



Bon Voyage! **The Stress-Free,** **Automated Message** **Explainer (SAME)**

Let the robots do the *same ol'* work
so you don't have to!

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what's the difference?

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There are so many messages in my inbox!

The SIA¹ Customer Service Centre receives thousands of feedback, compliments and complaints every week



Help me sort out the messages!



**KrisFlyer &
PPS Club (KF)**



**Lounge, Catering,
Amenity (LCA) kits**

Top 2 most-asked topics



Others

Objectives: SIA management hired us to

- (1) Develop a predictive model to **automatically sort messages into the 3 topics**, and
- (2) Highlight the **frequently mentioned words and their sentiments** in KF and LCA



SAME: **Always on the move**

SAME is a natural language processing model: it can do the same enquiry sorting as we can, but around the clock!

The Approach



Data extraction

1 **KrisFlyer & PPS Club**
Earning miles, redeeming miles, status in SQ's frequent flyer programs

2 **SQ-Operated, Partner and Contract Lounges**
Information about airport lounges you can access when flying SQ.

3 **SQ Catering and Amenities**
All you have to know about Menus, Amenity kits and so on when you're onboard SQ.

Small Talk
Talk about stuff that don't fit anywhere else here

16k (40%), 10k (24%), 14k (36%) data rows respectively



Modelling

Try different **pre-processing, vectoriser, models** & pick the best combination



Sentiment Analysis

Find the **top comments of KF and LCA**, and **analyse the sentiments** of these comments to derive insights

Strap in! Diving into the data

The screenshot shows a forum thread on SQTalk. The thread title is "Sticky: Qualifying as EG for the first time" and it was started by Vtac82 on 14 May 2008. The thread contains three posts. The first post is by Vtac82, the second by SuperJonJon, and the third by SuperJonJon. The thread is titled "Sticky: Qualifying as EG for the first time" and it was started by Vtac82 on 14 May 2008. The thread contains three posts. The first post is by Vtac82, the second by SuperJonJon, and the third by SuperJonJon.

Forum Thread Details:

- Thread Title:** Sticky: Qualifying as EG for the first time
- Started by:** Vtac82, 14 May 2008, 14:32 PM
- Topics:** LATEST ACTIVITY, MY SUBSCRIPTIONS, PHOTOS

Post 1 (Vtac82):

14 May 2008, 05:43 PM #4

Originally posted by Nick C:
There's no need to panic. Been in a similar situation before. What you see on KF is based on what you need to accumulate to renew your KF membership thru April 2009. The system tracks your mileage accurately, you will be upgraded to KF-G after your upcoming trip.

Yeah? What was concerning was that even the lady on the phone told me I had to collect further 48,064 to get KF G.

Anyway, if that's the case, does that mean by transit in Singapore I can get Gold Lounge access?

From experience, flying with SQ, miles are credited immediately upon departure, that means that by the time I transit in Singapore, I'll be Gold, wonder if I can gain access to Gold lounge.

Either way, I'll keep you guys posted on Sunday.

Post 2 (SuperJonJon):

14 May 2008, 05:44 PM #5

Firstly, Welcome to SQTalk, Vtac82!

Krisflyer miles and Elite miles are actually two separate things. Elite miles are earned only thru flying but Krisflyer miles can be earned through credit cards, hotel stays, etc.

I am guessing that you have flown about 26,936 miles or slightly more. When you hit 25K or slightly above, you achieved KF Silver in May thus the counter has reset. You have flown 1,936 miles since achieving Silver and would require another 23,064 miles to get Gold.

I hope that I am wrong and you get your KF Gold this Sunday.

Post 3 (SuperJonJon):

14 May 2008, 05:46 PM #6

Originally posted by SuperJonJon:
Firstly, Welcome to SQTalk, Vtac82!

Krisflyer miles and Elite miles are actually two separate things. Elite miles are earned only thru flying but Krisflyer miles can be earned through credit cards, hotel stays, etc.

I am guessing that you have flown about 26,936 miles or slightly more. When you hit 25K or slightly above, you achieved KF Silver in May thus the counter has reset. You have flown 1,936 miles since achieving Silver and would require another 23,064 miles to get Gold.

I hope that I am wrong and you get your KF Gold this Sunday.

Tier bonuses also does not count towards Elite qualification / requalifications.

1. Properly spelled words

2. Absence of Singlish

3. Replies to comments will duplicate words

4. Thread titles are important as they contain many key words

What's the Model Answer?

1

Lemmatisation vs Stemming vs None:

2

Pre-processing & vectorisation trials:

Bolded – different from baseline

Steps	Baseline	Alternate 1	Alternate 2	Alternate 3
Pre-processing	<ul style="list-style-type: none">Basic data cleaning	<ul style="list-style-type: none">Basic data Cleaning	<ul style="list-style-type: none">Basic data CleaningRemove duplicated sentences	<ul style="list-style-type: none">Basic data CleaningRemove duplicated sentences
Vectorisation	CountVectoriser	TFIDF	CountVectoriser	TFIDF
Model	Multi-Naïve Bayes	Multi-Naïve Bayes	Multi-Naïve Bayes	Multi-Naïve Bayes

Best performing pre-processing & vectorisation combination

3

Modelling trials:

Model	Multi-Naïve Bayes	Random Forest	XGBoost	SVM
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* The alternate 1, 2, 3 also had 4 additional stopwords (iirc, imo, imho, btw) extracted from a "SQTalk Abbreviations, Slangs, Definitions, Phrases" thread. This had little effect on the model performance.

