



DIAMONDS

Service Like No Other





Showing insightful trends,



Emitting the beauty of data



How to use data to
enhance the
efficiency of
customer service?

Key Features



Auto-generated
Answers



SMART Priority
Filtering



Interactive
Dashboard



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7%

Of Facebook posts can be answered by SingTel's FAQs

Avg ~1000

Facebook posts per month

Avg ~70

Facebook posts per month which can be covered by FAQs

Avg ~840

Facebook posts per year which can be covered by FAQs

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Priority Filters

- Category (e.g. tech support, request, complain, enquiry, compliment, others)
- Junk/Spam
- Priority Score

Priority Algorithm

Following the guidelines from ITIL (Information Technology Infrastructure Library), our tickets are prioritized based on **urgency** and **business impact**.

	Tech Support	Request	Complaint	Enquiry	Compliment	Suggestion/ Others
Negative						
Neutral						
Positive						






 Highest
 High
 Medium
 Low
 Lowest

Table: Priority Matrix



Priority Algorithm



Category		Label		Sentiment	
Tech Support	4	Broadband	2	Negative	1.5
Request	3	Lifestyle Services	1.3	Neutral	1.2
Complaint	3	Mobile	2	Positive	1
Enquiry	2	Others	1		
Compliment	1	Telephone Services	1.5		
Suggestion/Others	1				



Priority Algorithm

$$\text{Score} = \text{Category} + \text{Sentiment} * \text{Label} \\ + \text{Num_of_Days_Since_Posted} * 0.1$$



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Key Parties



Entry Level



Management Level



The Solution



DATA



KNOWLEDGE



ACTION



Entry Level Dashboard DEMO

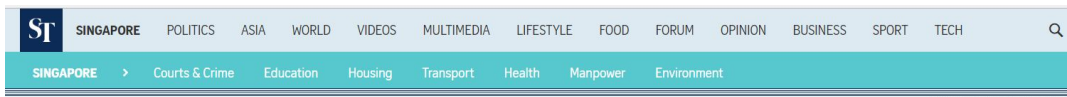
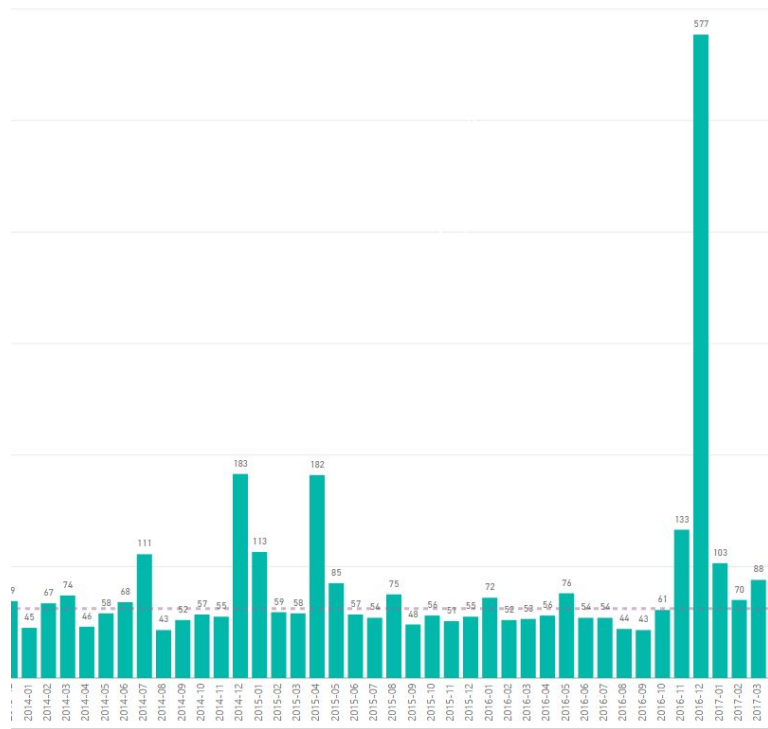


Management Level Dashboard DEMO



Insights from Data Analysis

- 1) 2 main spikes in #posts - Apr 2015 and Dec 2016
- 2) More than 500% increase in number of posts compared to average
 - a) Significantly higher number of *broadband* posts

