

# Project Report

## **New Car Buy Recommender System**

Masters of Technology in Intelligent Systems

Module: Machine Reasoning

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## Executive Summary

We, Singapore citizens and residents, are known globally for our 5 C desires- car, condo, cash, credit card, and country club. Car ownership in Singapore for Singaporean Residents is a cherished aspiration. [Ref 1] captures the sentiment in international press. Closer home, Singaporean love affair for car ownership is evident from weekend newspaper advertisements.

However, Car ownership is a costly affair for us, primarily because of the need to control traffic, ranking Singapore as one of the costliest places to own and maintain a car. [Ref 2]

Our “love” affair with car ownership is so skewed that MAS (monetary authority of Singapore) has to step in to enforce financial prudence in Singaporeans. [Ref 3]

A wide range of choices of car brands offering a bewildering range of possibilities across ride, comfort, safety, gadgetry etc are available to Singaporeans. This is thanks to our open and services driven economy.

Above all aspects, makes the buying decision process a very confusing and suboptimal affair, leaving much room for improvement.

Our project, New car buy recommendation system, is an effort to improve the car buy decisioning system in Singapore. Our recommendation system today suggests car budget and car make/type to look for, thus narrowing down the choice process quite significantly.

We used a combination of expert interviews as well as carefully designed surveys to define a set of rules that help make optimal decision on car buying. To capture the fuzziness in buying decision, we reinforced thus defined/mined rules with the concept of certainty factor.

We used the KIE for user input capture, KIE DROOLS for decision rule, and additionally Orange pack as well as Excel, Tableau and Google sheet analytics for offline data exploration and data mining.

The solution we have presented is a first attempt to bring objectivity and optimality to car the decisioning process. Hence, we prepared a roadmap to capture further objectivity and augmented intelligence in car buying process.

Our achievement is happy Singapore car buyers, where they find the best ride suited to their needs and aspirations that the money they can realistically afford!

## Problem

### Definition

Which Car should I buy? Car Features, Japanese/European, fuel efficiency, second car, family needs are just some of the fuzzy questions that a car buyer goes through? What is the best decision considering various individual, family, societal needs, financial constraints Vs car ownership costs.

Above is the predicament household question in Singaporean car buyer minds. Unfortunately, no clear and robust decisioning system at hand.

Hence, we formulated the problem statement, thus:

***A prospective car buyer weighs in various questions like above, and makes a subjective decision which we observe is not the most optimal.***

### Description

To further describe the problem involves an appreciation of following aspects for the problem space:

1. The buying process is subjective and hence fuzzy. The concept of Certainty Factor has to be introduced.
2. There is sufficient synthesized human knowledge on car buying that must be captured as part of decisioning system
3. Buying process has “crowd wisdom” and nuggets of knowledge have to be unearthed. Given that the buying process entails whole population of Singapore, hence surveys are clearly must and needed.

### Scoping for the project

#### Inclusions

##### *Business*

1. Only Singapore market car buying behaviour and decisioning constraints are captured. Application formulation does not apply to other countries.
2. Singapore New car buying process is captured
3. Today's financial regulations on car loan
4. Expert interview and survey data mining

### *Technical*

1. Application relies on making strong inferences from the survey and single expert interview (this is clearly an inferential limitation)
2. Application is offline today, based on KIE rules engine
3. Offline inclusion of survey data mining results

### *Exclusions*

1. Not recommending individual model make (E.g. Toyota Camry 2.5 Hybrid Vs Mercedes Benz C-Class 180 Avantgarde). Models pricing and Model inventory/life cycle are volatile.
2. resale Car buying process
3. web interactive and public domain
4. Differentiation between car owners and prospective car buyers not made

Please review the roadmap recommendations to address some of the exclusions listed above.

## Solution

The solution involves a three-step knowledge definition and synthesis process:

1. Acquisition & Discovery
2. Structuring
3. Modelling & Representation

### Knowledge Discovery & Acquisition

Knowledge Acquisition constituted following 3 step process:

- a) Step 1: Self-study and self-articulation: There are two areas of knowledge acquisition here:
  - a. Individual and family/Friends buying experience and post facto (post buying challenges). This formed the dimensions of decision making process, the dimensions nature – whether definite or fuzzy contributors to buying logic. This step also involved reviewing government rules and regulations regarding loan eligibility [Ref 3] and minimum cash down for car buying.
  - b. Leading car portals: [Ref 5]. We leveraged SGCarMart portal and paid special attention to their car comparison and car catalogue pages. These give curated car buyer knowledge acquisition process
- b) Step 2: Expert feedback: Equipped with our understanding and appreciation of car buying process, its dimensions and challenges, we sought feedback of a Subject Matter Expert. [Ref 4]. We conducted an in-depth interview conducted with Jay, who carries a sound decade plus continued experience in Singapore car buying market as a senior new car sales manager. Synthesizing the interview outcomes, we:
  - a. Confirmed evident rules in step 1
  - b. un-earthed un-conventional buying rules, amounting to 100% certainty factor. We also identified “grey” buying areas in our interview. Besides, we got a good understanding on finances and finance behaviour related to car buying process. E.g. women buyers predominantly look for fuel-efficient cars.

