




# PREDICTING VEHICLE SERVICE INTERVALS USING MACHINE LEARNING




CMPT-3830 - Machine Learning  
Work Integrated Learning I - Fall  
2025

# Team members

Parminder Singh  
Rajveer Singh  
Jaswinder Singh  
Farhan Mohammed  
Manoj Bhatta

A decorative graphic consisting of several parallel white lines of varying lengths, slanted diagonally upwards from left to right, located on the right side of the slide.

# What we will be talking about-

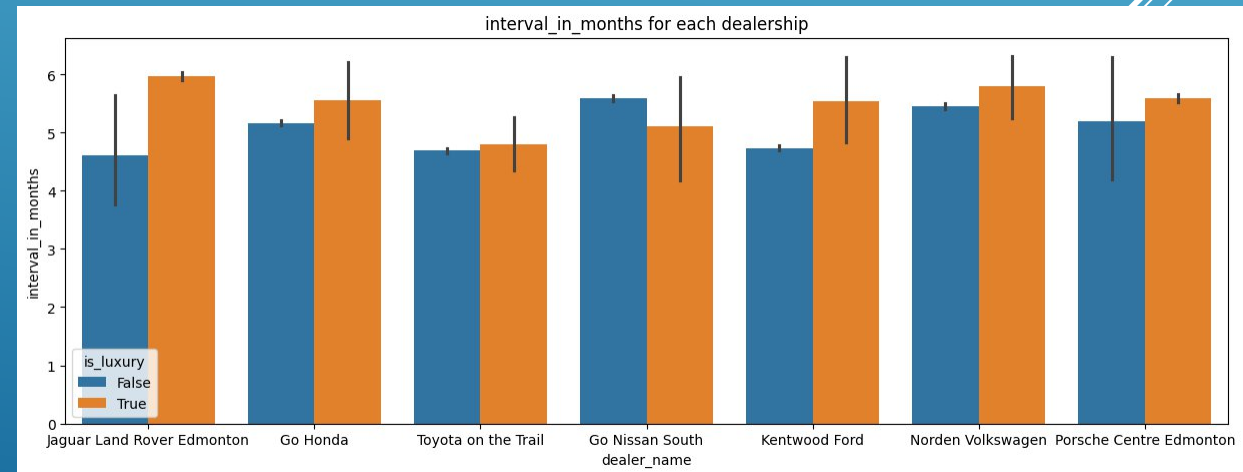
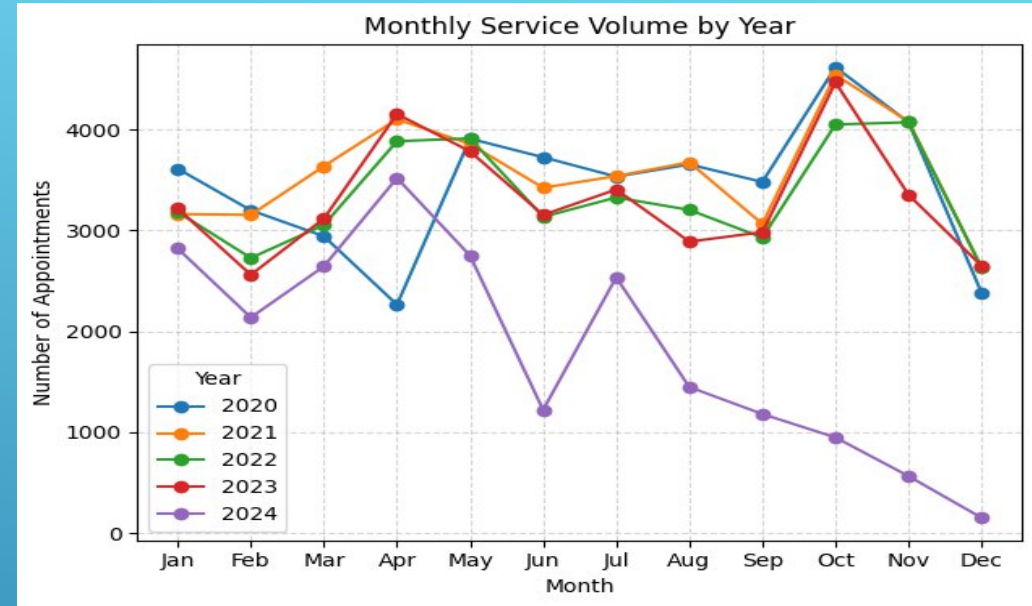
- Phase 1 highlights
  - Client's feedback for phase
  - Improvements and insights after client's feedback
  - Power Bi dashboard
  - ML application
- 
- Several thin, white, parallel diagonal lines are positioned in the bottom right corner of the slide, extending from the bottom edge towards the right edge.

