



### Maven Halloween Challenge

**Goal:** Use data to find the 3 types of Halloween candy that will make you the most popular house on the block.

**Details:** Using online votes ranking 85 types of candy, your task is to find the 3 treats you'll give out on Halloween to guarantee that trick-or-treaters of all tastes find something they'll love and present the data to back up your decision.

## **Approach**



Python Pandas: Load and View the data.



**Machine Learning:** PCA and jitter to prepare the data for the scatter plot.



**Tableau:** Explore the scatter plot and make recommendations

#### **Python Pandas**

```
import pandas as pd
      candies = pd.read csv('candy-data.csv')
      candies.head()
         competitorname chocolate fruity caramel peanutyalmondy nougat crispedricewafer hard bar pluribus sugarpercent pricepercent winpercent
[2]:
               100 Grand
                                                                                                                        0.732
                                                                                                                                     0.860
                                                                                                                                             66.971725
      0
                                        0
                                                                  0
                                                                                                0
                                                                                                              0
            3 Musketeers
                                                                                                                                             67.602936
     1
                                        0
                                                 0
                                                                  0
                                                                                                0
                                                                                                              0
                                                                                                                        0.604
                                                                                                                                     0.511
               One dime
      2
                                 0
                                        0
                                                 0
                                                                  0
                                                                          0
                                                                                          0
                                                                                                0
                                                                                                     0
                                                                                                              0
                                                                                                                        0.011
                                                                                                                                     0.116
                                                                                                                                             32.261086
      3
             One quarter
                                                                          0
                                                                                                              0
                                                                                                                        0.011
                                                                                                                                     0.511
                                                                                                                                             46.116505
      4
               Air Heads
                                                                                                                        0.906
                                                                                                                                             52.341465
                                 0
                                                 0
                                                                  0
                                                                          0
                                                                                                     0
                                                                                                              0
                                                                                                                                     0.511
      flavours subset = candies.iloc[:,1:-3]
      flavours subset.head()
[3]
```

]:		chocolate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer	hard	bar	pluribus
	0	1	0	1	0	0	1	0	1	0
	1	1	0	0	0	1	0	0	1	0
	2	0	0	0	0	0	0	0	0	0
	3	0	0	0	0	0	0	0	0	0
	4	0	1	0	0	0	0	0	0	0



#### **Machine Learning**

#### Principal Component Analysis (PCA)

```
from sklearn.decomposition import PCA
     pca = PCA(n_components=2)
     pca.fit(flavours_subset)
[6]:
             PCA
     PCA(n components=2)
     pca.explained variance ratio
     array([0.45736631, 0.15454393])
     These 2 newly created colums out of the 9 original columns are able to capture around 60% of the information in the original 9 colums, which is a decent representation.
     candies_2d = pd.DataFrame(pca.transform(flavours_subset))
     candies_2d.head()
[8]:
         1.305021 -0.335978
         1.123628 -0.294698
         0.034733 -0.204088
         0.034733 -0.204088
                                                                                                                                                   1st PC
        -0.469043 -0.565369
```

```
[8]: candies_2d = pd.DataFrame(pca.transform(flavours_subset))
      candies_2d.head()
 [8]:
                0
      0 1.305021 -0.335978
      1 1.123628 -0.294698
      2 0.034733 -0.204088
      3 0.034733 -0.204088
      4 -0.469043 -0.565369
      candies_2d.plot(kind='scatter',x=0,y=1)
[19]: <Axes: xlabel='0', ylabel='1'>
           1.00
           0.75
           0.50
           0.25
       ı
           0.00
         -0.25
         -0.50
         -0.75
         -1.00
```

since some of the candies have similar values of these 9 features they are overlapping so we should add a little bit of random noise to have less overlap