Row Labels	Sum of Fuel	Count of Gender	
Female	15	16	94%
Male	30	42	
(blank)			
<b>Grand Total</b>	<b>45</b>	<b>58</b>	

*Table 1: Female drivers a CF\_features\_economy = 1*  
(see table 3 for the whole list of CFs)

- c) Step 3: End-user Survey: Having understood key matters related to car buying process, and also knowing grey areas in our thus acquired knowledge above, we went to survey creation. Sample Survey questionnaire is in [Ref 6]

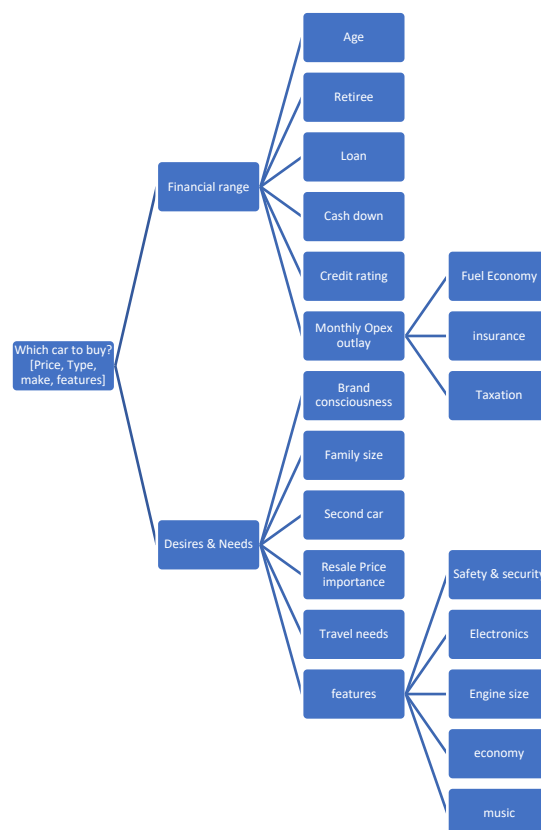
In Summary, our project’s knowledge acquisition tree looks like below:



(Figure 1: Knowledge Acquisition for car buying recommendation system)

### Knowledge Structuring – Dependency diagram

Synthesized Knowledge from Knowledge discovery and acquisition process is represented by dependency diagram (figure 2) below.



(Figure 2: Dependency diagram: how to buy a new car in Singapore)



## Knowledge Modelling & Representation

### Buying budget behaviour and calculation: Derived Rules

Car budget recommendation is based on knowledge rules extracted from 3 step processes of Knowledge discovery and acquisition. Here following facts are important for defining the recommendation on car buying budget:

- Singapore regulation stipulates a maximum 70% loan. Min 30% must be cash down
- Financial lending companies and car companies ecosystem brings more incentives if a minimum 30% percentage of car value is taken as loan
- Younger people generally prefer maximum loan
- Older people generally prefer no to minimum loan

Accordingly, we defined, confidence factor (CF) to model end buyer fuzziness in financial clarity Vs the observed behaviour in field and government rules. The CF defined are:

1. CF to take 70% loan = CF\_70\_loan
2. CF to take 30% loan = CF\_30\_loan

Based on user inputs, CFs are updated based on a combination of survey and expert rules.

For the three identified age brackets, the mappings are as follows:

Scenario	Customer age group mapping	CF_70_loan	CF_30_loan
Scenario 1	Age > 40 & Age < 50	0	0
Scenario 2	Age > 50	-1	0.8
Scenario 3	Age < 40	1	-0.5

*Table 2: Mapping of Age to CF 70/30*

Associated car budget logic for each of the scenarios:

#### Scenario 1:

IF

(CF\_70\_loan == 0 && CF\_30\_loan == 0)

THEN

Price\_Max = Cash\_Max/(1- loan\_max)

Price\_Min = Cash\_Min/(1-loan\_min)

Where Max/Min represents user inputs/gut feel on max min loan and cash down

#### Scenario 2:

```

IF
    (CF_70_loan <= -0.5 && CF_30_loan >= 0.5)
THEN
    Price_Max = Cash_Max
    Price_Min = Cash_Min/(1-30%)

```

### Scenario 3:

```

IF
    (CF_70_loan >= 0.5 && CF_30_loan <= -0.5)
THEN
    Price_Max = Cash_Max/(1- 70%)
    Price_Min = Cash_Min/(1-70%)

```

### Certainty factors

As discussed in earlier sections, the buying process is inherently fuzzy and is inclusive of multiple factors. Hence, we have defined the following variables to capture the fuzziness.

Certainty Factor (CF)	Explanation (leaning towards)	min	max	initialized to
CF_30_loan	taking 30% loan	-1	1	0
CF_70_loan	taking 70% loan	-1	1	0
CF_Make_EUR	a European make car - e.g. Mercedes	-1	1	0
CF_Make_JP	a Japanese make car - e.g. toyota	-1	1	0
CF_Make_MY	a malayisa make car	-1	1	0
CF_Make_KOR	a korean make car	-1	1	0
CF_Make_American	an American make car	-1	1	0
CF_Type_Hatch	a hatchbatch e.g. Honda Jazz	-1	1	0
CF_Type_SUV	an SUV e.g. Subaru forester	-1	1	0
CF_Type_MPV	an MPV like Honda Stream	-1	1	0
CF_Type_Sedan	a Sedan e.g. Toyota camry	-1	1	0
CF_Type_Stationwagon	a stationwagon e.g. Volvo V60	-1	1	0
CF_features_SS	safety and security leading car	-1	1	0
CF_features_economy	higher fuel economy car	-1	1	0
CF_features_engine	higher engine performance i.e. high Torque, l	-1	1	0
CF_features_gadgets	electronic gadgets like vision, car parking aid	-1	1	0
CF_features_music	higher quality music instruments	-1	1	0

Table 3: Certainty factor for the car recommender system

### User inputs:

Car recommendation system end user inputs are captured as follows:

Input	Description	Data type	Initiatized to	Values
Age	Age of buyer	numeric	18	18-80
Gender	Gender of buyer	Binary	0	M = 1, F = 0
Num_People_family	Number of family members in the hous	Numeric	0	
is_second_car	is the car being bought second car	Binary	0	0= 1st car, 1= 2nd car
Cash_min	Min cash user thinks has for car buy	Numeric	0	0 to 500000
Cash_max	Max cash user think has for car buy	Numeric	0	0 to 500000
Loan_min	Min loan user think shall need for car b	percent	0%	0%-70%
Loan_max	Max loan user think shall need for car b	percent	70%	0%-70%
Weekly_Usage	Weekly car usage	Binary	0	0 = low usage, 1 = high usage
Brand_Consciousness	Is brand consciousness a factor in decis	Binary	0	1=brand conscious, 0 = not
monthly_car_expenses	Monthly expense include fuel, maintena	Binary	0	1= high, 0 = low

Table 4: User inputs table for the car recommender system

Car recommendation system application outputs:

Car recommendation system outputs are captured by following table:

#### Defined Rules

Based on Expert interview and Survey results (Certainty Factor), below diagram's capture the evolution of the CFs based on inputs (table 4) given by an end user of the car recommendation system application.

First, a few examples of user survey derived Certainty factors:

#### Example 1: user survey derived CF recommendation

Sum of Fuel	Column Labels							
Row Labels	American	European	Japanese	Korean	Others	(blank)	Grand Total	
25-40			8	7	4	5	24	
40-50	1	2	9	3	2		17	
50 and above		1	3				4	75%
(blank)								
Grand Total	1	11	19	7	7		45	

Table 5: >50 years old users car make preference

#### Example 2: user survey derived CF recommendation

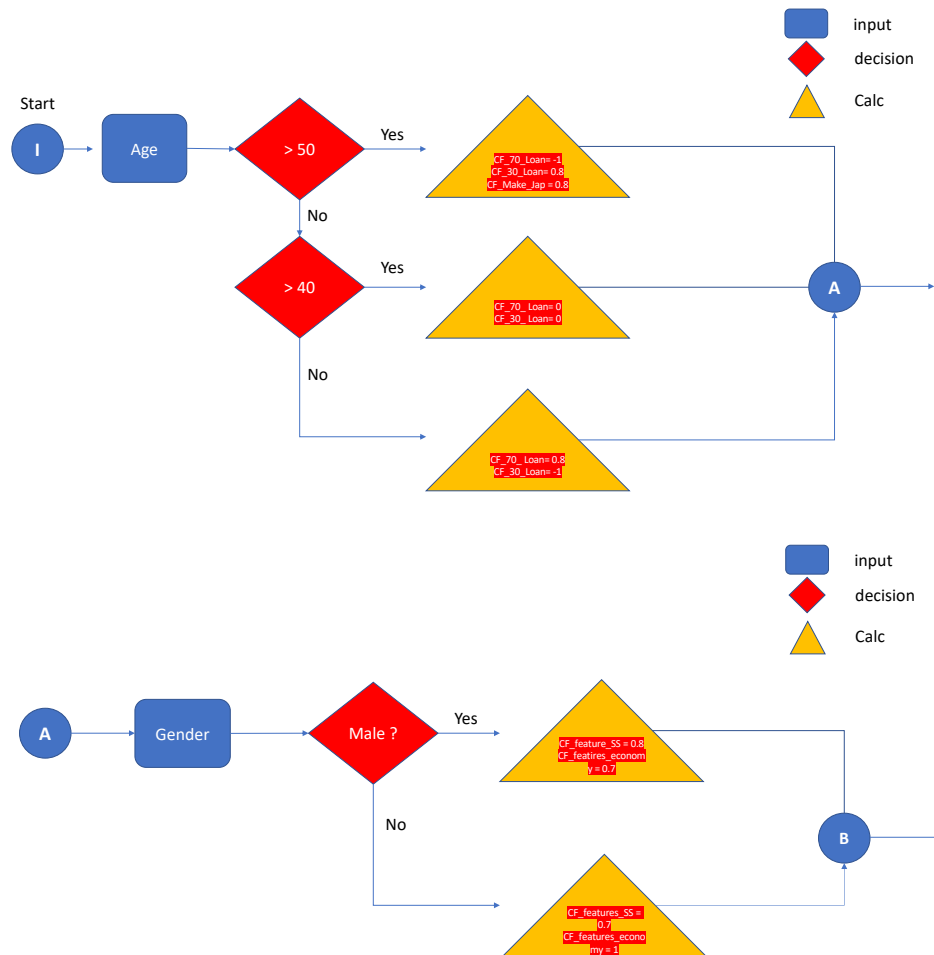
Older people prefer Japanese, resulting in derived rule CF\_Make\_Jap = 0.8 (see detailed flowchart in tab

