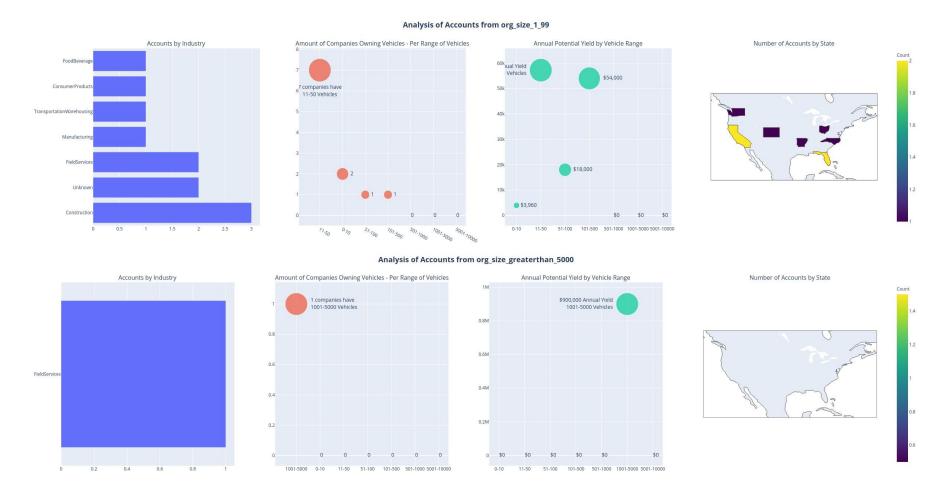
# Identifying High-Potential Accounts for Sales Targeting

A Data-Driven Approach

Vitoria Lima

# **TAL (Target Account List):** 30 High-Potential accounts yielding jointly potentially ~5.4 mio USD yearly\*



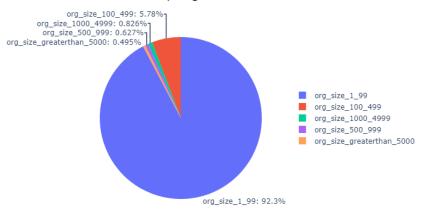
# Agenda:

- 1. Data Overview
- 2. Feature Enhancing and Selection
- 3. Model: Additive Logistic Regression
- 4. Prediction Results
- 5. Insights and Business Impact
- 6. Limitations
- 7. Q&A

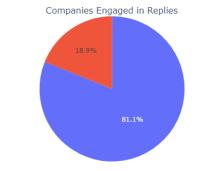
Vitoria Lima 3 of 20

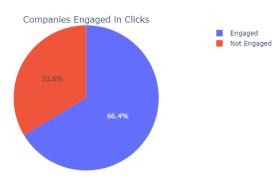
# 1: Data Overview >3k companies across the 6 given datasets

#### Distribution of Accounts by Organization Size

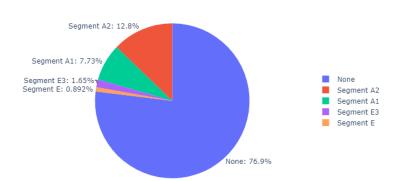








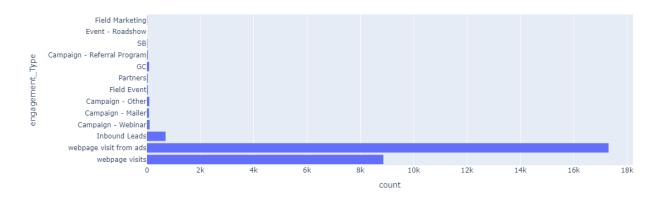
#### Distribution of Companies by Segments