1.5

1.0

0.0

-0.5

-1.0

0.5

0

### **Jitter (Random Noise)**

since some of the candies have similar values of these 9 features they are overlapping so we should add a little bit of random noise to have less overlap

```
[38]: import numpy as np

candies['x_jitter'] = candies_2d[0] + np.random.randn(85)*0.1

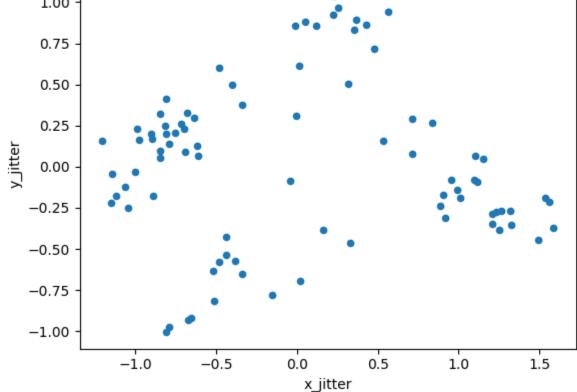
candies['y_jitter'] = candies_2d[1] + np.random.randn(85)*0.1

candies.plot(kind='scatter',x='x_jitter',y='y_jitter')

[38]: <Axes: xlabel='x_jitter', ylabel='y_jitter'>

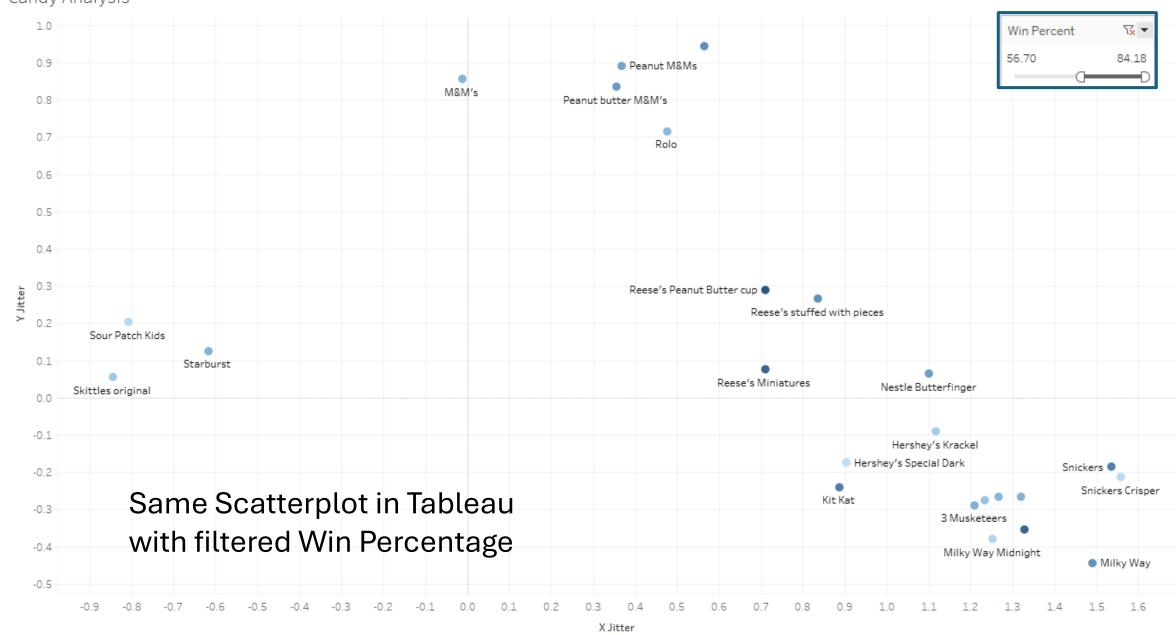
1.00 -

0.75 -
```

