int64[pyarrow](3), string[pyarrow](4)
memory usage: 45.5 MB
```

```
In [ ]: df["InvoiceDate"]=pd.to_datetime(df["InvoiceDate"])
```

```
In [ ]: #Adding customers first transaction date
df.loc[:, "Start_month"]=df.groupby("Customer ID")["InvoiceDate"].transform(lambda x: x.min())
```

```
In [ ]: #finding difference of every transaction from startdate of customers in months
from dateutil.relativedelta import relativedelta
```

```
def diffdates(x,y):
    return int(abs((relativedelta(x,y).years*12)+relativedelta(x,y).months))

df.loc[:, "Months_Since_Join"]=df.apply(lambda x:diffdates(x.Start_month,x.InvoiceDate ),axis=1)
df.head()
```

Out[ ]:

	Invoice	StockCode	Description	Quantity	InvoiceDate	Price	Customer ID	Country	Revenue	Start_month	Months_Since_Join
0	536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	2010-12-01 08:26:00	2.55	17850	United Kingdom	15.30	2010-12-01 08:26:00	0
1	536365	71053	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01 08:26:00	0
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	2010-12-01 08:26:00	2.75	17850	United Kingdom	22.00	2010-12-01 08:26:00	0
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01 08:26:00	0
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01 08:26:00	0

```
In [ ]: df.loc[:, "Start_month"]=df["Start_month"].transform(lambda x: x.strftime("%Y-%m"))
df.head()
```

Out[ ]:

	Invoice	StockCode	Description	Quantity	InvoiceDate	Price	Customer ID	Country	Revenue	Start_month	Months_Since_Join
0	536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	2010-12-01 08:26:00	2.55	17850	United Kingdom	15.30	2010-12-01	0
1	536365	71053	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01	0
2	536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	2010-12-01 08:26:00	2.75	17850	United Kingdom	22.00	2010-12-01	0
3	536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01	0
4	536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	2010-12-01 08:26:00	3.39	17850	United Kingdom	20.34	2010-12-01	0

```
In [ ]: #Cohort matrix with retention values
cohort_visual_df=df.groupby(["Start_month", "Months_Since_Join"])[ "Customer ID"].apply(pd.Series.nunique).reset_index()
fig=px.imshow(cohort_visual_df,pivot(index="Start_month",columns="Months_Since_Join",values="Customer ID"),color_continuous_scale="viridis",text_auto=True)
fig.update_layout(xaxis_title="Months Since First Transaction",yaxis_title="Cohort",title={
    'text' : 'Monthly Cohort for Retention',
    'x':0.5,
    'xanchor': 'center'
})
fig.show("notebook")
```



```
Churn_rate=('Frequency',lambda x:(1-(x>1).sum()/len(x))),
Avg_lifespan=('lifespan','mean')
).reset_index()

cohort_summary
```

Out[ ]:

	Start_month	cohort_size	Avg_sales	Avg_Frequency	Churn_rate	Avg_lifespan
0	2010-12-01	885	36.422175	9.398870	0.125424	267.748023
1	2011-01-01	417	229.024754	5.163070	0.184652	208.083933
2	2011-02-01	380	36.631088	4.107895	0.228947	171.002632
3	2011-03-01	452	28.076714	3.564159	0.278761	140.685841
4	2011-04-01	300	27.500523	3.080000	0.333333	113.050000
5	2011-05-01	284	241.196904	2.883803	0.306338	99.954225
6	2011-06-01	242	83.920088	2.731405	0.355372	82.946281
7	2011-07-01	188	29.477879	2.351064	0.393617	57.186170
8	2011-08-01	169	32.904727	2.118343	0.455621	42.366864
9	2011-09-01	299	35.094298	1.989967	0.521739	26.655518
10	2011-10-01	358	25.035718	1.698324	0.653631	11.645251
11	2011-11-01	323	25.476888	1.359133	0.736842	3.597523
12	2011-12-01	41	153.916796	1.048780	0.975610	0.000000

Churn CLV= ( Avg Sales \* Avg Frquency ) / Churn Rate

Assume Profit Margin= 0.10 (10%)

```
In [ ]: cohort_summary["CLV_Churn"]=((cohort_summary["Avg_sales"]*cohort_summary["Avg_Frequency"])/cohort_summary["Churn_rate"])*0.10
```

Lifetime CLV = (Avg Sales \* Avg Frequency per year \* Customer Lifetime )

since we have only 1year data if we expect customer lifetime = 2years

