

# Broadview

Ad free Video Channel for Content Creators

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# Agenda

1. Business challenges
2. Markov Model Description
3. Customer Segmentation
4. Customer Lifetime Value (CLV)
5. Broadview Reward Vector
6. Broadview Revenue: With Policy & No Policy

# Business Challenges

- Leader in space, but spending aggressively and ultimately not profitable
- Facing increasing competition due to developing economy
- Want to spend money smarter based on customer study

# Acquistion vs. Retention

- Retaining: Understanding good vs. bad customers
- Customer churn analysis
- Cannot be defined using supervised learning

# Model Description- RFM model

## RFM METRICS

### RFM Metrics:



#### **RECENCY**

The *freshness* of customer activity.

e.g. time since last activity



#### **FREQUENCY**

The *frequency* of customer transactions.

e.g. the total number of recorded transactions



#### **MONETARY**

The *willingness* to spend.

e.g. the total transaction value

# Broadview's RFM model

- All customers assigned a “tuple” score
- Recency – Time since last activity
- Frequency- Total no. of recorded transactions
- Monetary value: Total transaction value each year/month

# Customer Segmentation

- RFM tuple infinite to finite state space
- Segmentation – Executive, Business, Personalized
- Define cut-off pt. for Churn – Probability of customer return is low

# Broadview's Markov Model

- Purchase behavior -> Customer transition to diff states
- Historical data -> Probability of state transition, (i,j)

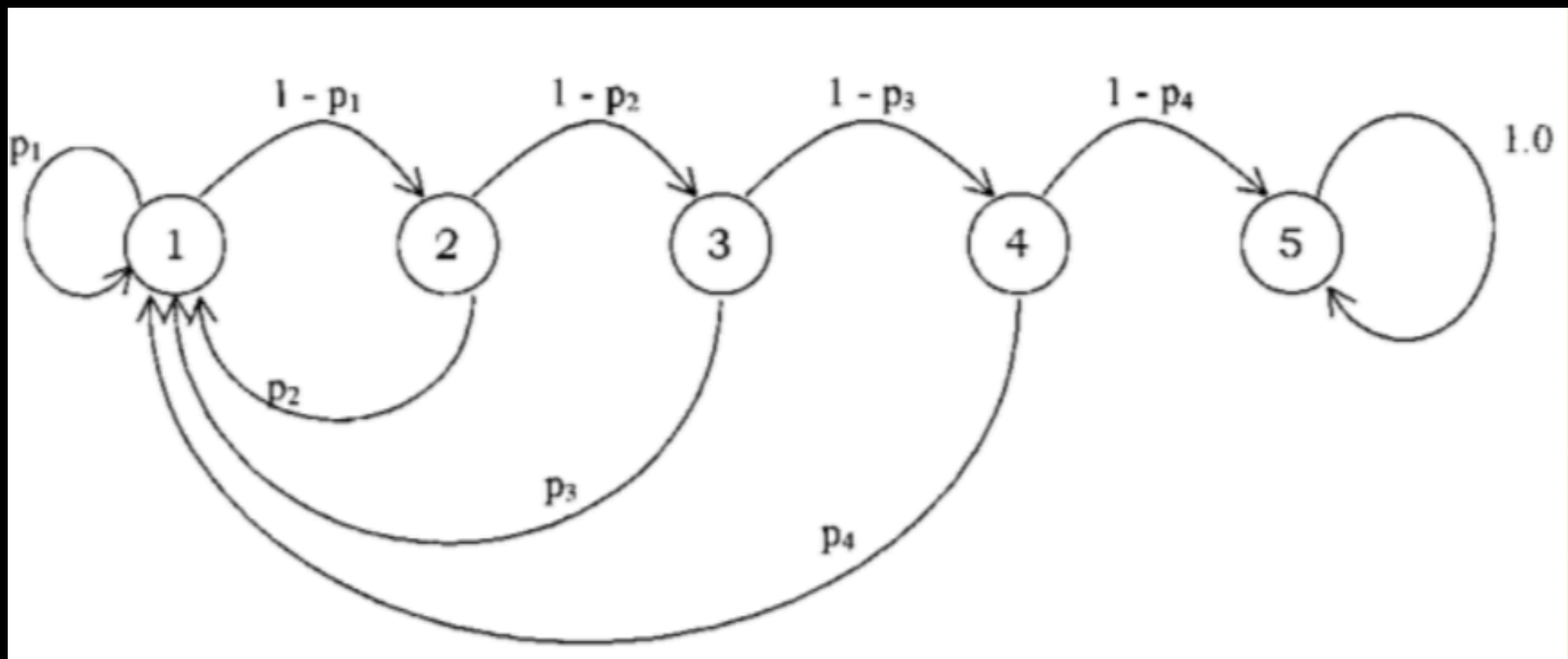
$P(i, j)$  = probability of transitioning from state i to state j over some time period