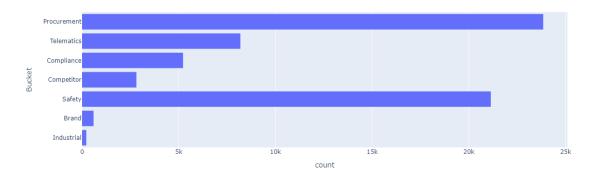
Vitoria Lima 4 of 20

### 1.1: Data Overview

#### Distribution of Engagements per Type



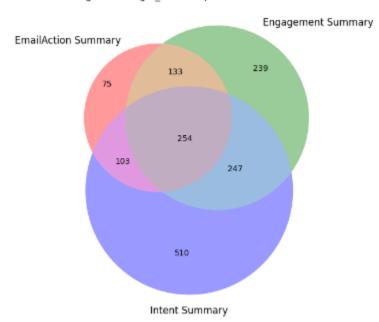
#### Distribution of Engagements per Type



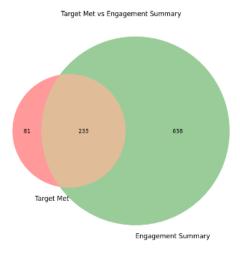
Vitoria Lima 5 of 20

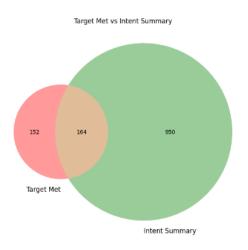
# 1.2: Data Sparsity

Venn Diagram of target\_met Companies in Different Summaries









Vitoria Lima 6 of 20

# 1.3: Training Data and Feature Engineering

This has resulted in:

Train: 254 observations (out of which 122 sales)

Test: 132 observations

- Enhancement: Feature Engineering
  - New features: lag and rolling features, days between engagements, average amount of days between engagements, days since last engagement, average weekly engagement rate, etc.

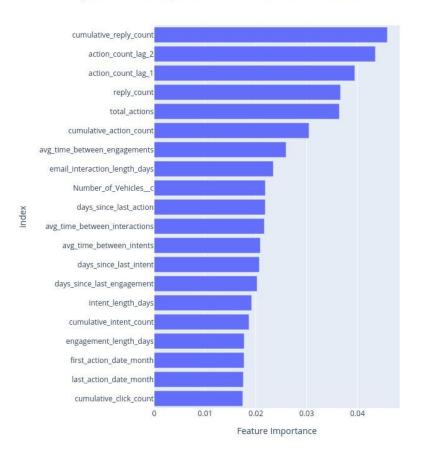
...resulting in ~100 total features.

