Customer Term Deposit Predictor

Meagan Gardner, Elshaday Yoseph, Shell Ou, Henry Ha

Land Acknowledgement

With gratitude, we would like to acknowledge we are located on the traditional, unceded, ancestral territories of the x mə k əyəm (Musqueam), $S\underline{k}w\underline{x}w$ ú7mesh (Squamish), and səlilwəta (Tsleil-Waututh) First Nations, whose historical relationships with the land continue to this day.

Presentation Overview

• Meagan: Background and research question

• Elshaday: Exploratory Data Analysis

• Shell: Logistic Regression Model

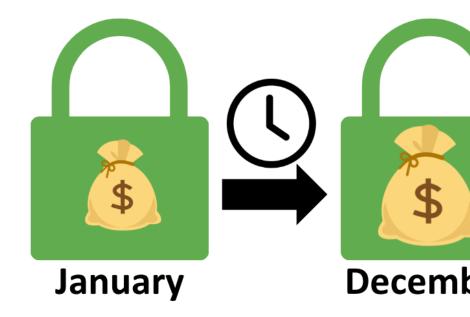
• Henry: Results and conclusion

Canada's Investment Landscape

- 79% of Canadians recognize the importance of investing
- Only 48% actually invest annually
- Over half of those that don't invest state a fear of losing money
- The Opportunity: Low-risk investment options!

What is a Term Deposit?

• A low-risk, steady way to grow your savings



ullet Keeps initial investment secure

Our Research Goal

We aim to determine whether a machine learning model can predict if a customer will agree to open a term deposit

- This will lead to:
 - More targeted campaigns
 - Cost effectiveness
 - Improved customer relationships.

Exploratory Data Analysis

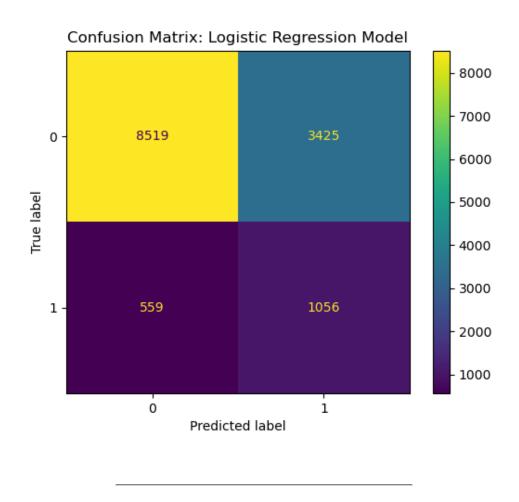
Elshaday Slides (to be updated)
Elshaday Slides (to be updated)
The Model
Shell Slides (to be updated)

Shell Slides (to be updated)

Results and Discussion

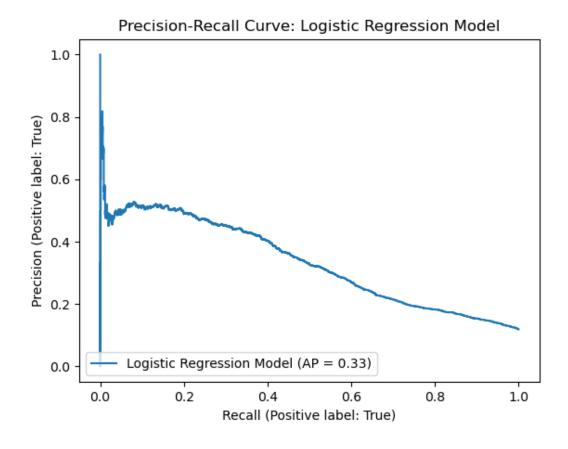
Confusion Matrix

- Key Insights
 - Strong performance for non-subscribers.
 - Low precision for subscribers (0.24).
 - Highlights true positives, true negatives, false positives, and false negatives.



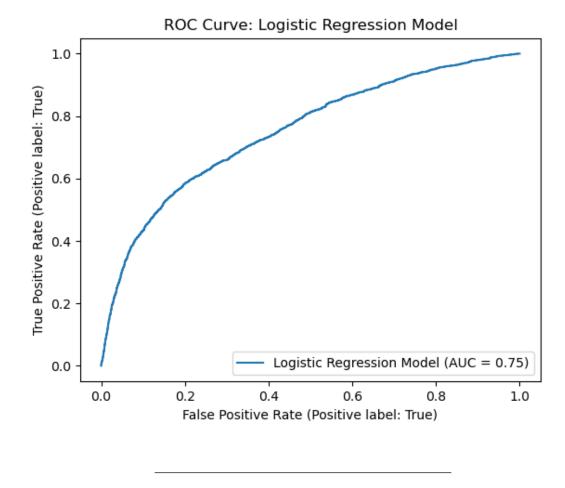
Precision-Recall Curve

- Key Insights:
 - Precision-recall trade-off is critical.
 - Impact of threshold adjustments on precision and recall metrics.



ROC Curve

- Key Insights:
 - AUC score of 0.75 suggests moderate predictive power.
 - Useful for assessing model performance across thresholds.



Limitations

- Key Points:
 - Dataset Imbalance: Significant imbalance affects precision. Precision Issues: Low precision for term deposit subscribers. Model Scope: Logistic regression may not capture complex relationships. Feature Limitations: Lack of features like transaction history impacts performance.

Practical Implications

- Applications:
 - Optimize marketing strategies using model predictions.
 - Efficient resource allocation.
- Challenges:
 - Low precision increases costs by targeting non-subscribers.
 - Improvements in precision are needed for better efficiency.

Future Research Directions

- Suggestions:
 - 1) Advanced Models: Use Random Forests or Gradient Boosting Machines to improve precision and accuracy.
 - 2) Additional Features: Include transaction history or customer engagement metrics.
 - 3) Economic Context: Integrate external factors such as interest rates or market conditions.

Questions?			