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How we use machine learning in Project Natural Language Search

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Our Speakers



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What you will learn today

Machine learning can be used to solve many different problems.

We will show you examples of applied machine learning in Project Natural Language Search.

We will walk through high level examples before diving into one example in depth.

Introduction

What is Project Natural Language Search?

Project Natural Language Search is a natural language platform for machine data that delivers Natural Language Search, Understanding and Generation for Splunk and SQL data.



Natural Language Search



Communicate instantly
in charts, text and speech



Access anywhere
with type, touch, voice

Demo

The screenshot shows a laptop screen displaying the Splunk mobile application. The title bar reads "My Recent Queries - Splunk". The URL in the browser is https://app.askpony.com:8000/en-US/app/babel_fish/index#/searches/3b25d5697be247f6ac941b166c5439cd/recent. The top navigation bar includes links for Apps, J, D, B, C, M, O, R, X, C, LTE Spec, Beta, A, BFO, SFISH, ASKPONY, ES, AWS, F, Pike, CONF2018, and Other Bookmarks. The user profile shows Dipock Das, 63 messages, Settings, Activity, Help, and a Find search bar. Below the header, there are tabs for Search, Dashboards, Workspaces, Data Sources, and Settings. A sidebar on the left features a search bar labeled "Search in Splunk Workspace" and a "Recent Searches" section. The main content area displays recent searches grouped by date:

- May 24, 2018**
 - sales in london yesterday
May 24th 2018, 2:15 pm
- May 23, 2018**
 - show network traffic today by second between 1AM and 3AM
May 23rd 2018, 1:11 pm
 - login failures today
May 23rd 2018, 1:10 pm
- May 3, 2018**
 - rogue wireless devices on the network
May 3rd 2018, 7:41 pm
- May 2, 2018**

Overview of Machine Learning in Project Natural Language Search

- ▶ Natural Language Understanding
 - ▶ Visual Interface Determination
 - ▶ Data Driven Drill Downs
 - ▶ Handling Ambiguity
 - ▶ Natural Language Speech Recognition
 - ▶ Search Query Recommendation

Project NLS: Natural Language Understanding



What is Natural Language Understanding?

- ▶ Understand the intent behind a textual query
 - ▶ NLU is a hard AI problem!

Intent: {sum} {sales} {product} {timegrain} {city} {time range}

Project NLS turns Intents into SPL (or SQL)

Intent: {sum} {sales} {product}
{timegrain} {city} {time range}



SUM sales BY Day

WHERE Product = Cappuccino, AND
City IN Vancouver, San Francisco,
San Jose,
AND timerange is (today minus 10
days) to today



```
| tstats allow_old_summaries=t summariesonly=t
SUM("All_Sales.grossSales") AS "All_Sales.grossSales"
FROM datamodel=Retail.All_Sales
WHERE (((("All_Sales.productName"=="cappuccino")
AND ("All_Sales.city"
IN ("vancouver","san francisco","san jose")))
AND ((earliest=1533427200)
AND(latest=1534291199)))
BY "All_Sales.city" _time span=1d
| eval All_Sales__time_date=strftime(_time, "%Y-%m-%d")
| table All_Sales__time_date "All_Sales.city" "All_Sales.grossSales" |
sort 150 "All_Sales.grossSales" DESC, All_Sales__time_date DESC,
"All_Sales.city" DESC
```

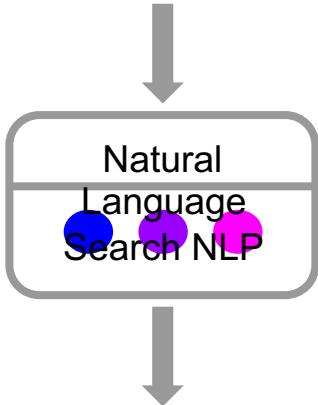
Project NLS machine learning obliterates hard coded rules systems