# Broadview's Markov Model

- Build matrix  $P$  of transition probabilities
- Size of matrix  $P = n \times n$ ,  $n$ =no. of states
- Assumption: Customer churn is permanent
- Absorbing state: Transition back to current stat

# Sample Markov Model



# Markov Model- Predictions

- State vector - “ $V_p(t)$ ” at time-step  $t$
- $V_p$  is of size  $n$ ;  $n$  is number of states
- $V_p[i] =$  proportion of customers in state  $i$
- $V_p(t+i) = V_p(t)P$

# Customer Lifetime Value (CLV)

- Customer Value over infinite time frame
- $CLV = -ve \rightarrow$  Customer may churn
- Marketing budget decisions



# CLV Calculation

## Step 1: Reward Vector

- $R[i] = \text{Customer value at state } i$
- $R = NC - M$
- $NC = \text{Net Contribution} = \text{Money spent that year}$
- $M = \text{Marketing Cost}$

$$\mathbf{R} = \begin{bmatrix} NC - M \\ -M \\ -M \\ -M \\ 0 \end{bmatrix}$$



# CLV Calculation

## Step 1: Reward Vector

$$\mathbf{R} = \begin{bmatrix} NC - M \\ -M \\ -M \\ -M \\ 0 \end{bmatrix}$$

- Customer with recency 0 = Non-Zero NC
- Unchurned Customers recency  $\geq 1$
- = -ve Reward
- Churned Customers = 0 Marketing Cost
- = 0 Reward



# CLV Calculation

## Step 2: Value Vector over time T

$$\mathbf{V}^T = \sum_{t=0}^T [(1 + d)^{-1} \mathbf{P}]^t \mathbf{R}$$



# CLV Calculation

## Step 3: Extending timeframe to Infinity

$$\mathbf{V} \equiv \lim_{T \rightarrow \infty} \mathbf{V}^T$$
$$= \{\mathbf{I} - (1 + d)^{-1} \mathbf{P}\}^{-1} \mathbf{R}$$





# CLV Calculation

## Step 4: Policy Adjustment

- Churned Customer = Stop marketing costs
- Customers with –ve CLV = Net loss to Broadview
- =Adjust policy to stop marketing costs