**Client-** Go Auto is a large, Canadian-owned family automotive group, one of the biggest in Canada, selling new/used vehicles, parts, and service across many brands (Ford, Mercedes, Nissan, etc.) through 70+ dealerships, offering a large inventory, 7-day service etc.

**Problem Statement** - analyzing Go Auto's service records to understand service frequency, dealership patterns, and differences between luxury and non-luxury vehicles.

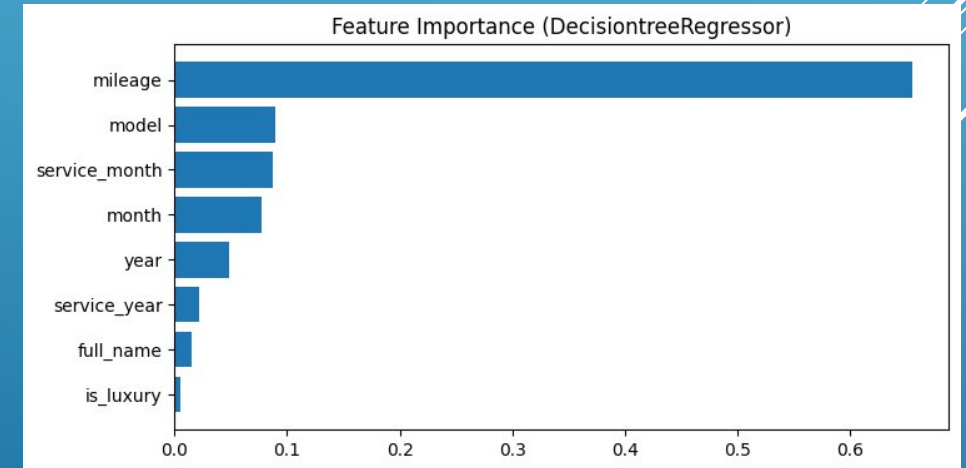
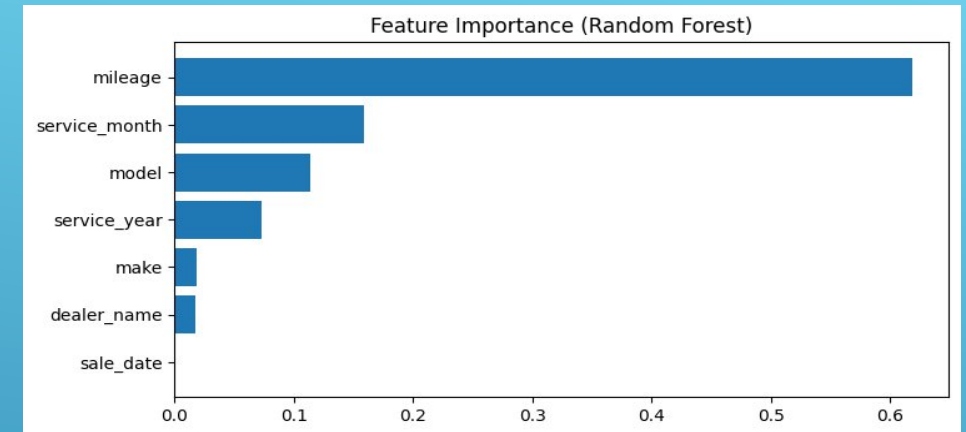
# Phase 1 highlights-

- Converted cost ranges to numeric midpoint values.
- Standardized make names using mapping dictionary.
- Dropped missing values and cleaned anomalies.
- Defined list of luxury brands.
- Created is\_luxury boolean column.
- Enabled comparison between luxury and regular cars.
- Compared luxury vs regular service intervals.
- Analyzed dealership performance and service volume.
- Explored mileage class impact on service frequency.