Show me **daily sales** for **San Francisco** in **September**

What were **daily sales** in **SF last month**

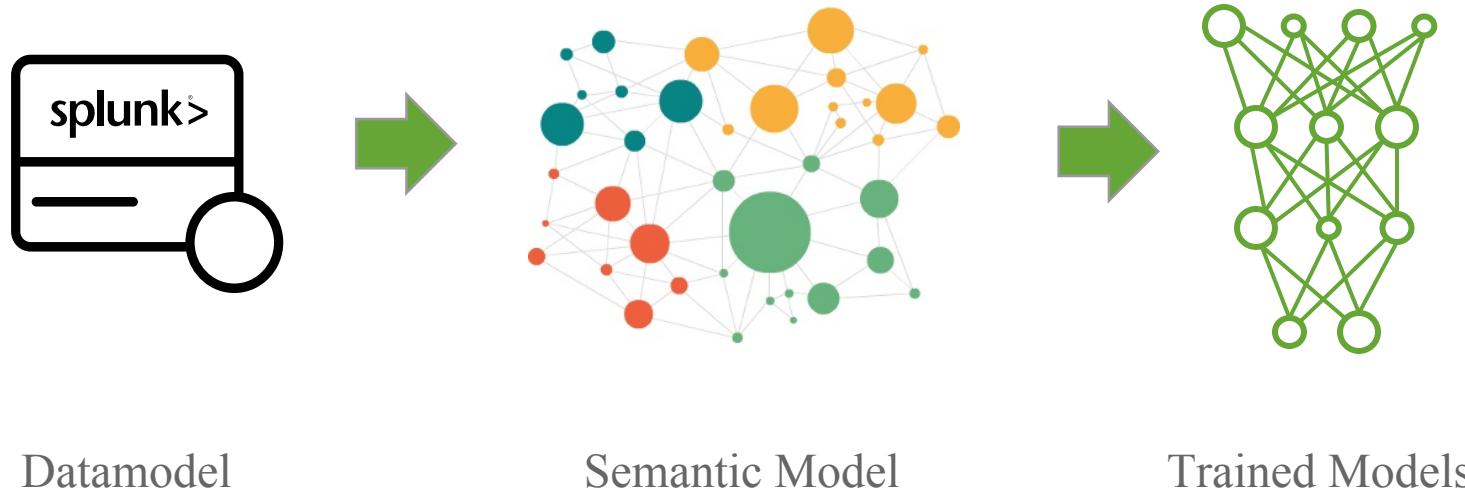
Display the **Vancouver** store **gross sales** for the **last 30 days** by **day**