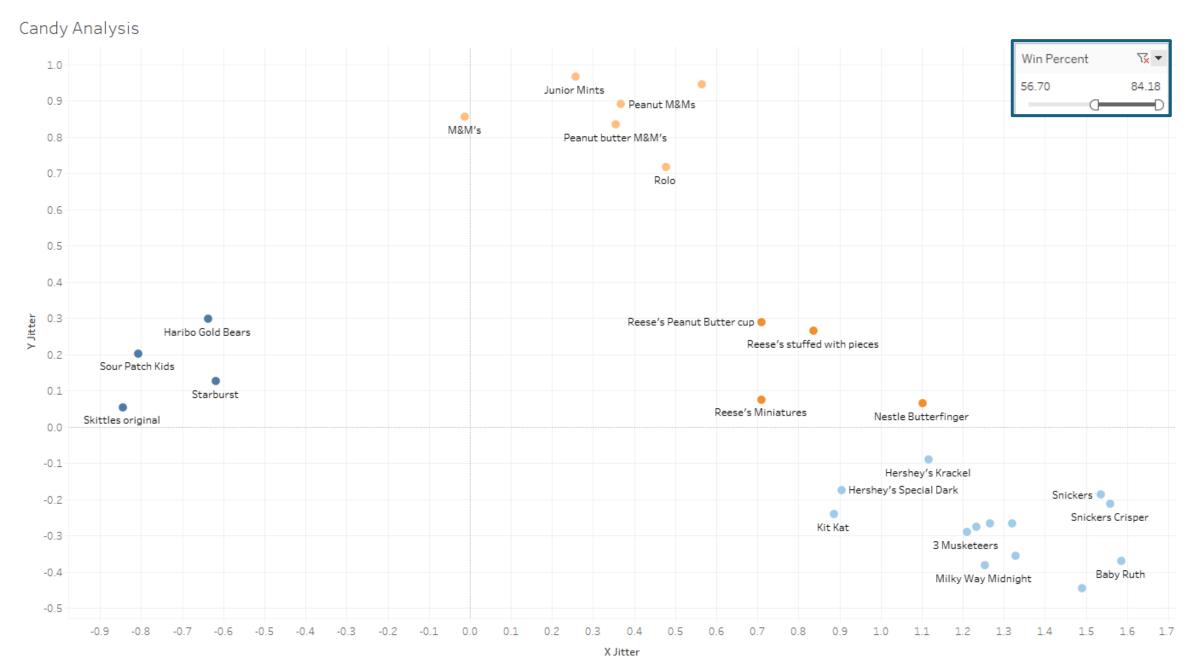


#### **Tableau**



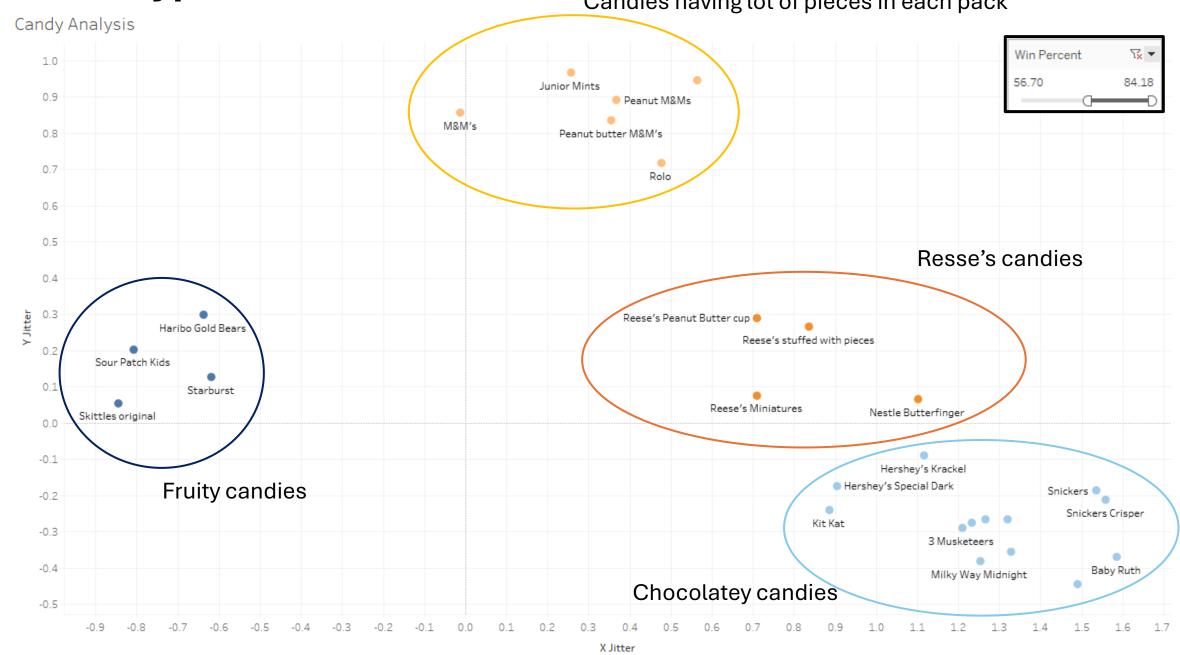