## Phase 1 encoding and modeling

- Converted columns to numeric types.
- Dropped irrelevant columns for modeling.
- Label-encoded make, model, dealer\_name.
- Defined X (features) and y (target: interval\_in\_months).
- Performed 80/20 train-test split.
- Models used: Linear Regression, Random Forest, Decision Tree, XGBoost.
- Evaluated using MSE, RMSE,  $R^2$ .
- Random Forest & DecisionTree performed best.

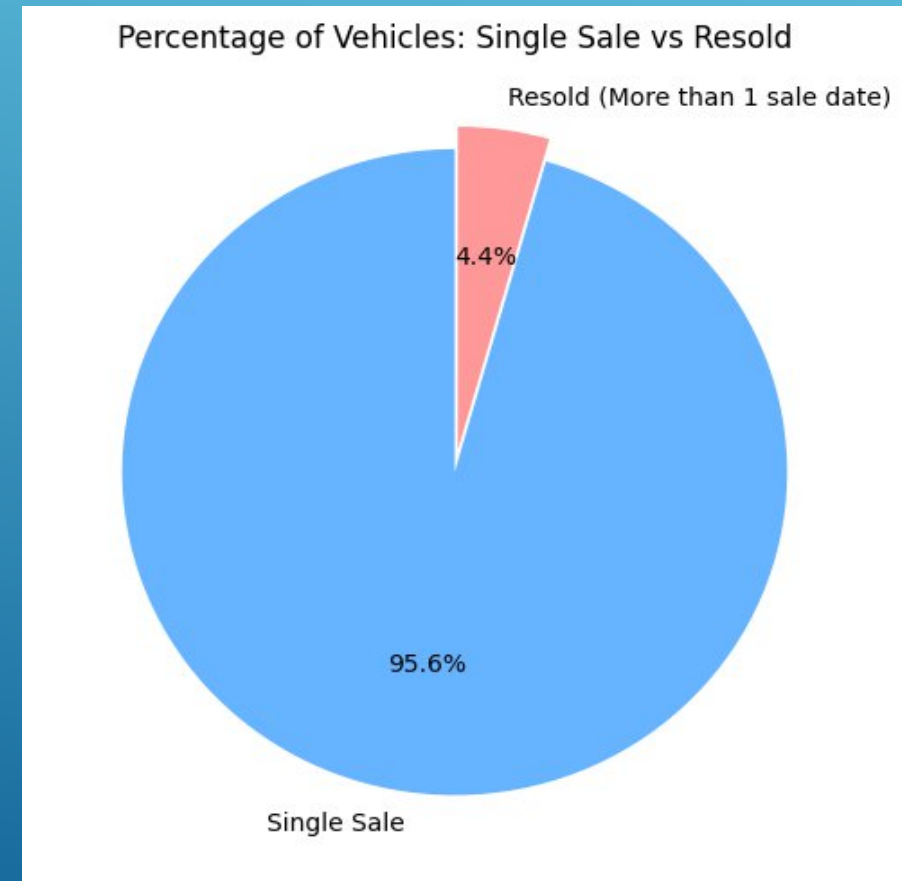
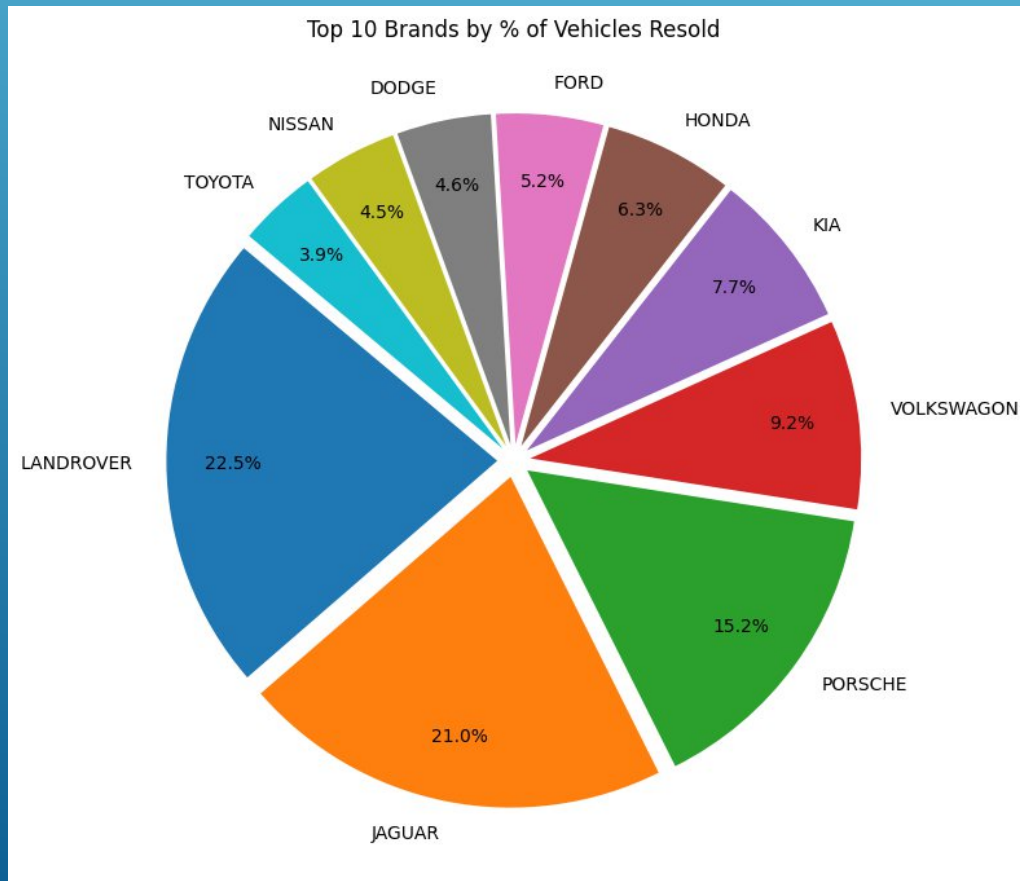