Get me **San Jose takings** for the **last four weeks** by **day**

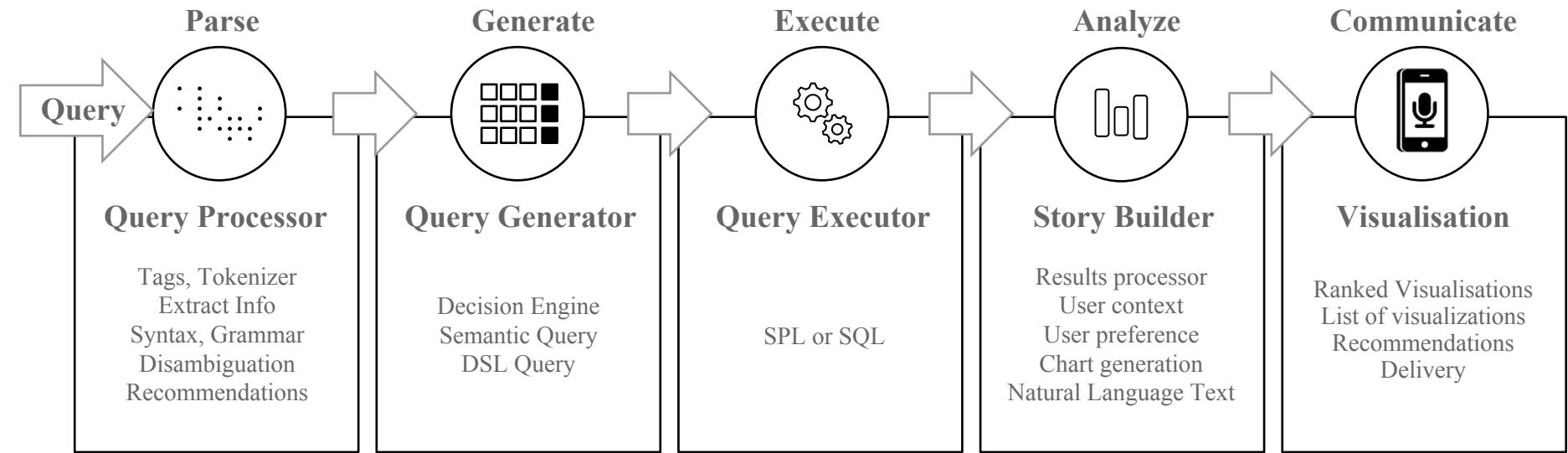


sales, city, time period, time grain

Model Driven Natural Language Understanding



Natural Language Processing Pipeline



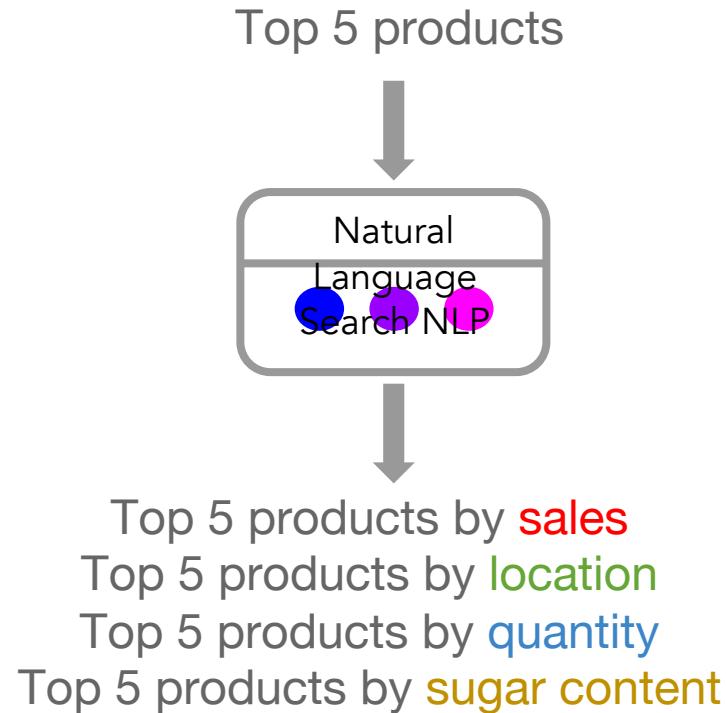
Ontology

Search Model (entities, relationships, synonyms, common names)
Intent, Statement, Responses, User Profiles, Knowledge Graph, ML Models, Templates

Project NLS: Handling Ambiguity



Natural Language search machine learning disambiguates a query



Project NLS: Visual Interface Determination



What's wrong with User Interface Design?

First Name

Last Name

Gender

Date of Birth

Month: Day: Year:

Country

Desired Email Address @

Choose a Password

Re-type Password

Contact Email (optional)

Security Question

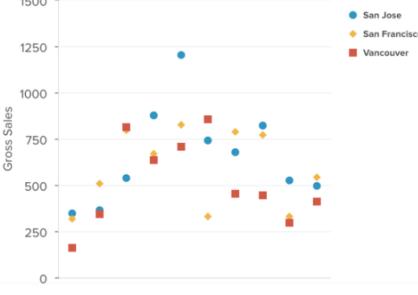
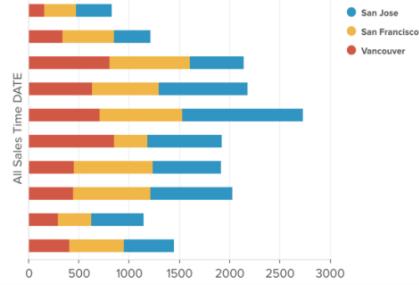
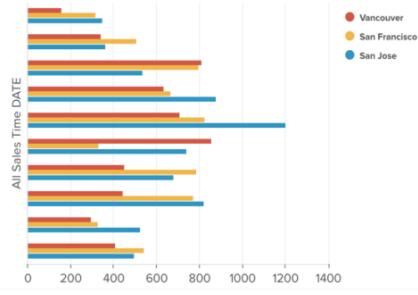
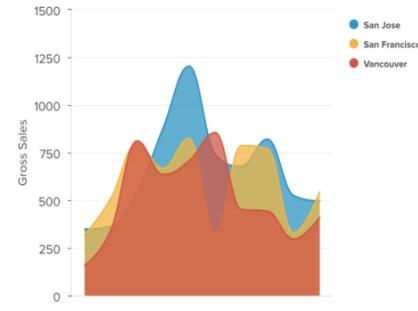
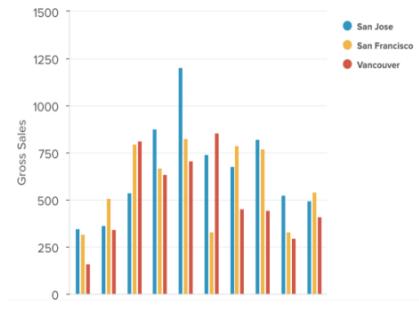
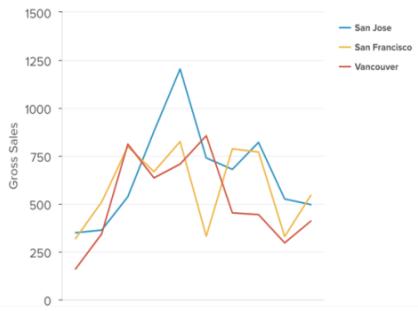
- ▶ We can do better than sort alphabetically..
- ▶ Afghanistan - 0.48% of World's population
- ▶ Default using browser getCurrentPosition()?