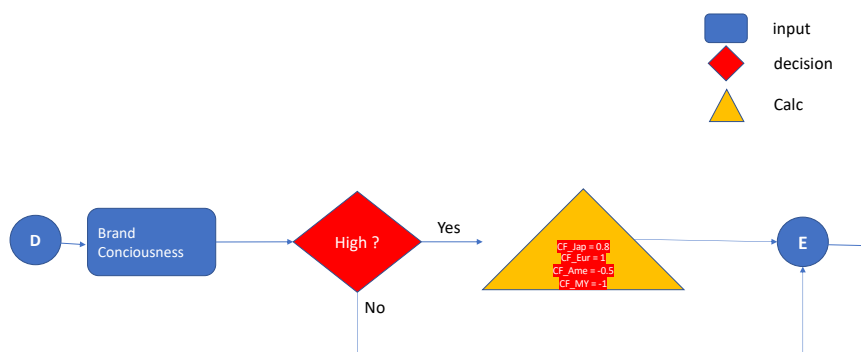
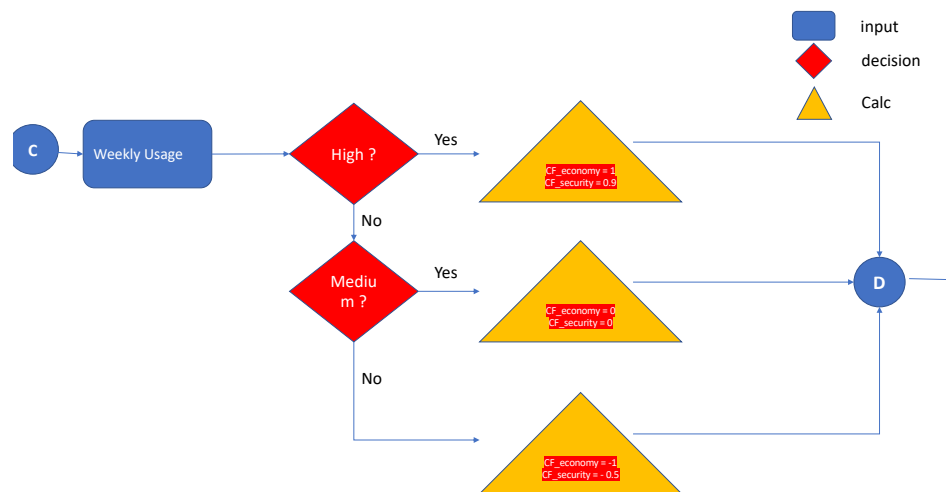
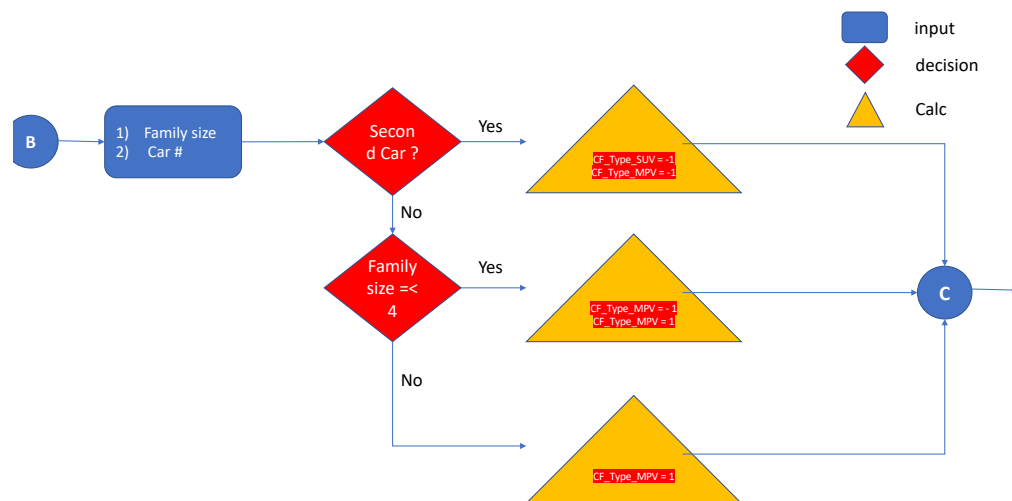
Row Labels	Sum of SS	Count of Gender	
Female	11	16	
Male	36	42	86%
(blank)			
Grand Total	47	58	

Table 6: Males prefer safety and security.

Refer table 1 for female preference

Finally, here is the flowchart for user input based CF influence.





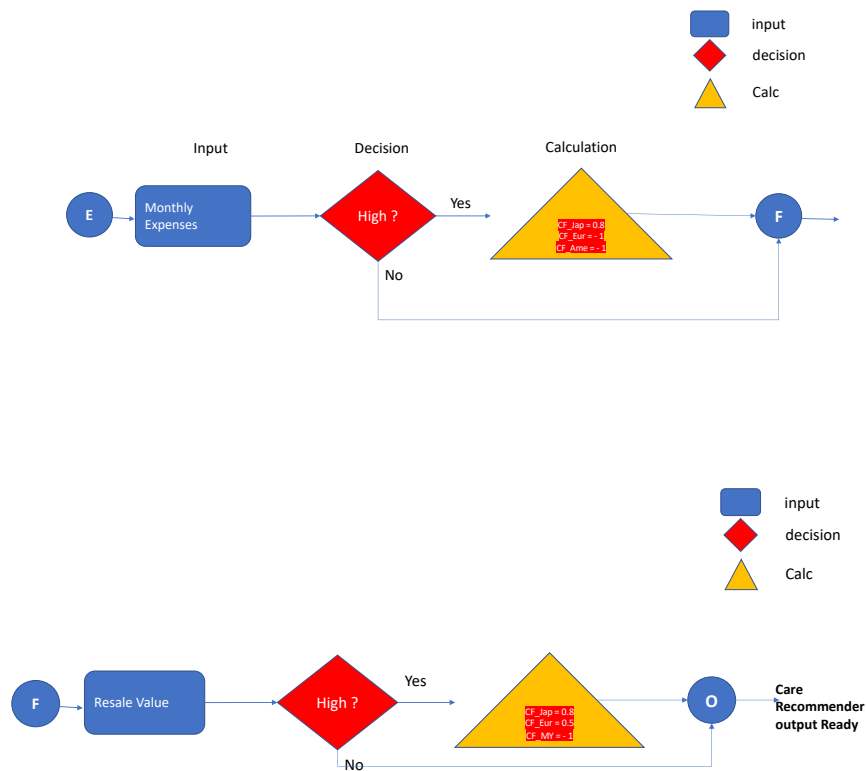


Table 7: Flowchart for Rule Definition engine to update the Certainty Factor

Finalized CFs will then generate the recommendation to the car recommendation system application end user.