Vitoria Lima 7 of 20

### 2: Feature Selection

#### Random Forest



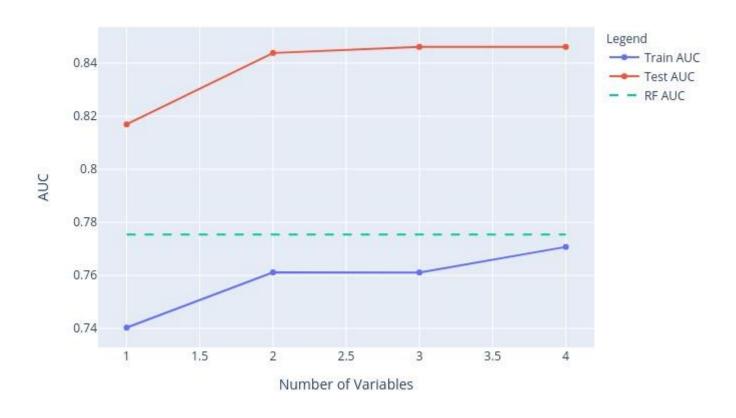


Vitoria Lima 8 of 20

# 3: Modelling

#### Additive Logistic Regression

#### AUC Scores for Top 4 Features



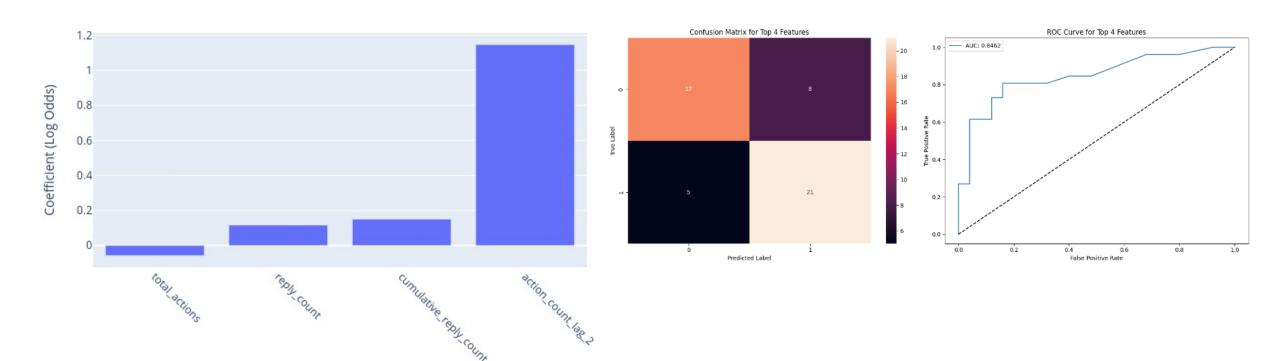
Vitoria Lima 9 of 20

# 3.1: Modelling

Feature

#### Additive Logistic Regression

LR Coefficients for Top 4 Features

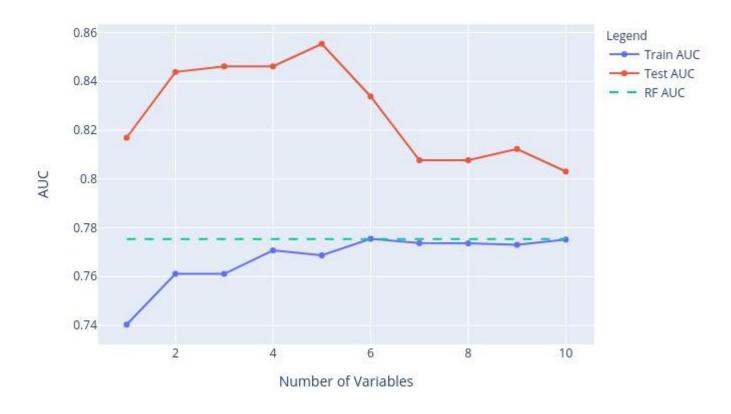


Vitoria Lima 10 of 20

# 3.2: Modelling

#### Additive Logistic Regression

#### AUC Scores for Top 10 Features

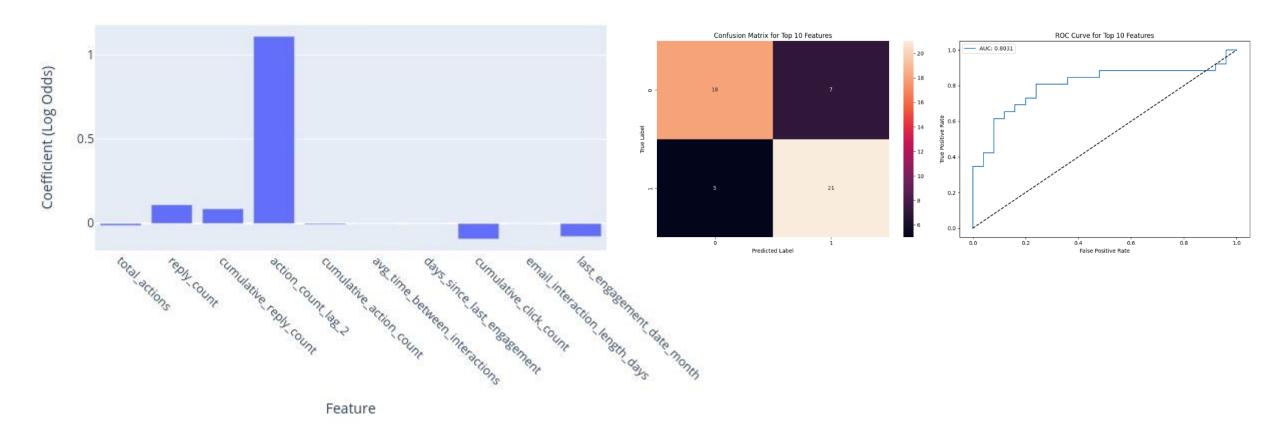


Vitoria Lima 11 of 20

### 3.3: Modelling

#### Additive Logistic Regression

LR Coefficients for Top 10 Features



Vitoria Lima 12 of 20

#### 4: Prediction Results

- Model with 4 predictors:
  - 40 potential costumers
- Model with 10 predictors:
  - 35 potential costumers