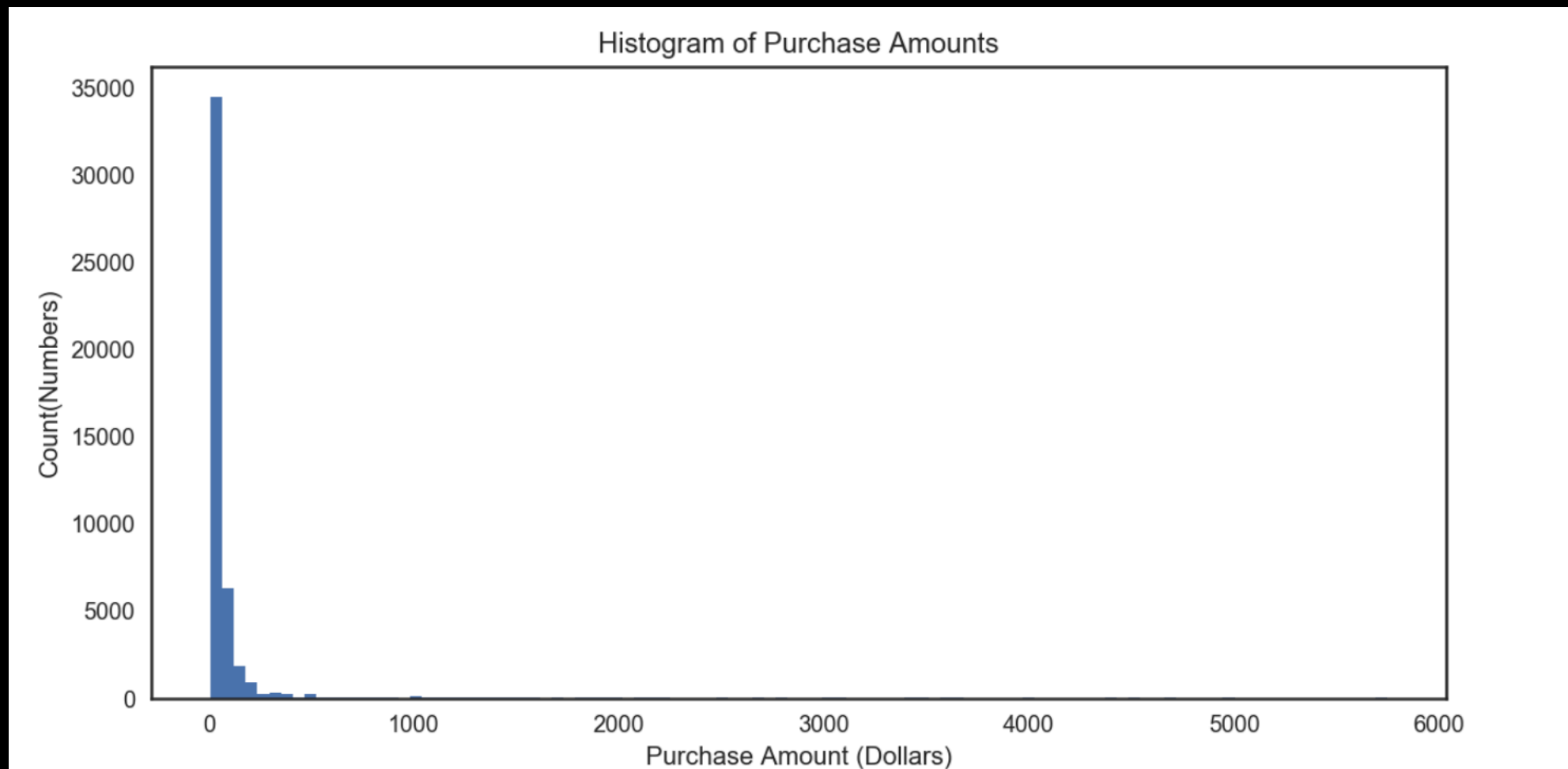
# Broadview Case Specifics



# Broadview Transaction Data

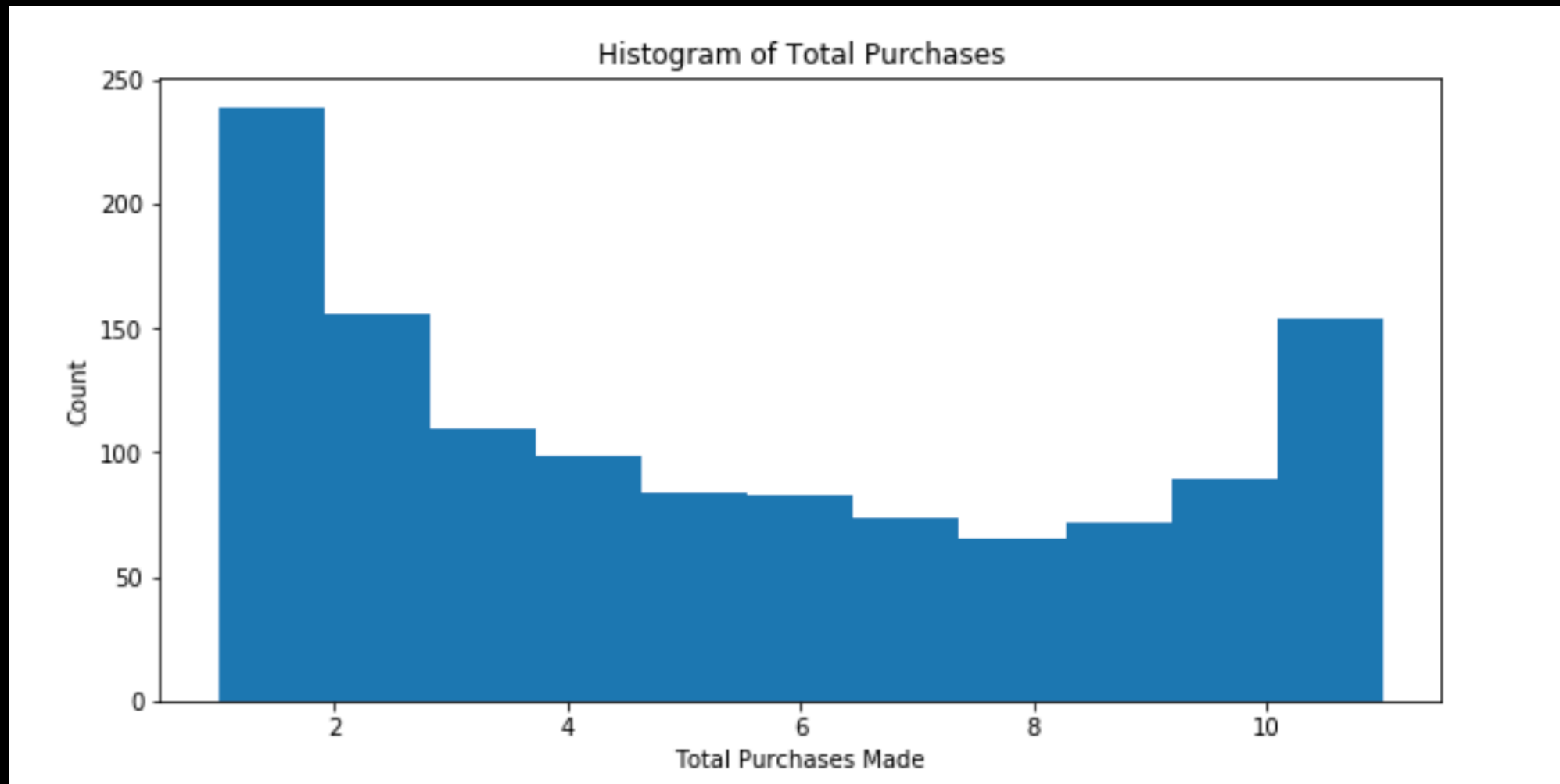
- No. of records = 44966
- Column Format = cust\_id, purchase\_amt, date\_purchase
- Data range = 2005 to 2015

# EDA – Purchase Amounts



Purchase data right skewed

# EDA – Total Purchases



Median = 5

# EDA – Boolean Matrix

year	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015
cust_id											
10	1	0	0	0	0	0	0	0	0	0	0
80	1	0	1	0	1	0	1	0	1	1	1
90	1	1	1	1	1	1	1	1	1	0	0
130	1	0	1	0	0	0	0	0	0	0	0
190	1	1	1	1	1	0	0	0	0	0	0