# Client's feedback:-

- Suggestion to built logic for determining if VIN is resold to another customer
- A logic for ON/OFF Make to check if same VIN brand comes to same brand dealership

## Work and insights after client's feedback:-

- Built a model to know if vin is resold and get to know almost 1747 VINs were resold.

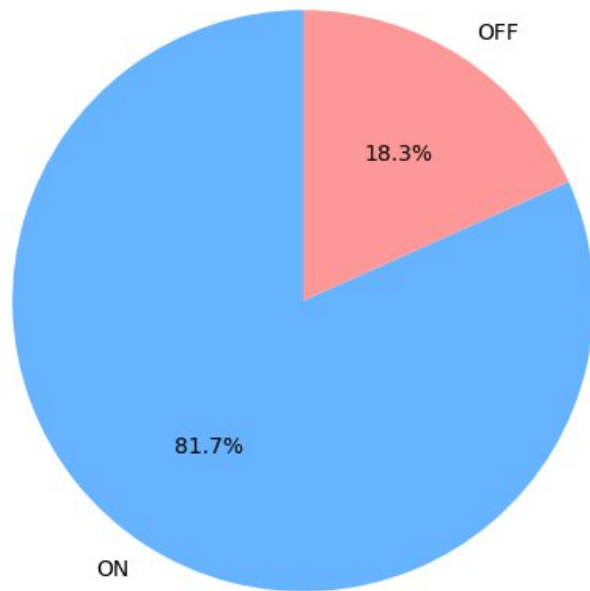




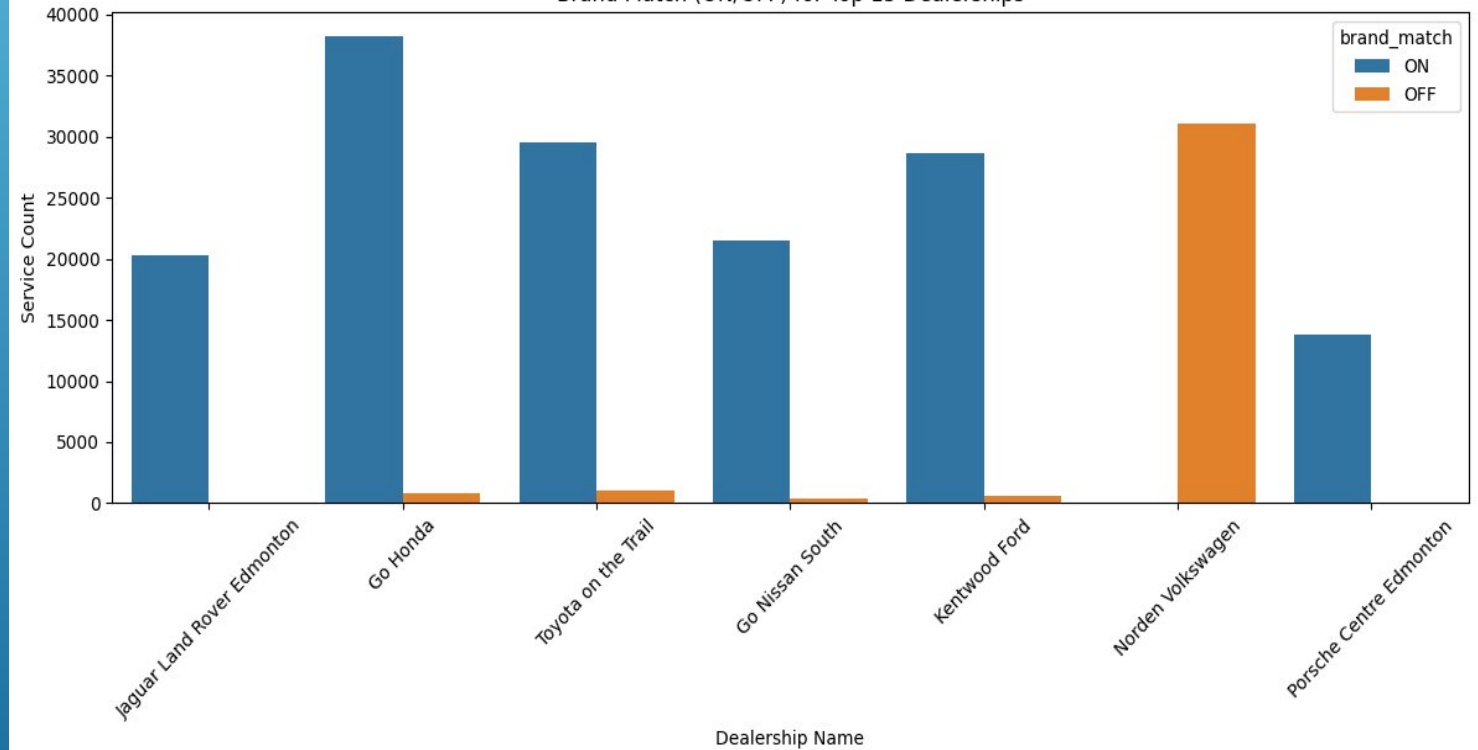
## ON/OFF Make function and insights:-

- Built a function to check
- About 81 percent of vehicle go their brand dealership for service.


Brand Match: Vehicle Brand vs Dealership



Brand Match (ON/OFF) for Top 15 Dealerships



## Phase 2:-

- Only used VINs which came back at least 1 time for service
  - Finalized RandomForest model and trained it with necessary features.
  - Built Streamlit application for prediction
  - Built Powerbi dashboard
- 
- A series of white diagonal lines of varying lengths and thicknesses, located in the bottom right corner of the slide, creating a modern, abstract graphic element.

dealer\_name

- ☐ Go Honda
- ☐ Go Nissan South
- ☐ Jaguar Land Rover Edmonton
- ☐ Kentwood Ford
- ☐ Norden Volkswagen
- ☐ Porsche Centre Edmonton
- ☐ Toyota on the Trail