Car recommendation system Outputs

Ouput	Output bracket	Possible values	Comments
Price	Price Min	Minimum car price	car price bracket
	Price Max	Maximum car price	car price bracket
Car country make	Country of car make	Japenese	In order of priority
		European	in order of priority
		Korean	in order of priority
		American	in order of priority
		Malaysian	in order of priority
Car Type	Type of car	Sedan	In order of priority
		SUV	In order of priority
		MPV	In order of priority
		Hatchback	In order of priority
		Stationwagon	In order of priority
Car Features	Recommended car features	Safety and securit	yes no
		Fuel Economy	yes no
		Engine	Yes no

Table 8: Output setup for the car recommendation system (in current version)

### System Design:

In today's MVP stage, the system logic is defined in KIE. This includes definition of the

- User input form (as per the user input table defined in Table 4)
- Task definitions and Data Objects
- Process flow including input corrections/data correction
- And finally guided decision tables to provide the end recommendations (as per table 8 for car recommendation system output)

### System Architecture considerations:

There are no system architecture considerations as we are in MVP stage. For the application to be launched to end users, following developments have to be defined:

- Mobile and web interface for user interactivity (Input and Output)
- Web server hosting (for the MVP functionality and logic)
- Security provisions for unauthorized access, DDoS etc attacks

## Inputs & Output dialogue screens

### User Input Form

Car Recommendation System  
User Input Box

Swipe to learn more

Age	>	Weekly Usage	>	Loan Min %	>
Gender	>	Brand Consciousness	>	Loan Max %	>
Number of family members	>	Monthly car exp	>	Cash Down Max	>
First or second car?	>			Cash Down Min	>

>

Swipe over to input

Submit

### Recommendation Screen

Car Recommendation System  
Recommendations for you based on your inputs & our AI engine

Swipe to learn more

We recommend you look for car in this budget range

Car Budget (max)

Car Budget (min)

We recommend that you will be satisfied with this make

Car Make

We recommend that you will be satisfied with this type

Car Type

We recommend following features in the car

Car feature 1

Car feature 2

Car feature 3

Coming soon – top three car models that will fit your needs best !

Feedback please



## Project Critical Analysis

### Key Survey and Expert Findings

- Housing type does not matter in car selection as was based on expert feedback
- Gadgets and Music does not matter much to the survey population. We believe that it's due to a reasonably OK music system in most of the cars in market, and is not a buying factor anymore.
- Women overwhelmingly look for 'fuel economy' of car
- Men overwhelmingly look for 'safety and security' in the car.
- Retirees overwhelmingly look for Japanese car
- The application does not consider the working status because it was explicitly suggested by expert that it does not influence car buying behaviour. However, detailed data based studies can be further taken to explore alternate hypothesis.
- Most of the buyer base is 40-50 years bracket
- Retirees mostly want to buy car with full cash down payment
- Young overwhelmingly take maximum loans

### Functionality and system short comings, pending issues

- Distinction of European cars – wide range of quality difference between European car brands
- Distinction between sedan categories – ideally luxury and mass market sedans are different markets
- specific tastes like Racing, Vintage cars
- Resale value of the car was not captured in the survey. While heuristics is available, i.e. Toyota and Mercedes fair well on resales value in Singapore market, since there was no supporting data from survey we have excluded it from user inputs. However, based on user inputs, we factor based on calibration.
- Next best alternative (NBA), top car make, type suggestions not implemented. E.g. Toyota Camry 2.5 Hybrid is the top recommendation based on inputs and knowledge decisioning system (i.e. Car recommendation system)
- Weekly/Daily fluctuations due to COE, flash sales
- OPC (off-peak car) or full COE based recommendations
- Ability to give meaningful recommendation with users providing only partial information
- System handles only male and female preferences (Transgender and other social /family structure preferences not curated)
- For retirees, we have not considered loan guarantors for loan. This is based on expert advise as well that most (~100%+) do cash down buying.

### System Limitations

- Lacks Visual appeal due to using KIE forms.

- User interactivity limited to capability afforded by KIE forms.
- System can't go live online (public internet domain) due to KIE capabilities and Public domain/public server hosting and budgets
- Inputs that shall have a finite range to represent, hence choices are made binary (e.g. security preference that can range from low, medium, high or on a scale of 1-10)

## Future Project Enhancements

### Scope Enhancements

- Singapore Used car market inclusion
- Expanding recommendations from car make, type, features, to top matching 3 cars (NBA) available in the Singapore market at that given point in time
- Expanding to nearby countries, e.g. South East Asian markets

### Functional Enhancements

- Representative surveys for Singapore car buying target population
- Machine learning module augmentation to identify
- Additional experts feedback
- Hypothesis testing scenarios

### Technical Enhancements

- Launching the mobile interface and web server (AWS) with a scale up on demand backend i.e. AWS IaaS/PaaS.
- Machine Learning module for offline integration based on model re-runs (surveys and user feedbacks) with AWS sagemaker.
- Integration with car dealers (e.g. C&C, Borneo motors etc) and car information aggregation websites like sgcarmart.com