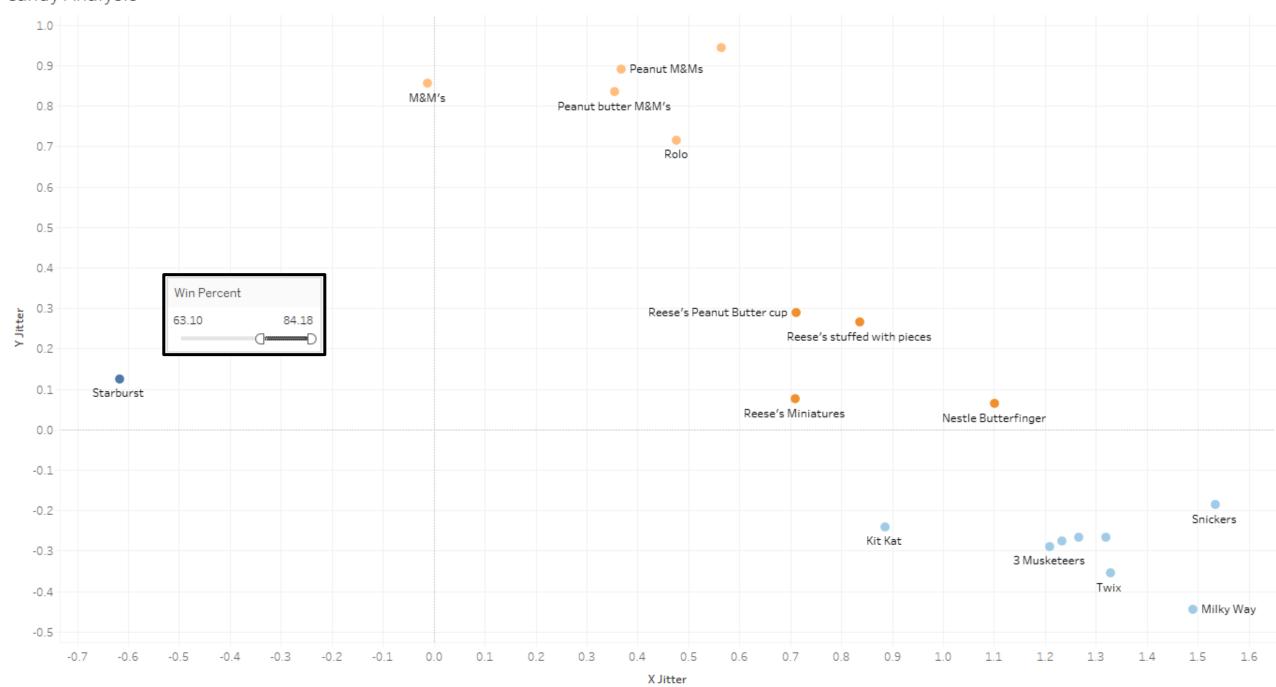
#### 4 Clusters