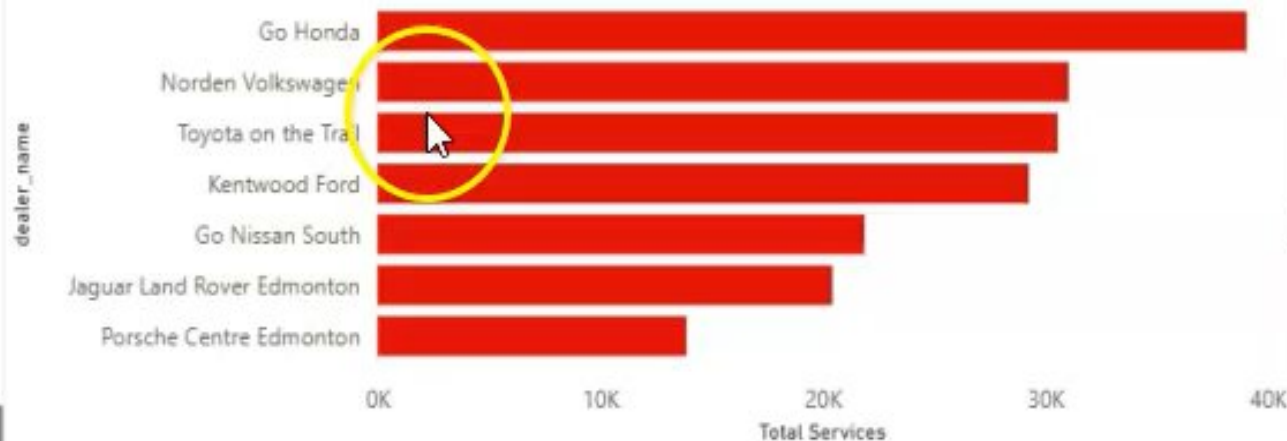
make

- ☐ ACURA
- ☐ ARCTIC MOTORS
- ☐ ASTONMARTIN
- ☐ AUDI
- ☐ BENTLY
- ☐ BMW
- ☐ BUICK
- ☐ CADILLAC
- ☐ CHEVROLET

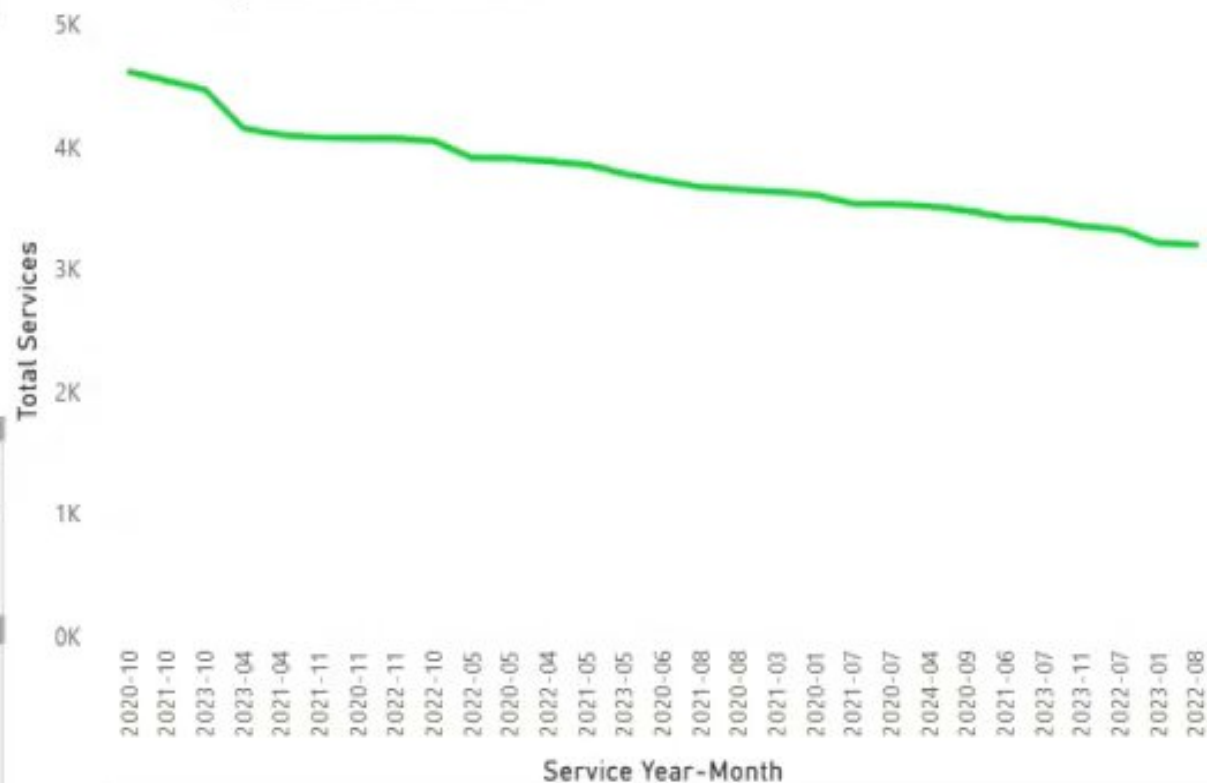
Service Year-Month

- ☐ 2020-01
- ☐ 2020-02
- ☐ 2020-03
- ☐ 2020-04
- ☐ 2020-05
- ☐ 2020-06

Total Services by dealer\_name



Total Services by Service Year-Month



186K