...30 in common

Vitoria Lima 13 of 20

### 4.1: Prediction and Lead Scoring

```
potential_accounts_high = potential_accounts[potential_accounts['Max_Proba']>0.7]
potential_accounts_high.head()
potential_accounts_low = potential_accounts[potential_accounts['Max_Proba']<0.7]
potential_accounts_low.head()

</pre>

0.0s
```

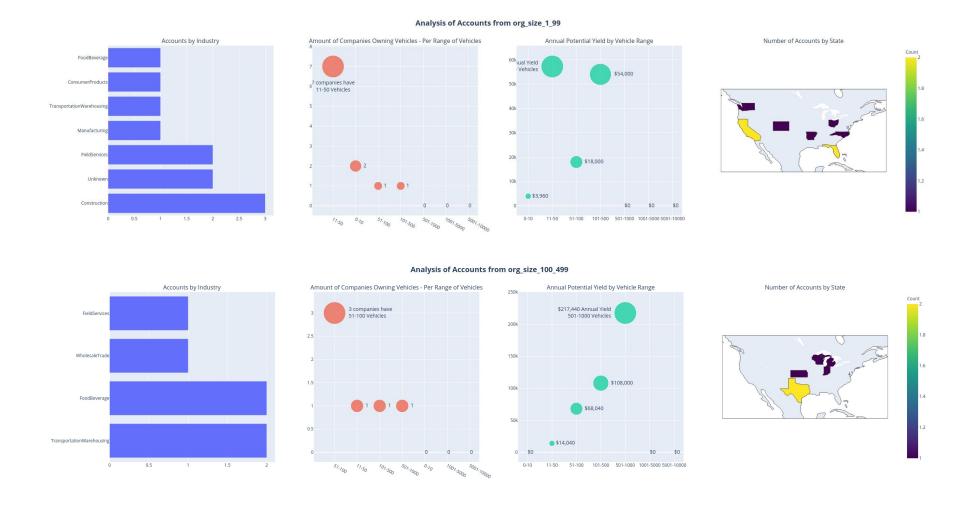
	Annual_Potential_Yield	Predicted_Proba_10	Predicted_Proba_4	Max_Proba
id_number				
1158	46800.0	0.778099	0.864448	0.864448
1369	3600000.0	0.832385	0.765538	0.832385
2868	10080.0	0.828805	0.789253	0.828805
2944	43200.0	0.754701	0.810326	0.810326
2464	66600.0	0.720021	0.797857	0.797857
1361	25920.0	0.719157	0.784024	0.784024
506	32400.0	0.691803	0.759803	0.759803
1567	5400.0	0.656276	0.754359	0.754359
2822	27360.0	0.748654	0.728473	0.748654
2792	19800.0	0.636146	0.728473	0.728473