## References

[Ref 1]: Singaporean love for cars

<http://money.com/money/4716340/worlds-most-expensive-city-cost-of-living-car-ownership/>

[Ref 2]: Singapore is consistently the costliest place to buy car

<https://www.startrescue.co.uk/news/motoring-news-2/10-most-expensive-countries-to-buy-and-run-a-car>

<https://www.sgcarmart.com/news/article.php?AID=10371>

[Ref 3]:

Singapore government regulations on car loans

<http://www.mas.gov.sg/News-and-Publications/Media-Releases/2016/MAS-Eases-Rules-on-Motor-Vehicle-Financing.aspx>

[http://www.ifaq.gov.sg/MAS/TOPICS/BANKING/Vehicle Loans/7419#FAQ\\_176191](http://www.ifaq.gov.sg/MAS/TOPICS/BANKING/Vehicle%20Loans/7419#FAQ_176191)

[Ref 4]:

Expert interview: Jay, Singaporean Car Sales manager with more than a decade of successful car selling experience in Singapore. Audio transcripts submitted.

[Ref 5]:

SGCarMart.com: is the leading Singapore car information portal.

[Ref 6]:

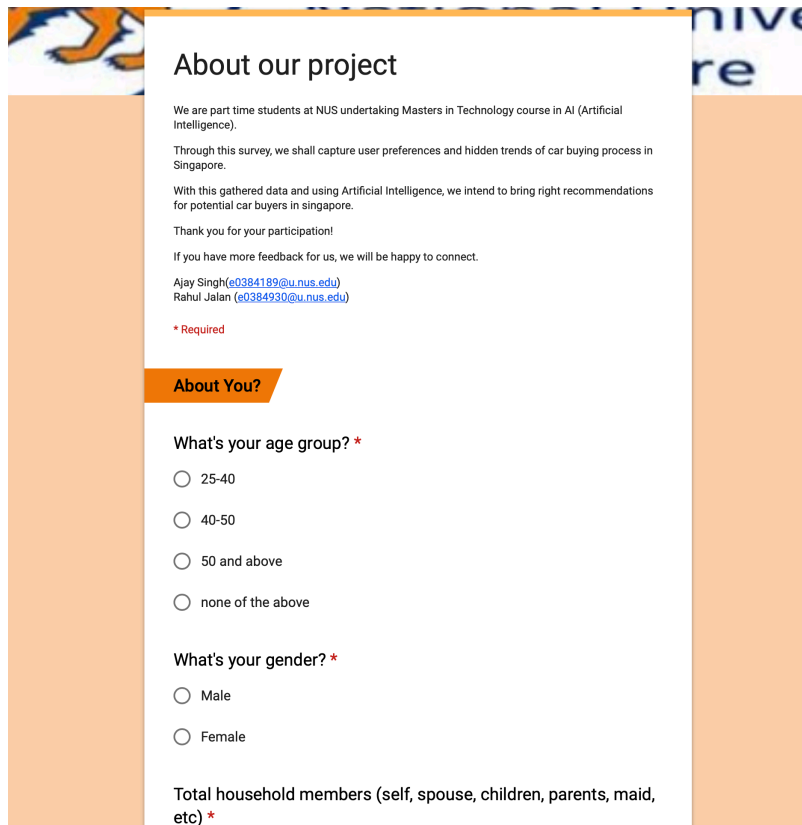
Survey to collect end buyer behaviour

<https://goo.gl/forms/Frt1Rh3q9ImSXXkY2>

## Appendices

### Appendix A: Survey

#### Survey Form:



**About our project**

We are part time students at NUS undertaking Masters in Technology course in AI (Artificial Intelligence).

Through this survey, we shall capture user preferences and hidden trends of car buying process in Singapore.

With this gathered data and using Artificial Intelligence, we intend to bring right recommendations for potential car buyers in singapore.

Thank you for your participation!

If you have more feedback for us, we will be happy to connect.

Ajay Singh([e0384189@u.nus.edu](mailto:e0384189@u.nus.edu))  
Rahul Jalan ([e0384930@u.nus.edu](mailto:e0384930@u.nus.edu))

\* Required

**About You?**

**What's your age group? \***

☐ 25-40

☐ 40-50

☐ 50 and above

☐ none of the above

**What's your gender? \***

☐ Male

☐ Female

**Total household members (self, spouse, children, parents, maid, etc) \***

Section 2 of 3

About your car finance /(or your future car)

Description (optional)

How much cash-down did you pay for your last car? (/or you will pay for your future car) \*

☐ 20,000 - 30,000 SGD

☐ 30,001-50,000 SGD

☐ 50,001-80,000 SGD

☐ 80,001 - 100,000 SGD

☐ More than 100,00 SGD

☐ Will not like to reveal

What % loan did you take on your last car buy? (/or you will pay for your future car) \*

Section 3 of 3

About your ride / (Or your future car)

Description (optional)