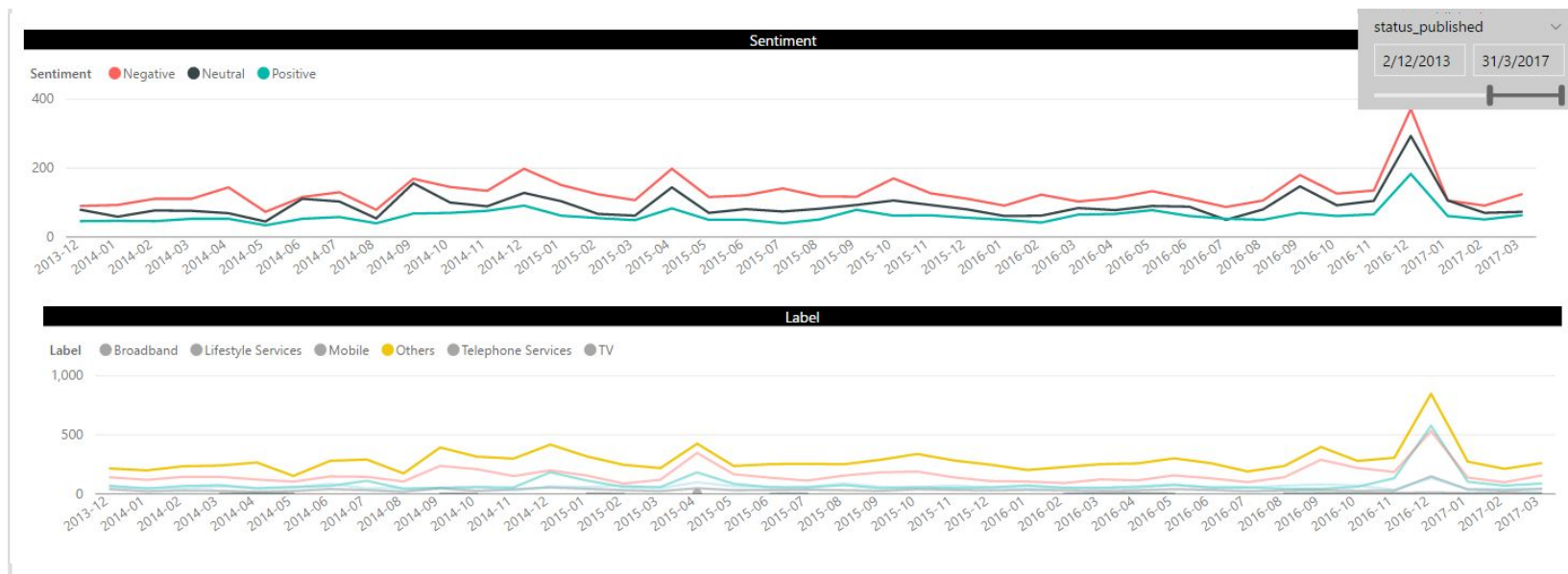


Singtel disruption: Fibre broadband services fully restored,
10% discount on broadband subscription for December



Insights from Data Analysis

3) Posts labelled under *Others* are mostly *negative* sentiments & are under *complaints*