Total of ~3.9mio USD\*

	Annual_Potential_Yield	Predicted_Proba_10	Predicted_Proba_4	Max_Proba
id_number				
2772	21600.0	0.677301	0.612369	0.677301
2796	56880.0	0.672899	0.607900	0.672899
1410	23400.0	0.594731	0.636298	0.636298
2347	54000.0	0.628731	0.561620	0.628731
1086	108000.0	0.570528	0.624376	0.624376
1042	7200.0	0.621915	0.547726	0.621915
767	7200.0	0.557432	0.612538	0.612538
775	7920.0	0.607234	0.477261	0.607234
945	9000.0	0.576478	0.528874	0.576478
911	18000.0	0.541531	0.561620	0.561620
2917	2160.0	0.506758	0.557370	0.557370
1670	14040.0	0.548986	0.499823	0.548986
1006	1800.0	0.536931	0.400959	0.536931
2797	108000.0	0.529697	0.452132	0.529697
2483	900000.0	0.510769	0.408200	0.510769
402	4320.0	0.507253	0.419324	0.507253
1882	16200.0	0.506069	0.419324	0.506069
2191	5400.0	0.468050	0.499823	0.499823
3005	23040.0	0.425038	0.442842	0.442842
1370	217440.0	0.442833	0.433591	0.442833

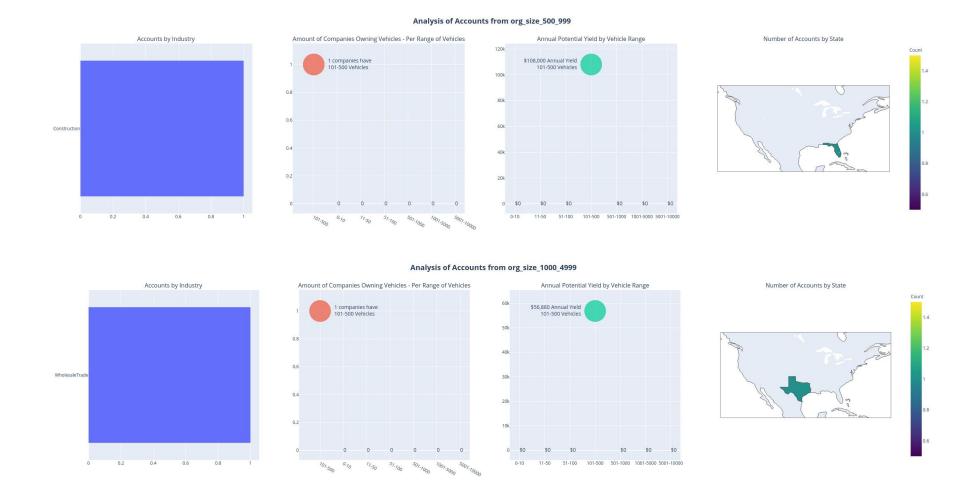
Total of ~1.6mio USD\*

# 5: Insights and Business Impact



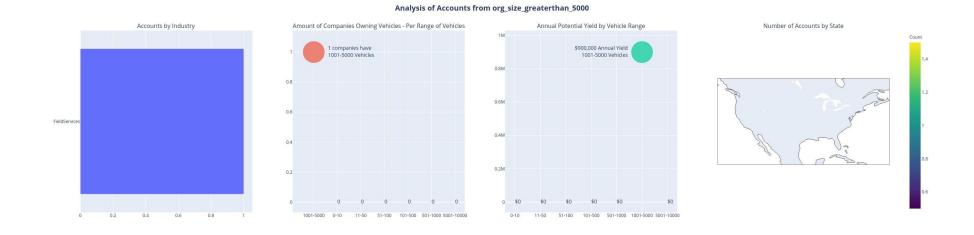
Vitoria Lima 15 of 20

# 5.1: Insights and Business Impact



Vitoria Lima 16 of 20

# 5.2: Insights and Business Impact



Vitoria Lima 17 of 20

# 5.3: Sales Cockpit



Vitoria Lima 18 of 20

#### 6: Limitations

- Sparsity of data:
  - Trained only on 250~ observations
  - Predicted only on 100~ observations
  - ...the database has ~3k companies!
- Quantitative insights into probability of accounts might not align with Sales' qualitative goals to acquire a client
  - More alignment necessary between Sales strategic goals to ensure synergic and aligned data modelling

Vitoria Lima 19 of 20

#### End of slides.

Vitoria Lima 20 of 20