Customer purchases/yr = 1 = Markov transaction period

# Recency Transition Matrix

	0	1	2	3	4	5	6	7	8	9	10
0	0.716090	0.28391	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
1	0.329261	0.00000	0.670739	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
2	0.137864	0.00000	0.000000	0.862136	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
3	0.086849	0.00000	0.000000	0.000000	0.913151	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000
4	0.074963	0.00000	0.000000	0.000000	0.000000	0.925037	0.000000	0.000000	0.000000	0.000000	0.000000
5	0.035842	0.00000	0.000000	0.000000	0.000000	0.000000	0.964158	0.000000	0.000000	0.000000	0.000000
6	0.057732	0.00000	0.000000	0.000000	0.000000	0.000000	0.000000	0.942268	0.000000	0.000000	0.000000
7	0.041162	0.00000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.958838	0.000000	0.000000
8	0.009585	0.00000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.990415	0.000000
9	0.020492	0.00000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.979508
10	0.000000	0.00000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000

For  $R \geq 7$ , Churn State Probability is  $< 5\%$

# Customer Segmentation

Recency:  $R = 0$  //  $<R < 7$  //  $R \geq 7$

Frequency:  $R \leq 5$  //  $F > 5$

Monetary:  $M \leq 30$  //  $30 < M \leq 50$  //  $M > 50$

RFM States =  $3 * 2 * 3 = 18$

States with  $R \geq 7$  = Absorbing States





# State Matrix

year	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015
cust_id											
10	6	12	12	12	12	12	18	18	18	18	18
80	5	11	4	10	4	10	4	10	4	1	1
90	4	4	4	4	4	1	1	1	1	7	7
130	5	11	4	10	10	10	10	10	16	16	16
190	5	4	6	4	4	10	10	10	10	10	16

Calculate RFM tuples from Boolean Matrix

Convert to State values using segments

# RFM Transition Matrix

year	2005	2006	2007	2008	2009	2010	2011	2012	2013	2014	2015
cust_id											
10	6	12	12	12	12	12	18	18	18	18	18
80	5	11	4	10	4	10	4	10	4	1	1
90	4	4	4	4	4	1	1	1	1	7	7
130	5	11	4	10	10	10	10	10	16	16	16
190	5	4	6	4	4	10	10	10	10	10	16

TOO LARGE to illustrate

# Broadview Reward Vector

```
[[ 46.03911622],  
 [ 46.03911622],  
 [ 46.03911622],  
 [ 46.03911622],  
 [ 46.03911622],  
 [ 46.03911622],  
 [-25.         ],  
 [-25.         ],  
 [-25.         ],  
 [-25.         ],  
 [-25.         ],  
 [-25.         ],  
 [  0.         ],  
 [  0.         ],  
 [  0.         ],  
 [  0.         ],  
 [  0.         ],  
 [  0.         ]]
```

Mean Purchase Amount,  $M = 71$  (approx)

Marketing Cost/Year,  $M = \$25$

# Broadview CLV

```
[ 164.97606887],  
[ 163.53654628],  
[ 132.88523581],  
[ 141.27641545],  
[ 133.1715939 ],  
[ 94.75529752],  
[ 13.41616067],  
[ 11.0259514 ],  
[ -7.17343942],  
[ -5.88976893],  
[ -4.71694682],  
[ -18.15269545],  
[ 0.          ],  
[ 0.          ],  
[ 0.          ],  
[ 0.          ],  
[ 0.          ],  
[ 0.          ]]
```

States 9 to 12= Unchurned but -ve CLV

Policy Conclusion = Stop Marketing Costs for them

# Broadview Revenue: No Policy

2016	117192.674744
2017	213756.539922
2018	274843.715024
2019	292909.959191
2020	295450.852444
2021	292481.817299
2022	287665.24994
2023	282355.970819
2024	277062.491102
2025	271977.620176

Total Revenue = 2.7M (approx)

# Broadview Revenue: With Policy



2016	242931.3269
2017	339495.192079
2018	387466.427388
2019	400117.133611
2020	399549.233561
2021	394034.374733
2022	386729.05203
2023	378889.023419
2024	371044.570756
2025	363435.328919

Total Revenue = 3.6M (approx)

# Broadview Revenue Calculation



Use  $T = 10$

Historical Data Observation = Avg. Customers/Yr =  
2144

# Thank you!