- Why not predict top 5 by signups?
- | | |
|----------------|-----|
| United States | 45% |
| United Kingdom | 28% |
| Canada | 16% |
| Australia | 5% |
| India | 2% |

Rest of World listed alphabetically

Using ML to display the right chart for the result

“daily sales of cappuccino in Vancouver, San Francisco and San Jose last 10 days”

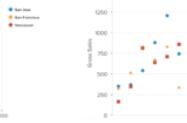
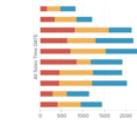
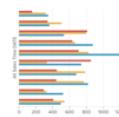
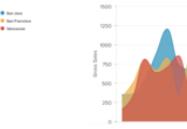
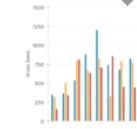
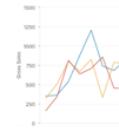
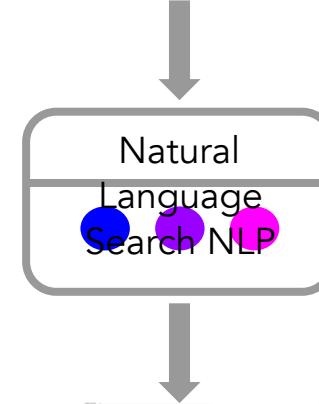


How ML Improves Visual Interface Determination

- ▶ Learn from past interactions
 - ▶ Recommend a ranked list of charts to user for a given query
 - ▶ Incorporate user's preference