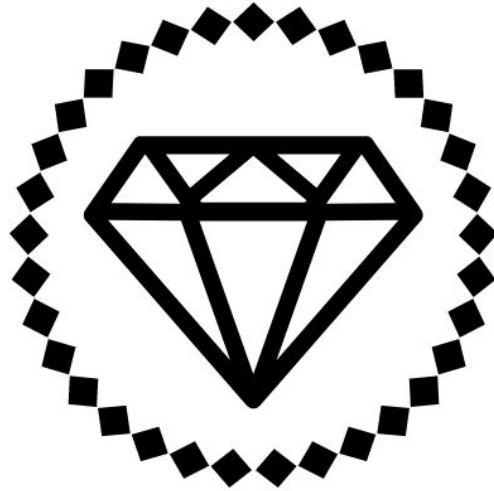
Key Learning Points

Our main challenges and takeaways

- Working with unclassified data
- Accuracy of model's categorization



DIAMONDS



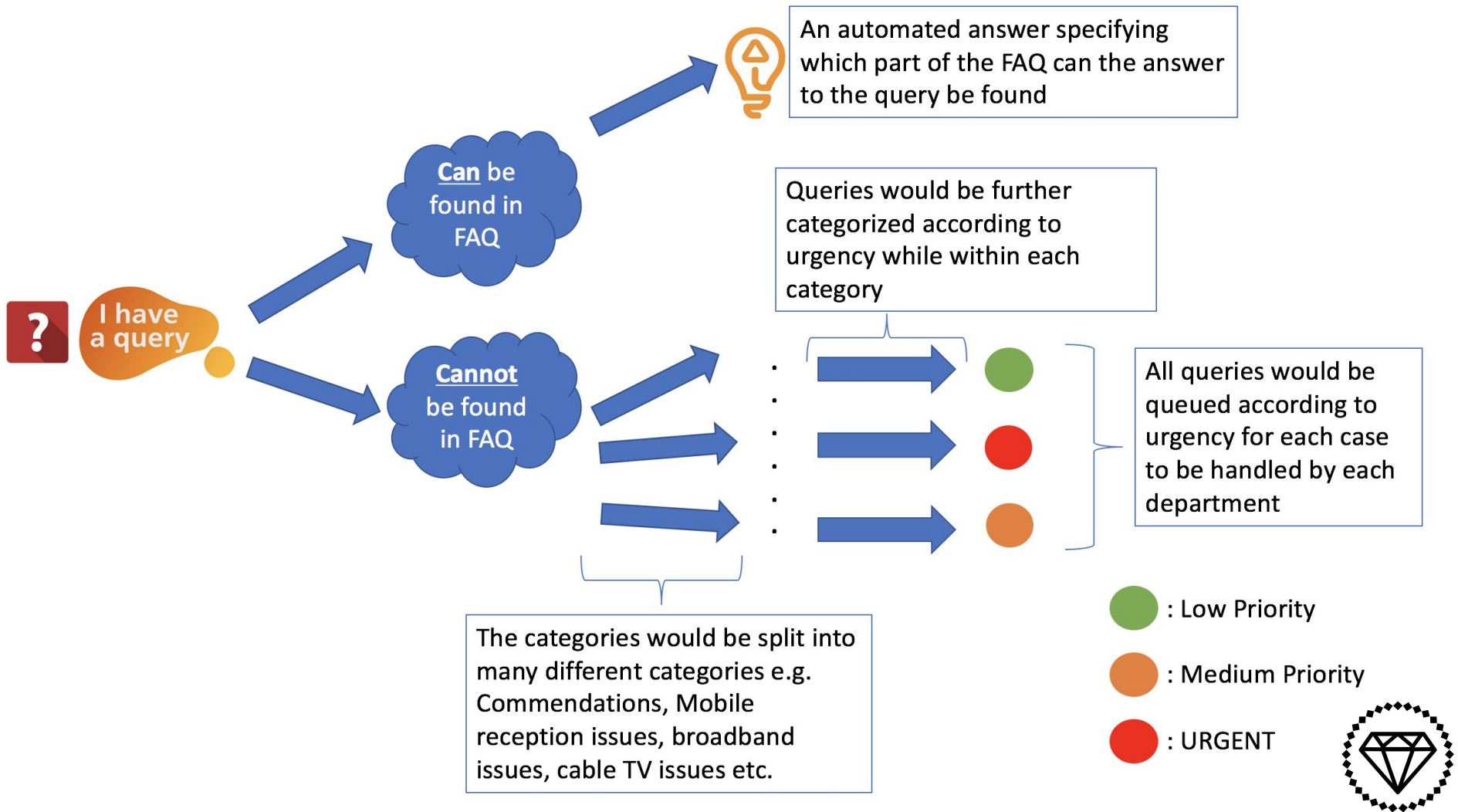
Service Like No Other

Thank You :)



Any Questions?





40% Training 60% Test

RF: (Mean 0.515)

```
> RF <- train_models(container, algorithms= c("RF"))
> RF_CLASSIFY = classify_models(container, RF)
>
> SVM_CV <- cross_validate(container, 10, "RF")
Fold 1 Out of Sample Accuracy = 0.5213549
Fold 2 Out of Sample Accuracy = 0.5109923
Fold 3 Out of Sample Accuracy = 0.4945881
Fold 4 Out of Sample Accuracy = 0.5150421
Fold 5 Out of Sample Accuracy = 0.5183541
Fold 6 Out of Sample Accuracy = 0.5322581
Fold 7 Out of Sample Accuracy = 0.5231748
Fold 8 Out of Sample Accuracy = 0.5088562
Fold 9 Out of Sample Accuracy = 0.5110783
Fold 10 Out of Sample Accuracy = 0.5188962
> |
```