Total Services

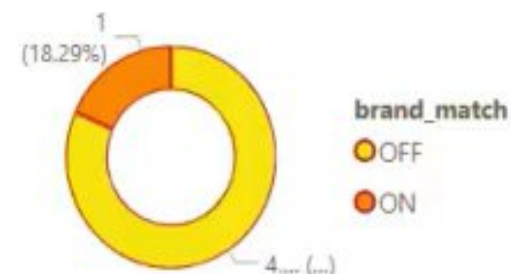
4.67

Median Interval (Months)

461.84

Average of cost

Brand Match % by brand\_match



Navigation

Go to:

- Overview
- EDA
- Analysis
- Prediction

# Go Auto Service Interval – Overview

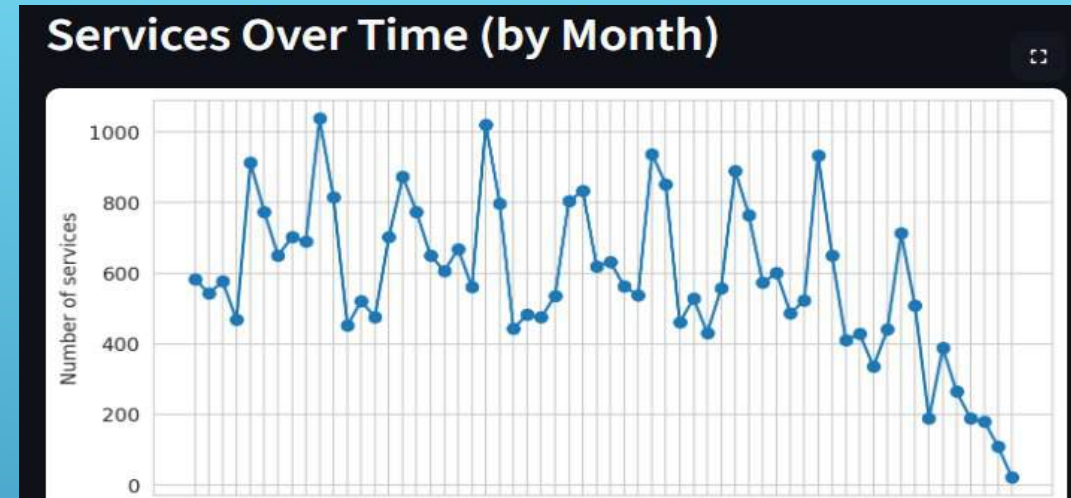
Total records	Unique VINs	Number of brands	Avg interval (months)
186,088	39,421	39	5.24