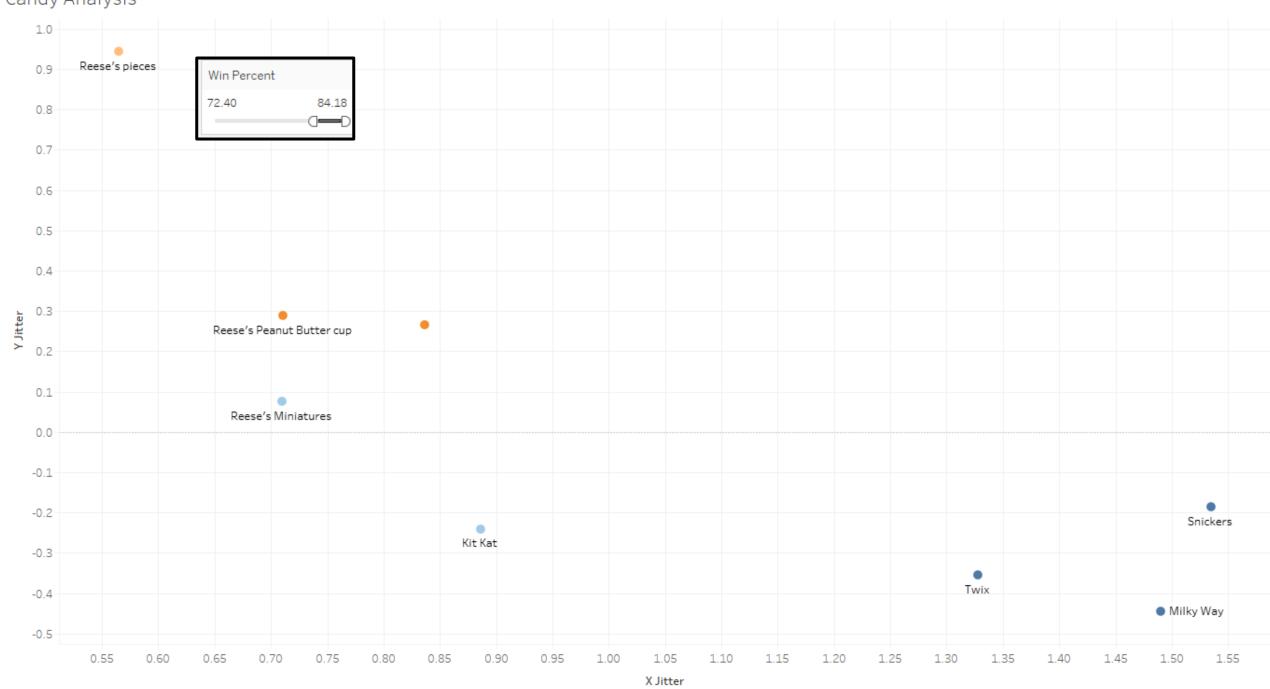


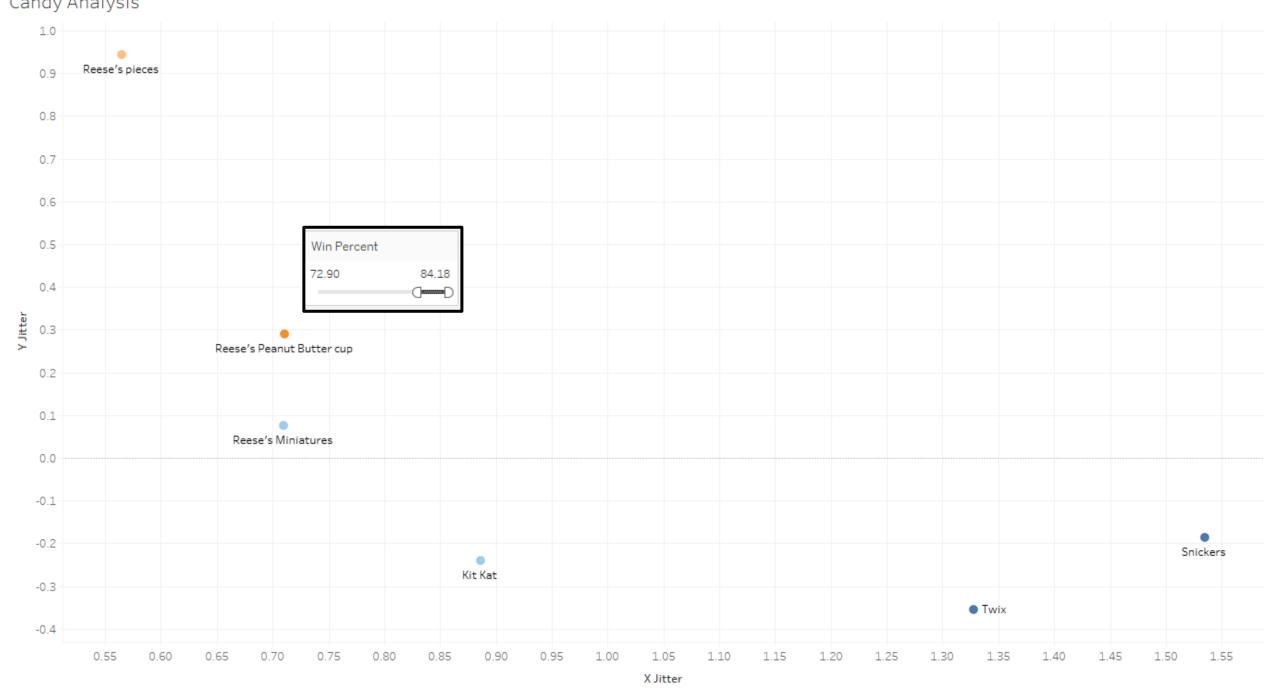
**Cluster types** 