What car you own today? (/Or plan to own) \*

☐ European

☐ Japanese

☐ Korean

☐ American

☐ Others

What's your car type? (/ or would be) \*

☐ MPV

☐ SUV

☐ Sedan

Survey raw data:  
Details attached in project zip folder

Excel File Edit View Insert Format Tools Data Window Help										
AutoSave ON										
Home Insert Draw Page Layout Formulas Data Review View Tell me what you want to do										
F10										
	A	B	C	D	E	F	G	H	I	J
1	Timestamp	What's your age group?	What's your gender?	Total household members (self, spouse, children, parents, maid, etc)	Are you a retiree?	Do you own a car	How much cash-down did you pay for your last car? (/or you will pay for your future car)	What % loan did you take on your last car buy? (/or you will pay for your future car)	How much are your other monthly car expenses? e.g. petrol, car tax, insurance (/or you will pay for your future car)	What car you own today? (/Or plan to own)
2	2019/03/03 8:08:14 PM GMT+8	40-50	Male	5-6	No	No	30,001-50,000 SGD	50-70%	medium (200-400 SGD)	Japanese
3	2019/03/03 8:10:27 PM GMT+8	40-50	Female	3-4	No	No	30,001-50,000 SGD	30-50%		Japanese
4	2019/03/03 8:11:14 PM GMT+8	50 and above	Male	1-2	Yes	No	50,001-80,000 SGD	30-50%	low (200-400 SGD)	Japanese
5	2019/03/03 8:37:22 PM GMT+8	25-40	Male	3-4	No	No	Will not like to reveal	Will not like to reveal		Others
6	2019/03/03 8:38:46 PM GMT+8	40-50	Male	3-4	No	Yes	30,001-50,000 SGD	30-50%	medium (400-600 SGD)	Japanese
7	2019/03/03 8:39:32 PM GMT+8	25-40	Male	3-4	No	No	30,001-50,000 SGD	0-30%	medium (400-600 SGD)	Japanese
8	2019/03/03 8:41:09 PM GMT+8	40-50	Male	3-4	No	Yes	30,001-50,000 SGD	30-50%	medium (400-600 SGD)	Japanese
9	2019/03/03 8:43:39 PM GMT+8	40-50	Male	5-6	No	No	Will not like to reveal	Will not like to reveal	low (200-400 SGD)	Others
10	2019/03/03 8:45:46 PM GMT+8	40-50	Male	3-4	No	Yes	80,001 - 100,000 SGD	0-30%	medium (400-600 SGD)	Japanese
11	2019/03/03 8:46:56 PM GMT+8	25-40	Male	3-4	No	No	20,000 - 30,000 SGD	50-70%	medium (400-600 SGD)	Korean
12	2019/03/03 8:48:40 PM GMT+8	40-50	Male	3-4	No	No	20,000 - 30,000 SGD	0-30%	medium (400-600 SGD)	Japanese
13	2019/03/03 8:49:33 PM GMT+8	25-40	Male	3-4	No	Yes	50,001-80,000 SGD	30-50%	high (more then 400 SGD)	Japanese
14	2019/03/03 8:53:22 PM GMT+8	50 and above	Male	6+	No	Yes	20,000 - 30,000 SGD	50-70%	high (more then 400 SGD)	Japanese
15	2019/03/03 8:55:04 PM GMT+8	25-40	Male	3-4	No	Yes	30,001-50,000 SGD	30-50%	medium (400-600 SGD)	Japanese
16	2019/03/03 8:56:35 PM GMT+8	40-50	Male	5-6	No	Yes	20,000 - 30,000 SGD	50-70%	medium (400-600 SGD)	Japanese

Survey cleaned data:

Excel File Edit View Insert Format Tools Data Window Help																
AutoSave ON																
Home Insert Draw Page Layout Formulas Data Review View Tell me what you want to do																
AV8																
	A	B	C	E	F	G	H	I	J	M	N	O	P	Q	R	S
1	#	Age	Gender	1-2	3-4	5-6	6+	Retiree?	Own car	Cash Down (Estimate)	Loan % (Estimate)	Car value (Estimate)	Monthly Car Ex	Country	Type	
2	1	40-50	1	0	0	1	0	0	0	40000	60%	\$ 100,000	L	Japanese	MPV	:
3	2	40-50	0	0	1	0	0	0	0	40000	40%	\$ 66,667	M	Japanese	SUV	:
4	3	50 and above	1	1	0	0	0	1	0	65000	40%	\$ 108,333	L	Japanese	Hatchback	:
5	4	25-40	1	0	1	0	0	0	0	65000	30%	\$ 92,857	L	Others	Others	:
6	5	40-50	1	0	1	0	0	0	1	40000	40%	\$ 66,667	M	Japanese	SUV	:
7	6	25-40	1	0	1	0	0	0	0	40000	30%	\$ 57,143	M	Japanese	SUV	:
8	7	40-50	1	0	1	0	0	0	1	40000	40%	\$ 66,667	M	Japanese	Sedan	:
9	8	40-50	1	0	0	1	0	0	0	65000	30%	\$ 92,857	L	Others	SUV	:
10	9	40-50	1	0	1	0	0	0	1	90000	30%	\$ 128,571	M	Japanese	SUV	:
11	10	25-40	1	0	1	0	0	0	0	25000	60%	\$ 62,500	M	Korean	SUV	:
12	11	40-50	1	0	1	0	0	0	0	25000	30%	\$ 35,714	M	Japanese	SUV	:
13	12	25-40	1	0	1	0	0	0	1	65000	40%	\$ 108,333	H	Japanese	SUV	:
14	13	50 and above	1	0	0	0	1	0	1	25000	60%	\$ 62,500	H	Japanese	Sedan	:
15	14	25-40	1	0	1	0	0	0	1	40000	40%	\$ 66,667	M	Japanese	Hatchback	:

Details attached in project zip folder

Survey Google analytics report

Details attached in project zip folder

## Appendix B: Car Recommendation System User Guide

User is guide to a dialogue box upon open the application on mobile or through a public URL. As discussed previously, the MVP is planned for KIE only. Functionality has to be exposed through a web API.

