Clustering Project

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Project Topics & Goals

Goal: Create two customer segmentations on demographic & financial customer information and visualize the clusters

Topics:

- Data Wrangling
- Data Visualization
- Data Preparation and Feature Engineering
- Dimensionality Reduction
- Unsupervised Learning

Data Engineering

Standard cleaning & EDA on datasets

Demographics: Customer information dataset

- Binning (e.g. Geography)
- Dummy Coding
- Min-max Scaler for specific columns

Financial: Combination of variables from transactions and account datasets

- Recorded transactions into categories (small, medium, high) based on distance from mean (sd)
- Removed categorical variables
- Standard Scalar
- Tried other combinations of variables and recoding. This was the final iteration

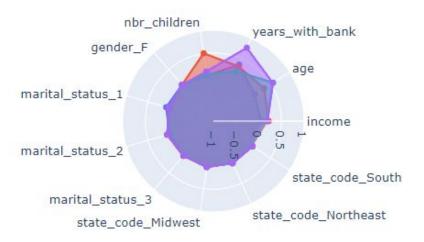
Demographic Segmentation

7 scenarios:

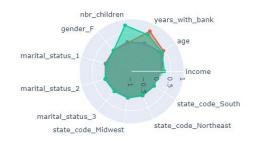
- Standard scaler with all features: segmentation based on marital status
- Standard scaler with all features: segmentation based on state bins
- Standard scaler with all features: big influence with gender feature
- Standard scaler with all features: big influence without dummies
- Standard scaler only 'income', 'age', 'years_with_bank', 'nbr_children' and multiplying the dummies with 0.5: income, nbre_children
- MinMax scaler only 'income', 'age', 'years_with_bank', 'nbr_children' and multiplying the dummies with 0.5: influence of marital status
- MinMax scaler only 'income', 'age', 'years_with_bank', 'nbr_children' and multiplying the dummies with 0.1 less influence

Demographic Segmentation

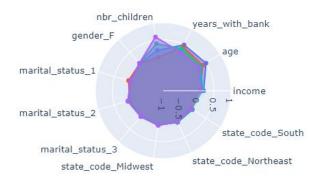
Kmeans (4)



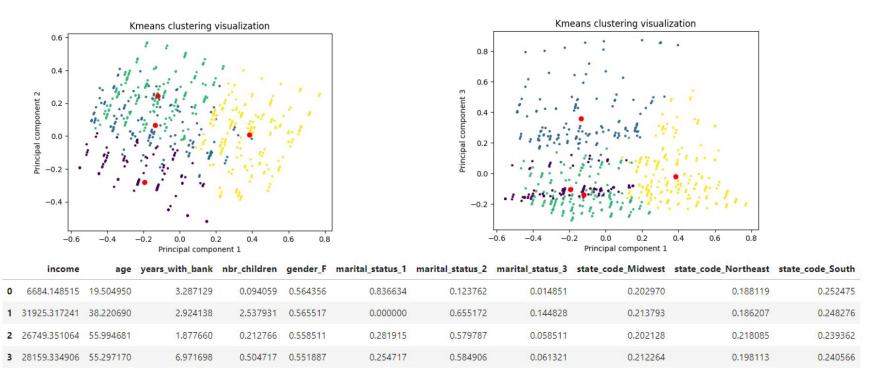
Hierarchical(3)



DBSCAN(4)-170



Demographic Segmentation



Financial Segmentation

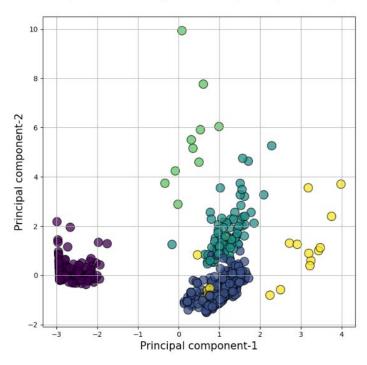
Segmentations:

- KMeans (selected)
- Hierarchical
- DBScan

Selected **K = 5 clusters** based on Silhouette & Elbow

Segmentation plotted to the right on first two PCs indicates 3 main classes and then 2 'outlier' / rarer classes

Class separation using first two principal components



Financial Segmentation (continued)

- Segment 0: Few transactions of any type (n=189)
- Segment 1: Many small transactions (n=359)
- Segment 2: Many transactions of all types (n=85)
- Segment 3: Many transactions & high savings balance (n=10)*
- Segment 4: Many transactions & high credit balance (n=22)*

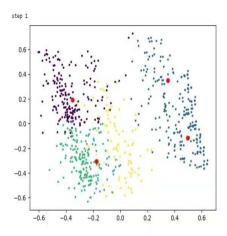


Fig: Polar chart, values rescaled to % of max means between clusters

Mean Values by Segment:

	trans_Category_large	trans_Category_medium	trans_Category_small	credit_balance	savings_balance	years_with_bank	acct_type_CC	acct_type_CK	acct_type_5V
class									
0	0.735849	1.358491	28,566038	715.837522	1016.999306	3.957672	0.597884	0.280423	0.761905
1	2.200557	6.637883	86.103064	939.505358	871.900046	3.699164	0.738162	1.000000	0.601671
2	13.411765	16.658824	67.752941	1476.993333	1437.573902	4.258824	0.705882	1.000000	0.482353
3	17.100000	14.600000	65.600000	703.050000	11172.385000	4.500000	0.800000	1.000000	1.000000
4	11.230769	10.692308	70.923077	5928,360455	2420.088000	3.818182	1.000000	0.590909	0.454545

KMeans Stepwise Visualization



Challenges

- Categorical / binary variables
- Running many models / checking different ways of clustering
- Data engineering challenges