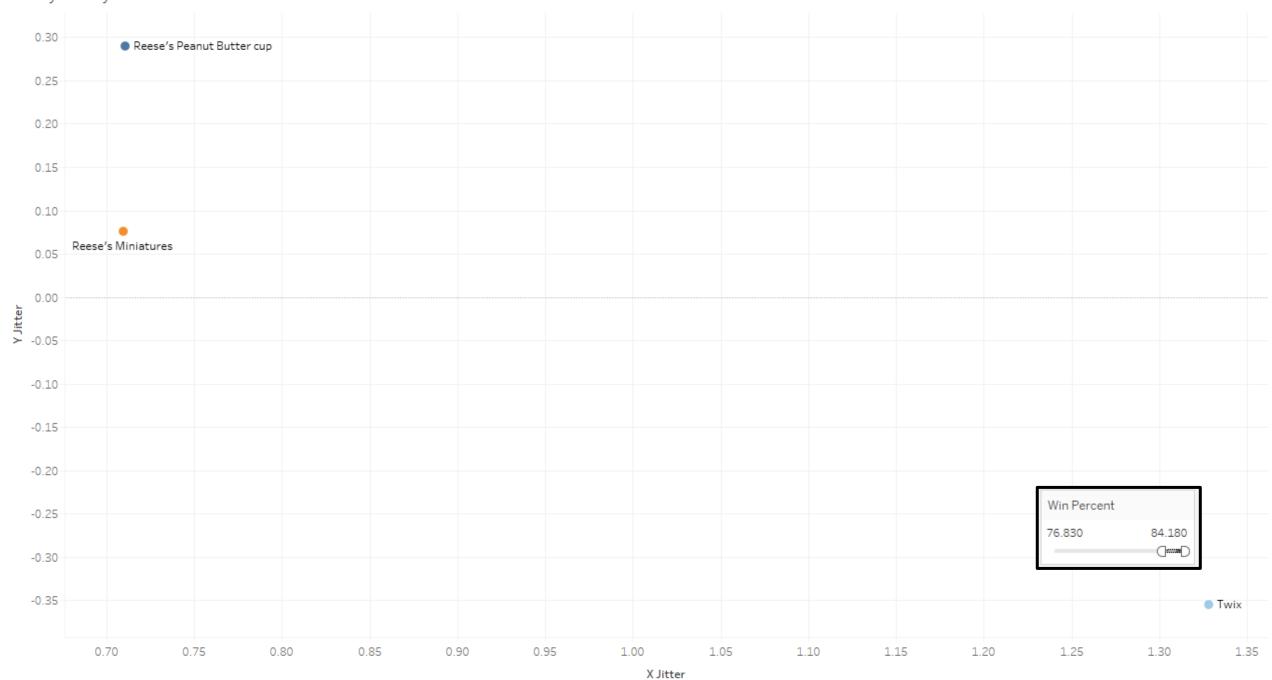
Candies having lot of pieces in each pack