The input screen will appear like below:

Car Recommendation System  
User Input Box

Swipe to learn more

Age	Weekly Usage	Loan Min %
Gender	Brand Consciousness	Loan Max %
Number of family members	Monthly car exp	Cash Down Max
First or second car?		Cash Down Min

Swipe over to input

Submit

At the input screen, please input all the input fields by swiping on the forward arrow mark. The inputs are either true/false or based on some numeric range, like highlighted below!



## Car Recommendation System User Input Box



Swipe to learn more

Age	number	Weekly Usage	High/Low	Loan Min %	Number in percent
Gender	M/F	Brand Consciousness	Yes/No	Loan Max %	Number in number
Number of family members	number	Monthly car exp	High/Low	Cash Down Max	Number in SG \$
First or second car?	F/S			Cash Down Min	Number in SG \$



Swipe over to input

Submit

The output, in today's MVP, will appear like this.

## Car Recommendation System

Recommendations for you based on your inputs & our AI engine



Swipe to learn more

We recommend you look for car in this budget range

Car Budget (max)

Car Budget (min)

We recommend that you will be satisfied with this make

Car Make

We recommend that you will be satisfied with this type

Car Type

We recommend following features in the car

Car feature 1

Car feature 2

Car feature 3

Coming soon – top three car models that will fit your needs best !

Feedback please

Point towards the information icon will provide additional details about a recommendation item.

Car Recommendation System  
Recommendations for you based on your inputs & our AI engine



Swipe to learn more

We recommend you look for car in this budget range

Max SGD budget

Car Budget (max)

i

Min SGD budget

Car Budget (min)

i

We recommend that you will be satisfied with this make

Car make –  
e.g. Japanese

Car Make

i

We recommend that you will be satisfied with this type

Car Type –  
e.g. SUV

Car Type

i

We recommend following features in the car

Car feature  
e.g. security  
& safety

Car feature 1

i

Car feature 2

i

Car feature 3

i

Coming soon – top three car models that will fit your needs best !

Feedback please

## Appendix C: Sample Scenario

Scenario: Value buyer, aged 30, male

A car buyer with following requirements captured in the user input form, ref appendix B for the user's guide, for user input dialogue box in mobile app/browser.

Input	Description	Data type	User inputs
Age	Age of buyer	numeric	30
Gender	Gender of buyer	Binary	Male
Num_People_family	Number of family members in the house	Numeric	2
is_second_car	Is the car being bought second car	Binary	No
Cash_min	Min cash user thinks has for car buy	Numeric	25000
Cash_max	Max cash user think has for car buy	Numeric	30000
Loan_min	Min loan user think shall need for car buy percent	percent	50
Loan_max	Max loan user think shall need for car buy percent	percent	80
Weekly_Usage	Weekly car usage	Binary	1
Brand_Consciousness	Is brand consciousness a factor in decision	Binary	1
monthly_car_expenses	Monthly expense include fuel, maintenance	Binary	0

Results in decision process calculation, ref Table 7, flowchart for the rule definition engine to calculate CF

Input	Description	Data type	User inputs	Basis CF calculations
Age	Age of buyer	numeric	30	
Gender	Gender of buyer	Binary	Male	CF_SS == 1
Num_People_family	Number of family members in the house	Numeric	2	CF_Type_SUV/CF_Type_MPV == -1
is_second_car	Is the car being bought second car	Binary	No	not relevant for CF influence in this case
Cash_min	Min cash user thinks has for car buy	Numeric	25000	\$83,333
Cash_max	Max cash user think has for car buy	Numeric	30000	\$100,000
Loan_min	Min loan user think shall need for car buy percent	percent	50	CF_30_loan == -0.5
Loan_max	Max loan user think shall need for car buy percent	percent	80	CF_70_loan == 1
Weekly_Usage	Weekly car usage	Binary	1	CF_features_economy == 1
Brand_Consciousness	Is brand consciousness a factor in decision	Binary	1	CF_European == 1, CF_Jap == 0.8
monthly_car_expenses	Monthly expense include fuel, maintenance	Binary	0	CF_European (updated) == -1, CF_Jap == 0.8

The car recommendation system, as defined in Appendix B, user guide will make following recommendations. With these recommendations, the budget is narrowed down, the car make, type and features narrowed down as well.

Output	Output bracket	Possible values	Recommendations
Price	Price Min	Minimum car price	\$83,333
	Price Max	Maximum car price	\$100,000
Car country make	Country of car make	Japanese	Japanese
		European	No
		Korean	Does not matter
		American	No
		Malaysian	No
Car Type	Type of car	Sedan	Yes
		SUV	No
		MPV	No
		Hatchback	Yes
		Stationwagon	No
Car Features	Recommended car features	Safety and security	Yes
		Fuel Economy	Yes
		Engine	Does not matter

Once we link the app to current market models and pricing, we can zero down to best 2-3 options available (future enhancement captured in Section Future Project Enhancements)

## Appendix D: Glossary

MAS	: Monetary Authority of Singapore
LTA	: Land Transport Authority of Singapore
COE	: certificate of entitlement. Granted by LTA Singapore. Each car must carry the COE. COE is a major component of car price.
TCO	: Total cost of ownership
EMI	: Equal Monthly instalment
CF	: Certainty factor
BHP	: Brake Horse Power
MVP	: Minimum Viable Product
KIE	: Knowledge is everything
DDoS	: Distributed Denial of Service
C&C	: Cycle and Carriage
NBA	: next best alternatives
AWS	: Amazon Web Services
IaaS	: Infrastructure as a Service
PaaS	: Platform as a Service
SUV	: Sports Utility Vehicle
MPV	: Multi Purpose Vehicle