

Pharmacy Claims – Fraud Detection

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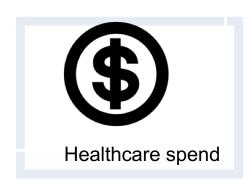
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Outline

- Background
 - Fraud Waste and Abuse
 - Current Status and Challenges
- Project
 - Objective
 - Steps
- Implementation
- Initial Results
- Next steps
- Q&A



Background



\$3.35 Trillion \$10K



67,000 700



4 billion 11.9%



950K \$6.5B

Background

Physician

Prescribes

Insurance Company

Pays Claim

Patient

Fills up

Pharmacy

Submits Claim



Fraud Waste and Abuse

- 3% to 10% of healthcare spend
- Possible Actors (Alone or in collusion)
 - Patient
 - Physician
 - Pharmacy



Current Status and Challenges

- Transaction level checks are in place
- Actor level checks are primitive
- Rule based checks are based on historical fraud
 - Fraudsters are innovative
- False positives are expensive



Project

Objective

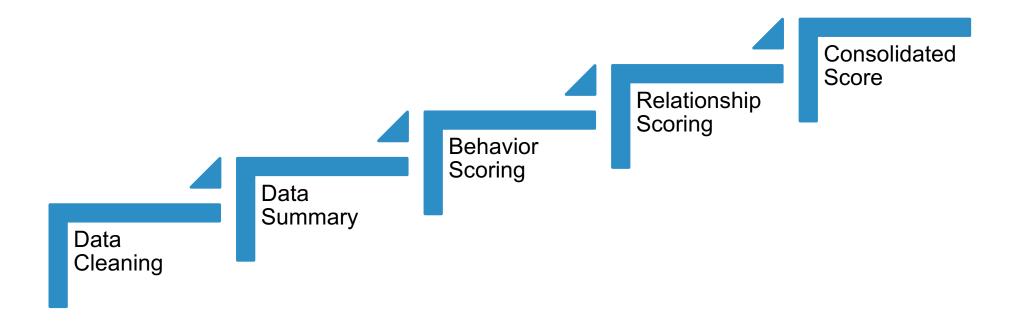
- Identify anomalous / possible fraudulent actors
- Rank them
 - Prioritize investigations

What did we do?

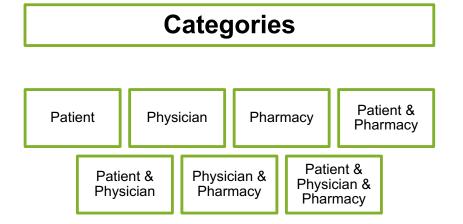
- Identify and rank anomalous behavior
- Identify and rank anomalous relationships between actors
- Generate consolidated scores to rank

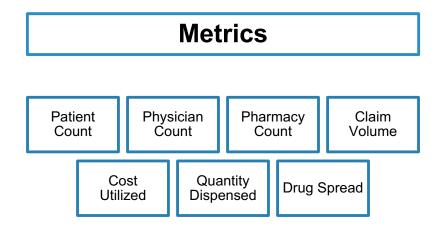


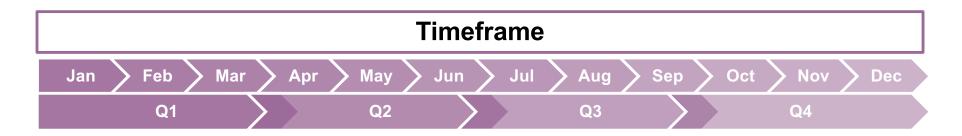
Steps



Data Summary







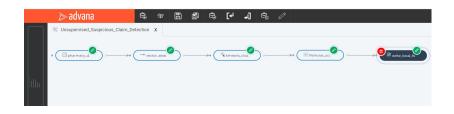


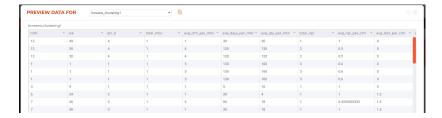
Anomalous behavior

- Grouping similar behavior
- Group Density
- Distance from Group Center

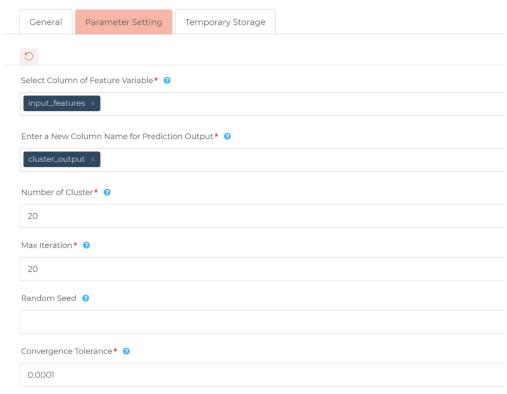


K Means Model





KMEANS_CLUSTERING1



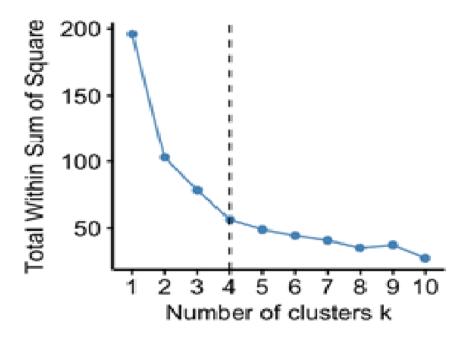


Optimal 'K'