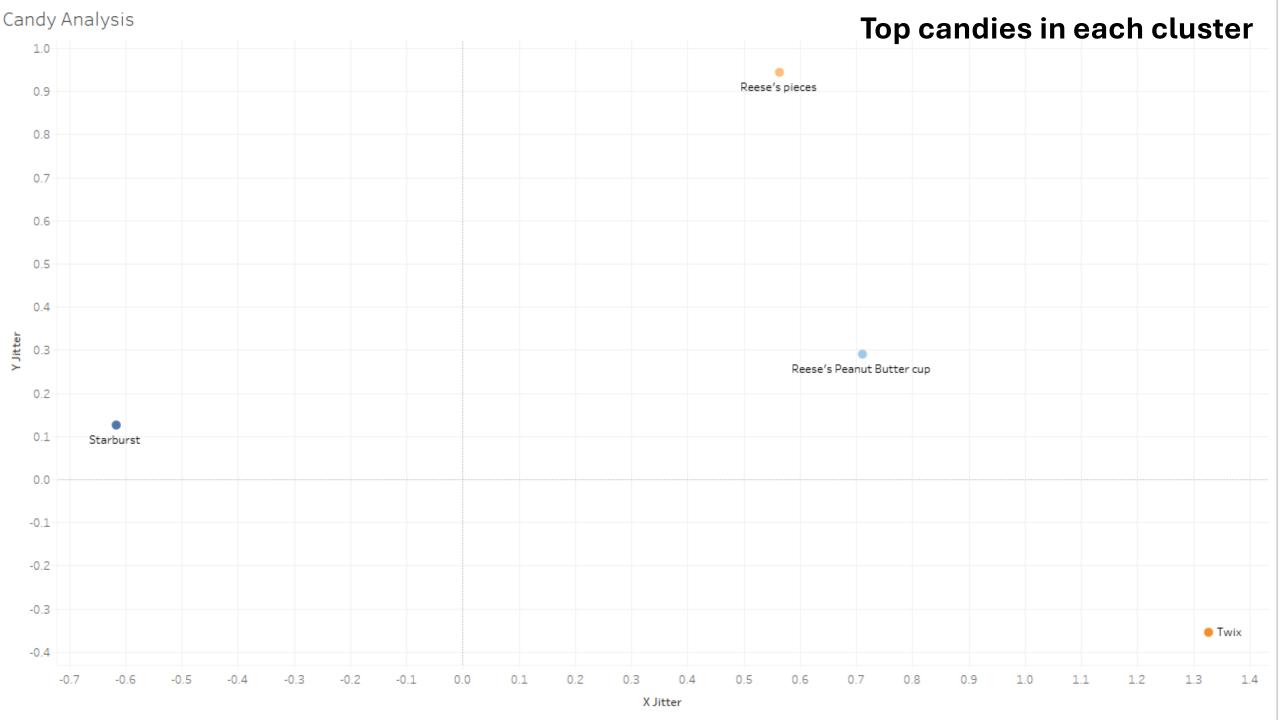


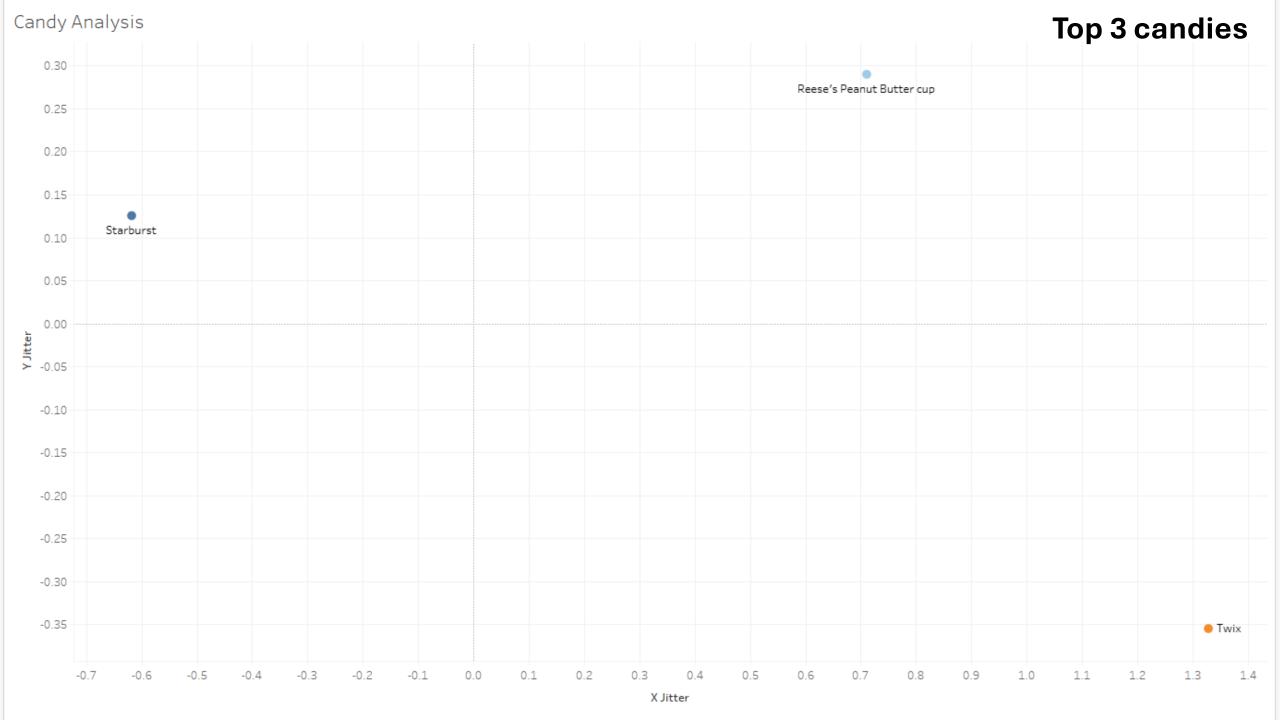












### **Top 3 recommendations**

- 1. Reese's Peanut Butter Cup
- 2. Twix
- 3. Starburst

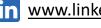








# Thank You





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