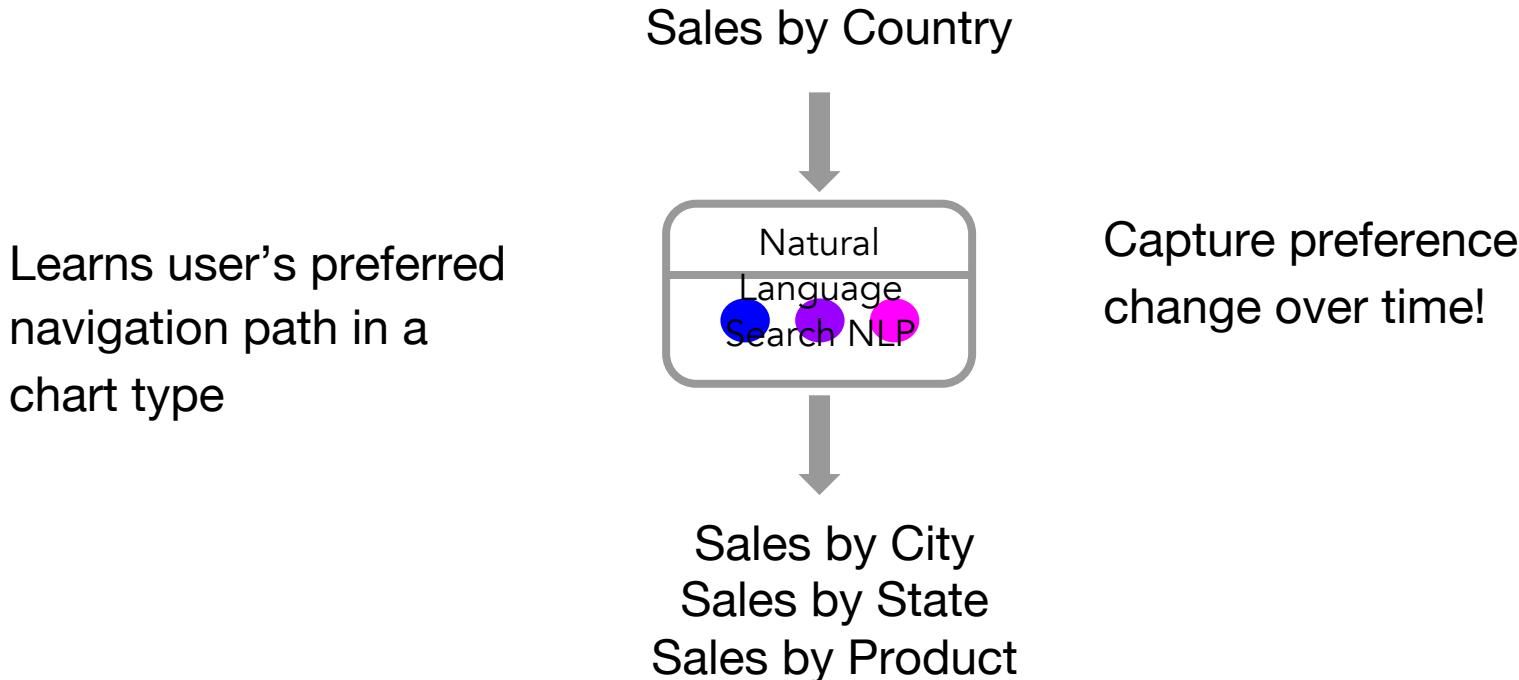
Vancouver: 234, 350, 500, 523 ...

San Francisco: 845, 820, 650, 723 .



What is Data Driven Drill Path?

Recommend drill path navigation in a selected visualization based on learned preference



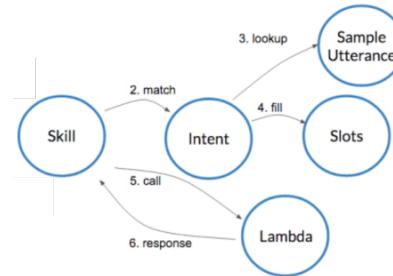
Project NLS: Natural Language Speech Recognition



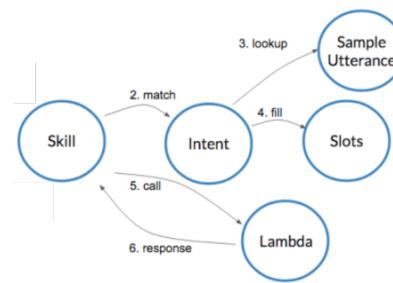
Natural Language Speech Recognition

- ▶ The User asks a question using a voice enabled device (VTT)
 - ▶ The User gets an answer from the device (TTS)