Elbow Method

- Compute cluster algorithm for different values of K
- Calculate WSS for each K and plot the curve
- Location of the Bend will be the optimal value of 'K'

Optimal Number of Clusters





Anomaly Score

- Density Factor = Size of the cluster / Total Size
- Distance Factor = Distance between data point and center of cluster / Distance of the farthest point in the cluster
- Score = Max of two scores
- Actor Score = Sum of level wise scores with weightages



Example

Patient Level Score card

	Cluster_Size_Factor	Cluster_Distance_Factor	Cluster_Size_Factor	Cluster_Distance_Factor	Cluster_Size_Factor	Cluster_Distance_Factor	Cluster_Size_Factor	Cluster_Distance_Factor	
Patient ID	Group 1	Group 1	Group 2	Group 2	Group 3	Group 3	Group 4	Group 4	Anomaly_Score
Patient - 1	1.0000	1.0000	0.9974	1.0000	0.9979	1.0000	0.9979	1.0000	0.9992
Patient - 2	1.0000	0.9985	0.9974	0.9997	0.9979	1.0000	0.9979	1.0000	0.9989
Patient - 3	0.9971	0.9997	0.9974	0.9997	0.9979	0.9998	0.9979	0.9998	0.9987
Patient - 4	0.9971	0.9990	0.9974	0.9997	0.9979	0.9998	0.9979	0.9998	0.9986
Patient - 5	0.9971	0.9972	0.9974	1.0000	0.9979	1.0000	0.9979	1.0000	0.9984
Patient - 6	0.9971	0.9962	0.9974	1.0000	0.9979	1.0000	0.9979	1.0000	0.9983

Physician Level Score card

	Cluster Size Factor	Cluster_Distance_Factor	Cluster Size Factor	Cluster Distance Factor	Cluster Size Factor	Cluster Distance Factor	Cluster Size Factor	Cluster Distance Factor	
Physician ID									Anomaly_Score
Physician - 1	0.9965	1.0000	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	1.0000
Physician - 2	0.9997	1.0000	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	1.0000
Physician - 3	0.9965	1.0000	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	1.0000
Physician - 4	0.9997	0.9998	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	1.0000
Physician - 5	0.9965	0.9998	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	0.9999
Physician - 6	0.9965	0.9998	0.9973	1.0000	0.9979	1.0000	0.9979	1.0000	0.9999

Pharmacy Level Score card

	Chuston Sino Foston	Cluster Distance Factor	Chuston Sine Foston	Cluster Distance Factor	Charten Sine Feater	Cluster Distance Factor	Chuston Sino Foston	Cluster Distance Factor	
Pharmacy ID	Group 1	Group 1	Group 2	Group 2	Group 3	Group 3	Group 4	Group 4	Anomaly_Score
Pharmacy - 1	0.9776	1.0000	0.9974	1.0000	0.9973	1.0000	0.9979	1.0000	1.0000
Pharmacy - 2	0.9776	1.0000	0.9974	1.0000	0.9973	1.0000	0.9979	1.0000	1.0000
Pharmacy - 3	0.9776	1.0000	0.9974	1.0000	0.9973	1.0000	0.9979	1.0000	1.0000
Pharmacy - 4	0.9994	0.9998	0.9998	1.0000	0.9998	1.0000	0.9999	1.0000	1.0000
Pharmacy - 5	0.9776	0.9998	0.9974	1.0000	0.9973	1.0000	0.9979	1.0000	0.9999
Pharmacy - 6	0.9776	0.9998	0.9974	1.0000	0.9973	1.0000	0.9979	1.0000	0.9999



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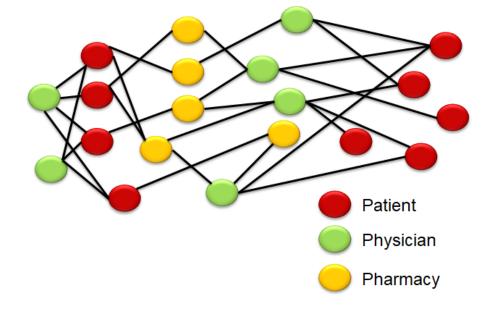
Anomalous Relationship

- Analyze
 - # of Connected Neighbors
 - # of Neighbors' Neighbors



GraphX

- Create a graph with multiple node types
 - Physician
 - Patient
 - Pharmacy



Calculate Neighbor Count

- First level Neighbor
 - Compute Immediate neighbor's degree from the each vertex
- Second Level Neighbor
 - Identify Neighbor's Neighbor degree and bring to the parent vertex
- Third Level Neighbor
 - Identify Second Level Neighbor's Neighbor degree and bring to the parent vertex



Anomaly Score

- Based on position in each count
- Consolidated score = Max of all the counts



Consolidate Score

 Sum of Behavior Score and Relationship Score with configurable weightages



Implementation – How it works?

- Web Application for the use of Analysts
- Features
 - Upload Claim Data
 - Run models
 - View Results Actor wise ranks
 - Action Tag as False Positive or Initiate Case
 - Feedback Input Investigate status



Initial Results

- Many findings would have escaped Rule based checks
- Initial investigation results prove less false positives
- Ranking weightages might need some tuning



Next steps

- Supervised techniques with investigation results
- Use additional data
 - Social media Twitter, Facebook, Review data
 - Address and Property data Zillow



Q & A





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