```
In [ ]: cohort_summary["CLV_Lifetime"]=cohort_summary["Avg_sales"]*cohort_summary["Avg_Frequency"]*2*0.10
cohort_summary
```

Out[ ]:

	Start_month	cohort_size	Avg_sales	Avg_Frequency	Churn_rate	Avg_lifespan	CLV_Churn	CLV_Lifetime
0	2010-12-01	885	36.422175	9.398870	0.125424	267.748023	272.936620	68.465457
1	2011-01-01	417	229.024754	5.163070	0.184652	208.083933	640.377007	236.494146
2	2011-02-01	380	36.631088	4.107895	0.228947	171.002632	65.725435	30.095331
3	2011-03-01	452	28.076714	3.564159	0.278761	140.685841	35.898084	20.013976
4	2011-04-01	300	27.500523	3.080000	0.333333	113.050000	25.410484	16.940322
5	2011-05-01	284	241.196904	2.883803	0.306338	99.954225	227.057775	139.112862
6	2011-06-01	242	83.920088	2.731405	0.355372	82.946281	64.501370	45.843949
7	2011-07-01	188	29.477879	2.351064	0.393617	57.186170	17.607058	13.860875
8	2011-08-01	169	32.904727	2.118343	0.455621	42.366864	15.298561	13.940701
9	2011-09-01	299	35.094298	1.989967	0.521739	26.655518	13.385325	13.967296
10	2011-10-01	358	25.035718	1.698324	0.653631	11.645251	6.505007	8.503752
11	2011-11-01	323	25.476888	1.359133	0.736842	3.597523	4.699308	6.925296
12	2011-12-01	41	153.916796	1.048780	0.975610	0.000000	16.546056	32.284987

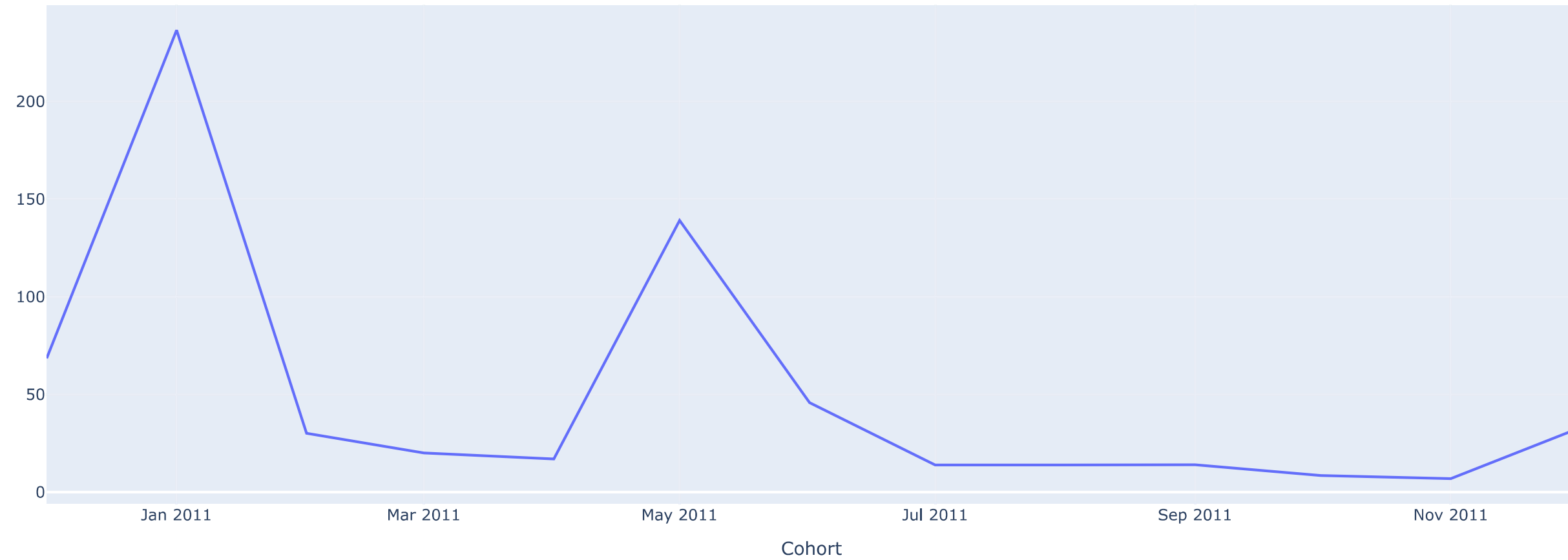
from above data

- CLV\_Lifetime= expected avg CLV from each Customer for 2years
- CLV\_Churn= Avg CLV From Each Customers with Churn

In [ ]:

```
fig=px.line(cohort_summary,x="Start_month",y="CLV_Lifetime")
fig.update_layout(xaxis_title="Cohort",yaxis_title="",title={
    'text' : 'CLV_Lifetime(~ 2 years, Profit Margin = 10%) ',
    'x':0.5,
    'xanchor': 'center'
})
fig.show("notebook")
```

CLV\_Lifetime(~ 2 years, Profit Margin = 10%)



```
In [ ]: fig=px.line(cohort_summary,x="Start_month",y="cohort_size")
fig.update_layout(xaxis_title="",yaxis_title="",title={
    'text' : 'New Customers',
    'x':0.5,
    'xanchor': 'center'
})
fig.show("notebook")
```

New Customers