NLU reduces the burden of skill building for so many scenarios^{©20}



Trained skill



Trained skill



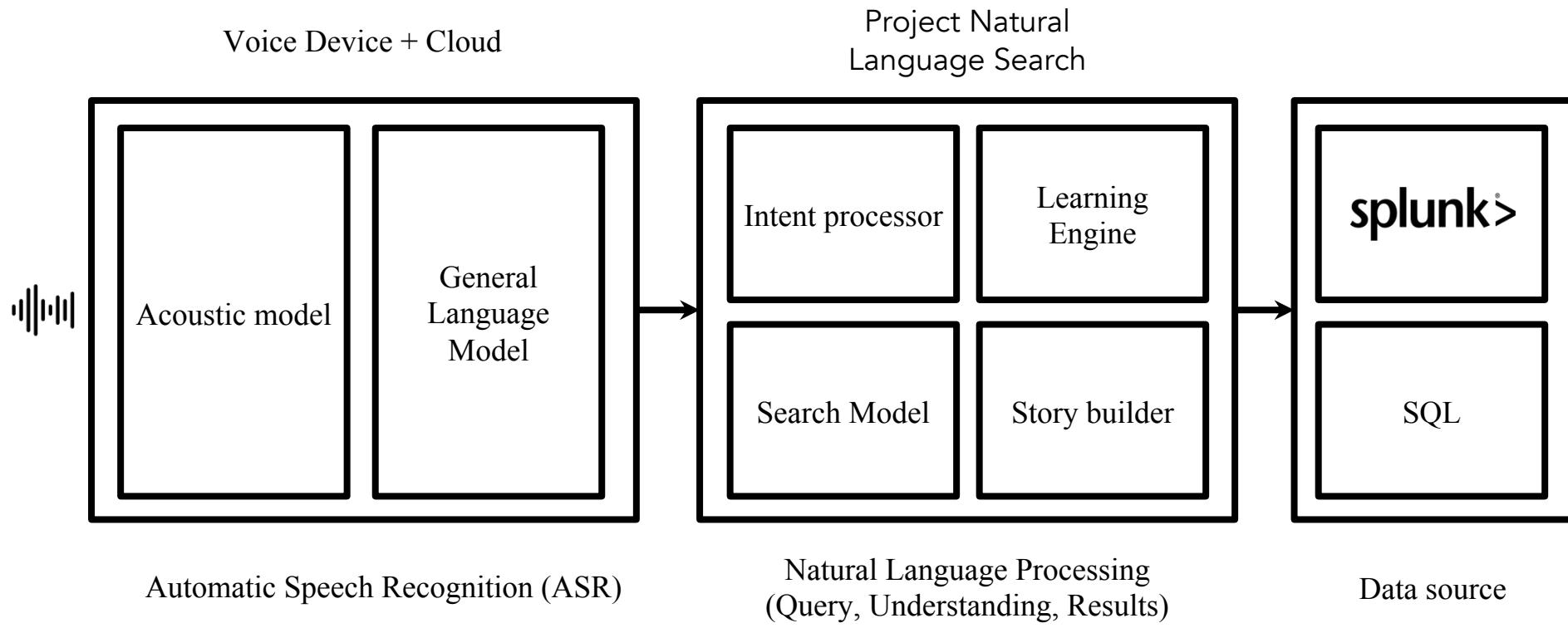
Data set



Data set

The current method for building speech (and chatbot) apps is to encode all the possible questions someone might ask.

Project NLS support for Voice enabled devices



Project NLS: Search Query Recommendation



Search Query Recommendation

Trending Search Recommendation:

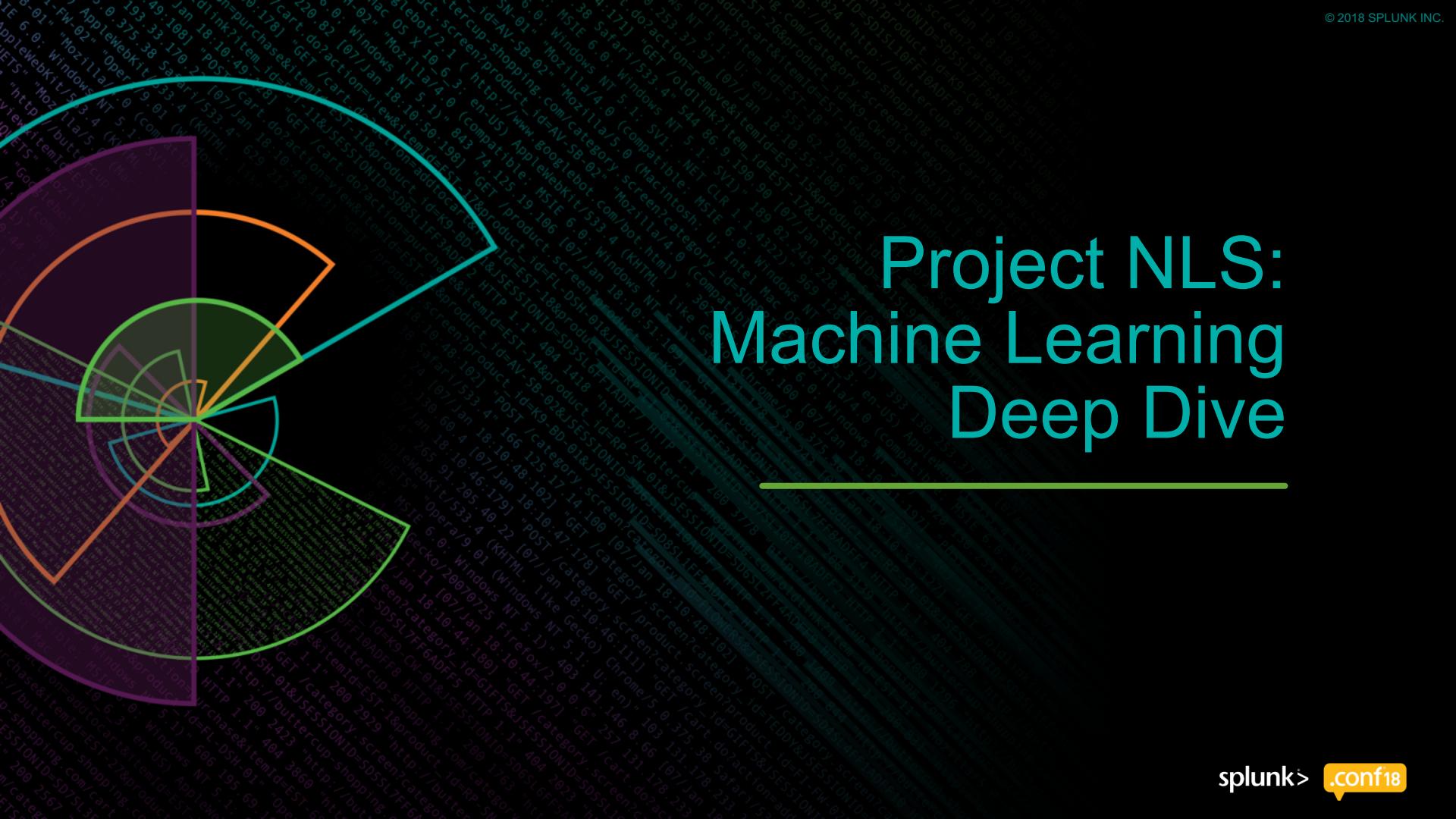
- ▶ Recommend top N search queries to user

Relevant Query Recommendation:

- ▶ Recommend top N search queries relevant to a selected search query

Demo

Project NLS: Machine Learning Deep Dive



Natural Language Search Machine Learning Architecture

Natural Language REST API

Natural Language Understanding

Natural Language Generation

Drill Path Navigation

Visual Interface Recommendation

Query Disambiguation

Trending Search Recommendation

Relevant Query Recommendation

Compute (Spark Cluster)

Model and State Persistence (HDFS)

Deep Dive: Trending Search Recommendation



Collaborative Filtering

make automatic predictions (filtering) about the interests of a user by collecting preference information from many users (collaborating)



in this case we observe the questions users ask about stores in specific city locations

Collaborative Filtering

	✓	✗	✓	✓
	✗	✓	✗	✗
	✓	✓	✗	✓
	✗	✗	✓	✓
	✓	✓	?	✗

Purple asked questions about
Seattle and **not London**

Pink asked questions about
San Francisco and **Seattle**

Green asked questions about
San Francisco, **Seattle** and
not London

Purple and Pink have asked
the same questions as Green
for matching location.

Based on the match, Green is
not likely to ask questions
about Vancouver



Trending Search Recommendation with Collaborative Filtering

 **splunk>enterprise** App: Jubilee ▾

 Search in Retail Workspace

Recommended Searches

daily sales last week in london

sales of coffee in seattle last week

hourly sales in london yesterday

daily sales in london last month

show me the trend in spicy chai sales over the last 3 months

 **splunk>enterprise** App: Jubilee ▾

 Search in Retail Workspace

Recommended Searches

hourly sales in london yesterday

sales of coffee in seattle last week

Show me the trend in spicy chai sales over the last 3 months

what were sales in the last 24 hours

show me sales quarter over quarter by product for 2015

Why was the Recommendations List updated?