BAGGING (Mean 0.497):

```
> SVM_CV <- cross_validate(container, 10, "BAGGING")
Fold 1 Out of Sample Accuracy = 0.4945881
Fold 2 Out of Sample Accuracy = 0.4883926
Fold 3 Out of Sample Accuracy = 0.5034216
Fold 4 Out of Sample Accuracy = 0.488081
Fold 5 Out of Sample Accuracy = 0.4927105
Fold 6 Out of Sample Accuracy = 0.5001516
Fold 7 Out of Sample Accuracy = 0.5111715
Fold 8 Out of Sample Accuracy = 0.5043452
Fold 9 Out of Sample Accuracy = 0.4988138
Fold 10 Out of Sample Accuracy = 0.4894377
>
```



50% Training 50% Test

RF: (Mean 0.514)

```
> RF <- train_models(container, algorithm="RF")
> RF_CLASSIFY = classify_models(container, RF)
>
> SVM_CV <- cross_validate(container, 10, "RF")
Fold 1 Out of Sample Accuracy = 0.5245025
Fold 2 Out of Sample Accuracy = 0.5051668
Fold 3 Out of Sample Accuracy = 0.5224903
Fold 4 Out of Sample Accuracy = 0.5373633
Fold 5 Out of Sample Accuracy = 0.5046062
Fold 6 Out of Sample Accuracy = 0.515433
Fold 7 Out of Sample Accuracy = 0.5105804
Fold 8 Out of Sample Accuracy = 0.5202723
Fold 9 Out of Sample Accuracy = 0.5082596
Fold 10 Out of Sample Accuracy = 0.5127042
>
> analytics = create_analytics(container, RF_CLASSIFY)
|
```

BAGGING: (Mean 0.498)

```
> RF <- train_models(container, algorithm="RF")
> RF_CLASSIFY = classify_models(container, RF)
>
> SVM_CV <- cross_validate(container, 10, "BAGGING")
Fold 1 Out of Sample Accuracy = 0.5114345
Fold 2 Out of Sample Accuracy = 0.490995
Fold 3 Out of Sample Accuracy = 0.4968722
Fold 4 Out of Sample Accuracy = 0.5224787
Fold 5 Out of Sample Accuracy = 0.4882615
Fold 6 Out of Sample Accuracy = 0.4986515
Fold 7 Out of Sample Accuracy = 0.4933495
Fold 8 Out of Sample Accuracy = 0.4980764
Fold 9 Out of Sample Accuracy = 0.4926254
Fold 10 Out of Sample Accuracy = 0.4972777
>
> analytics = create_analytics(container, RF_CLASSIFY)
```



60% Training 40% Test

RF: (Mean 0.515)

```
>  
> RF <- train_models(container, algorithms= c("RF"))  
> SVM_CV <- cross_validate(container, 10, "RF")  
Fold 1 Out of Sample Accuracy = 0.5038497  
Fold 2 Out of Sample Accuracy = 0.5195698  
Fold 3 Out of Sample Accuracy = 0.5070796  
Fold 4 Out of Sample Accuracy = 0.5142258  
Fold 5 Out of Sample Accuracy = 0.515343  
Fold 6 Out of Sample Accuracy = 0.5237812  
Fold 7 Out of Sample Accuracy = 0.5323952  
Fold 8 Out of Sample Accuracy = 0.4980888  
Fold 9 Out of Sample Accuracy = 0.5262201  
Fold 10 Out of Sample Accuracy = 0.5175282  
> |
```

BAGGING: (Mean 0.496)

```
Fold 10 Out of Sample Accuracy = 0.5175282  
> SVM_CV <- cross_validate(container, 10, "BAGGING")  
Fold 1 Out of Sample Accuracy = 0.50886  
Fold 2 Out of Sample Accuracy = 0.4938053  
Fold 3 Out of Sample Accuracy = 0.4862275  
Fold 4 Out of Sample Accuracy = 0.4872972  
Fold 5 Out of Sample Accuracy = 0.5160163  
Fold 6 Out of Sample Accuracy = 0.497022  
Fold 7 Out of Sample Accuracy = 0.4854634  
Fold 8 Out of Sample Accuracy = 0.4871717  
Fold 9 Out of Sample Accuracy = 0.4992544  
Fold 10 Out of Sample Accuracy = 0.5064275  
> |
```





Google Cloud Platform



IBM **Watson**

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<https://app.powerbi.com/view?r=eyJrIjozMzQ5OGQ5YTAtZTQ5Yi00MDRhLTgwMmEtMThkNzdmMDJiYzhhIiwidCI6ImI3YzgZyWYyLWYxM2QtNDU0Yi04NjAwLTlhZDhmZWxMDE3YyIsImMiOjEwfQ%3D%3D>

<https://badiamondsdash.shinyapps.io/dashboard/>