

# **Conversational Analytics**

Sept 2022

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## **Areas of Focus**

RHSH

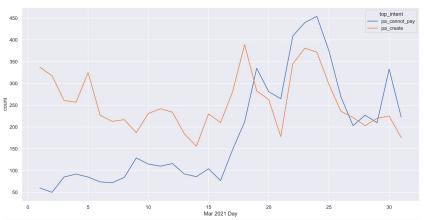
- → Basic time-series analyses: Traffic over defined periods of time
  - % volume of traffic by Intent / keywords / channel / etc.
  - Average confidences of predicted intents and actions
- → Topic Modeling
  - Overall corpus
  - By intent
- → NLU/conversation review
  - Layered keyword filtering / text clustering / word clouds
  - Model classification diffs
  - Dropoff intents/actions (where do conversations end)
- → Cross-Validation reviews (from Rasa's CLI tool)
  - Individual reports
  - Comparison reports

### **Time Series**



#### By intent traffic

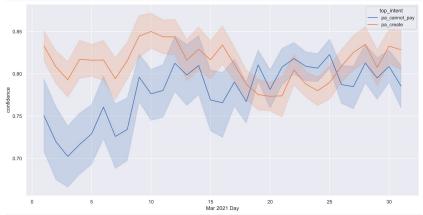
- → Prioritize by customer use
- → Gauge responses to ad campaigns
- → NLU updates properly re-directing



Ad campaign response

#### By average intent confidences

- → Review NLU updates
- → Identify intents potentially needing NLU sample improvements



Same ad campaign caused customers to use unexpected phrasing which also caused a drop in confidence levels of a related intent

### RASA

# **Topic Modeling**

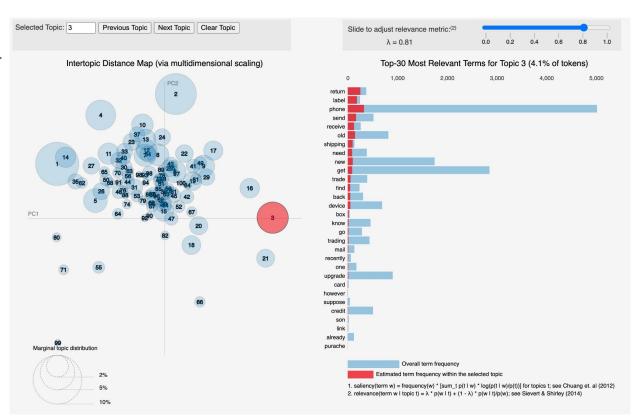
Lemmatization and other processing methods will have a major impact on this.

#### **Overall corpus**

- → Rough outline of potential services to provide
- → Identify common terminology within and between intents
- Observe shifts by news cycles, ad campaigns, etc

#### By intent traffic

- → Identify potential new story branches or intents - depending on grouping (dis)similarities
- → Discover phrasing and terminology to use in representative samples (eg. keywords)



Example topic model on the specific intent of "returns": identifies return labels as a major request

# **NLU / Conversation review**

### RHSH

#### **NLU review/Sample Gathering**

- → Keyword filtering determine % volumes of traffic with specific terminology to represent appropriately within NLU samples
  - "and"/ "or" pairings
  - Synonyms / slang
  - Word clouds/counts
- → Dropoff intents/actions
  - What was the last intent in the conversation, and was it served properly
  - Was the last action taken by the bot ignored and never responded to?
  - Broken actions/story pathing
- → Model Classification changes
  - Using utterances from conversations, classified by the current model, have the new model iterate through the utterances and re-classify them and compare the similarities/differences

text	intent	intent_prediction	confidence
It's not letting my choose October 15th	pa_change_dates	broken	0.924

Fig 9: Report Intent Errors table from "label errors.pdf"

intent	f1-score	precision	recall	samples	confused_with
pa_create	0.654	0.607	0.708	96	'pa_cannot_pay': 8 'pa_change_dates': 5

Fig 10: Report Intent Accuracy summary table from "label figs.pdf"

intent	confidence	count
line_add_byo	0.892	11
intent_prediction	confidence	count
affirm	0.902	28

Fig 13: tables for intent error confidence averages

intent	intent_prediction	count	confidence
out_of_scope	affirm	21	0.876

Summary of classifications



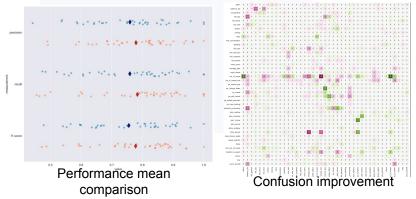
# **Cross-Validation Output Review**

#### **Individual Reports**

- NLU classification errors: review "missed" predictions for common terminology
- Confusion matrix of commonly mixed intents
- f1/precision/recall by intent correlated with number of samples and "commonly confused with..." from confusion matrix

#### **Comparison Reports**

- Performance shifts
- NLU prediction differences
- **Confusion Improvement Matrix**



intent	intent_prediction	count	confidence
out_of_scope	affirm	21	0.876
pa_edit_cancel	pa_change_dates	18	0.897

Commonly confused intents with a high average confidence during misclassification may indicate NLU samples need cleaning up