## Sample of data used to train the model

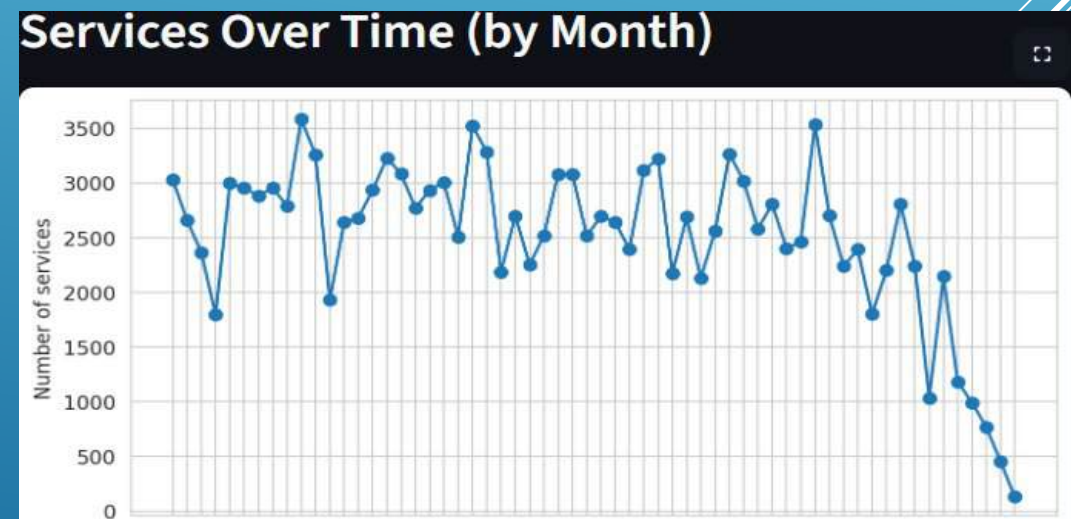
	service_date	make	vin	mileage	interval_in_months	is_luxury	dealer_name	cost
0	2021-11-30 00:00:00	LANDROVER	1000167976	14656	2.7333	<input checked="" type="checkbox"/>	Jaguar Land Rover Edmonton	50
1	2022-02-20 00:00:00	LANDROVER	1000167976	18253	7.7333	<input checked="" type="checkbox"/>	Jaguar Land Rover Edmonton	450
2	2022-10-10 00:00:00	LANDROVER	1000167976	33194	8.4333	<input checked="" type="checkbox"/>	Jaguar Land Rover Edmonton	650
3	2023-05-29 00:00:00	HONDA	1000197869	9342	1.8333	<input type="checkbox"/>	Go Honda	150
4	2023-07-23 00:00:00	HONDA	1000197869	15255	8.7333	<input type="checkbox"/>	Go Honda	750
5	2024-04-10 00:00:00	HONDA	1000197869	26453	1.8667	<input type="checkbox"/>	Go Honda	550
6	2024-06-05 00:00:00	HONDA	1000197869	29500	5.6667	<input type="checkbox"/>	Go Honda	50
7	2021-06-05 00:00:00	TOYOTA	1000350530	73090	2.4	<input type="checkbox"/>	Toyota on the Trail	150
8	2021-08-16 00:00:00	TOYOTA	1000350530	74322	9.5	<input type="checkbox"/>	Go Nissan South	50
9	2023-10-03 00:00:00	TOYOTA	1000402957	7209	0.4667	<input type="checkbox"/>	Toyota on the Trail	

# Conclusion

- Overall dealerships have more regular vehicles coming for service.
- Less number of luxury vehicle services is cause of less ownership and different drive behavior than regular cars.
- Mileage is dominant factor for car owners to return for service if no other issue.
- Still both luxury and regular vehicles comes around same time period.




Luxury vehicle graph



Regular car service graph



# Reflection

- Mostly return time is based on mileage and cost can be somewhat factor for some dealerships.
  - Feedback helped us improve and get some new insights.
  - In future, with more detail like service type can be helpful to get better prediction for return time of customer.
  - Exploring more about each customer driving pattern can be helpful for future improvements.
- 
- A series of white lines of varying lengths and slopes are positioned in the bottom right corner of the slide, creating a modern, abstract graphic element.

# Thank you Go Auto

For giving us opportunity to get familiar with real life experience of data science world problems and motivating us to improve.