1 Lemm vs Stem vs None

	Baseline 1	Baseline 2	Baseline 3
Pre-processing	Basic cleaning	- basic cleaning - lemm	- basic cleaning - stem
Vectoriser	CountVectoriser	CountVectoriser	CountVectoriser
Model	Multinomial Naive Bayes	Multinomial Naive Bayes	Multinomial Naive Bayes
macro-average ROC AUC	0.878	0.885	0.887
macro-average f1-score	0.737	0.747	0.752

Remarks:

- **Baseline 3 (remove NA, with stem) performed the best**
- **Macro-average ROC AUC (One vs the Rest) and f1-score (labels = KF, LCA) used as key metrics**
 - To compare the model's confidence to distinguish classifications, and focus on minimising false-positives and -negatives for KF and LCA

2 Preprocess, Vectoriser

	Baseline Model (from notebook 2)	Alternate 1 *Best performance*	Alternate 2	Alternate 3
Pre-processing	- Basic cleaning - Stem	- Basic cleaning - Stem	- Basic cleaning - Stem - Remove duplicated sentences	- Basic cleaning - Stem - Remove duplicated sentences
Vectoriser	CountVectoriser	TFIDF	CountVectoriser	TFIDF
Model	Multinomial Naive Bayes	Multinomial Naive Bayes	Multinomial Naive Bayes	Multinomial Naive Bayes
Macro-average ROC AUC	0.887	0.908	0.807	0.819
Macro-average f1-score (kf, lca)	0.752	0.759	0.625	0.630

Best performance: Basic clean, stem, TFIDF

Remarks:

- TFIDF had slightly better performance than CountVectoriser
- **Removing duplicated comments seemed to have an adverse effect** on model performance; this suggests that the comments that people reply to usually have crucial key words in them

Model Comparison

	Model 1 (from notebook 3)	Model 2	Model 3	Model 4 *Best performance*
Vectoriser	TFIDF	TFIDF	TFIDF	TFIDF
Model	Multinomial Naive Bayes	Random Forest	XGBoost	SVM
Macro-average ROC AUC	0.908	0.883	0.925	0.925
Macro-average f1-score (kf, lca)	0.759	0.719	0.807	0.815

Best performance: SVM

Remarks:

- **Model 3 performed the best** with the following params:
 - TfidfVectorizer(max_features=500, stop_words=stem_stopwords)
 - SVC(C=1, gamma=1, probability=True, random_state=42)
- SVM edged out due to its **slightly better f1-score**

Sentiment Analysis: Top Words

KrisFlyer & PPS Club

	count
miles	8899
SQ	8627
posted	7734
Originally	7367
PPS	5771
SIN	4730
flight	3692
KF	3616
just	3547
year	2822

Lounges, Catering, Amenity Kits (LCA)

	count
SIN	7319
lounge	4650
SQ	4337
posted	3819
Originally	3637
The	2213
served	2031
chicken	1928
flight	1872
vegetables	1805



1. Find top words
2. Find sentences that contain these words
3. Sort the sentiment score by:

Sentiment	Score	Assigned value
Very Negative	Under -0.5	1
Negative	Between -0.5 and -0.1	2
Neutral	Between -0.1 and 0.1	3
Positive	Between 0.1 and 0.5	4
Very Positive	Over 0.5	5

Sentiment Analysis: Results

Sentiment Score	KrisFlyer 'miles'	KrisFlyer 'PPS'	KrisFlyer 'KF'	LCA 'lounge'	LCA 'SQ'	LCA 'served'
Very negative (1)	<0.1	<0.1	<0.1	<0.1	<0.1	<0.1
Negative (2)	<0.1	<0.1	<0.1	<0.1	0.1	0.1
Neutral (3)	0.3	0.4	0.4	0.4	0.4	0.6
Positive (4)	0.6	0.6	0.6	0.5	0.4	0.3
Very positive (5)	<0.1	<0.1	<0.1	<0.1	<0.1	<0.1
Overall:	Positive	Positive	Positive	Positive	Positive	Positive

* Due to rounding, total may not add up to 1

Remarks:

- Sentiments for comments containing top words are **positive** → **these are areas of strength for SIA**
- Separate deep-dive into these comments can be conducted to find out the reasons why (e.g. good service, good exclusive deals for members, good food served etc.)

Conclusion



Model

Successfully developed SAME: a prediction model with macro-average ROC AUC (0.93) and f1-scores (0.82)



Sentiment Analysis

Comments with 'miles', 'PPS', 'KF', 'lounge', 'SQ', 'served' have **largely positive sentiments** → **areas of strength for SIA**



Next Steps (areas for future improvement)




- **Sort by topic (e.g. KF, LCA, others):** SAME can be used as a backend engine for a chatbot to sort incoming messages
- **Sort by type (feedback, complaints, compliments):** SAME + Sentiment Analysis
- **Incorporate Singlish** into sentiment analysis to improve performance
- Find **top words with negative sentiments** to identify and improve on areas of weaknesses

Thanks!

**Your work will never
be the *same* again**



Annex A - Macro Average f1-Score

Label	Per-Class F1 Score	Macro-Averaged F1 Score
 Airplane	0.67	$\frac{0.67 + 0.40 + 0.67}{3}$ = 0.58
 Boat	0.40	
 Car	0.67	

The macro-averaged F1 score (or macro F1 score) is computed using the arithmetic mean (aka unweighted mean) of all the per-class F1 scores.

In general, if you are working with an imbalanced dataset where all classes are equally important, using the macro average would be a good choice as it treats all classes equally. ([source](#))