Machine Learning Module Components

Trending Search Recommendation REST API

Ratings Generator

Model Builder

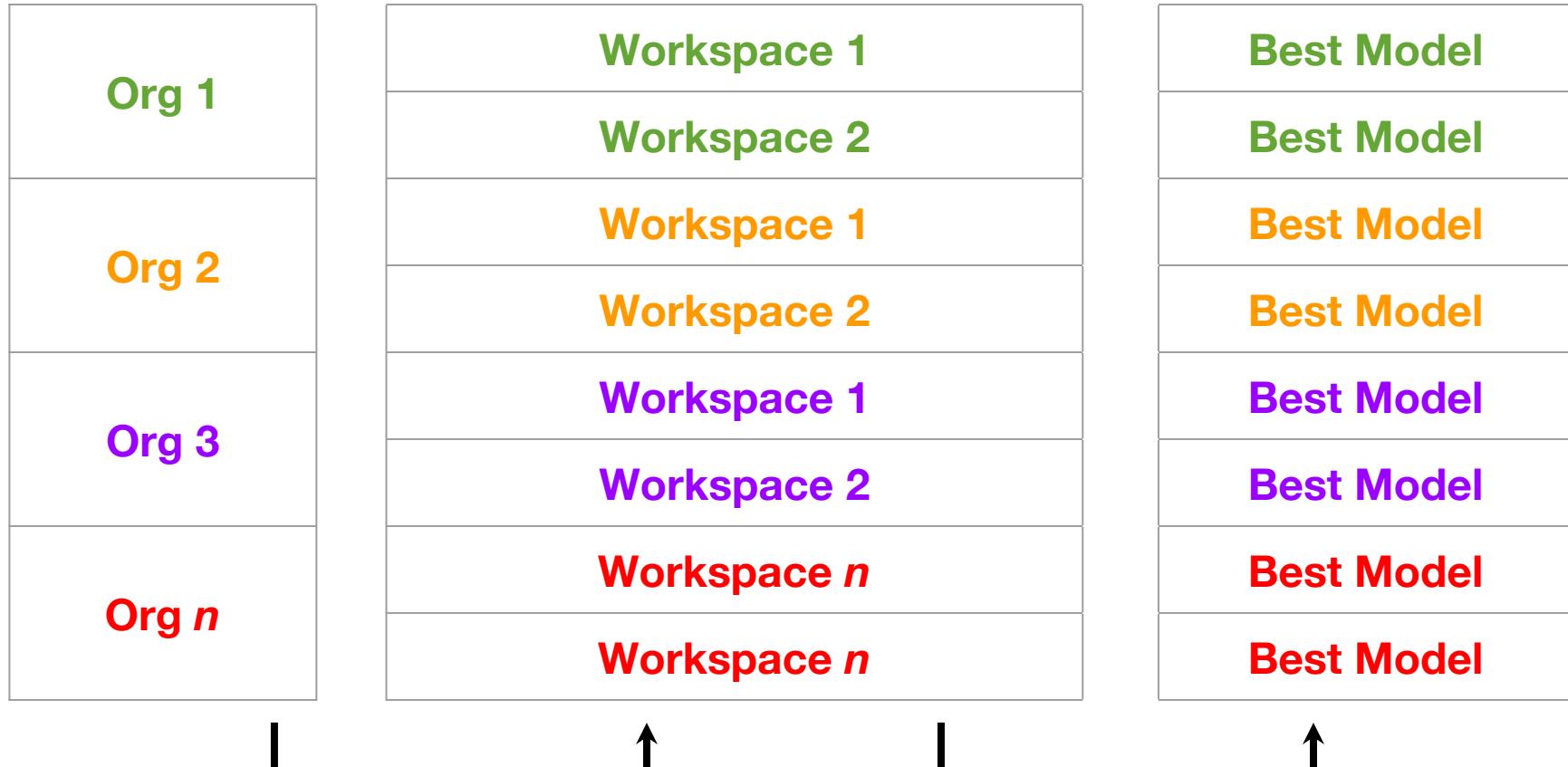
Predictor

Machine Learning Components

Compute (Spark Cluster)

Model and State Persistence (HDFS)

Model Persistence per customer per workspace

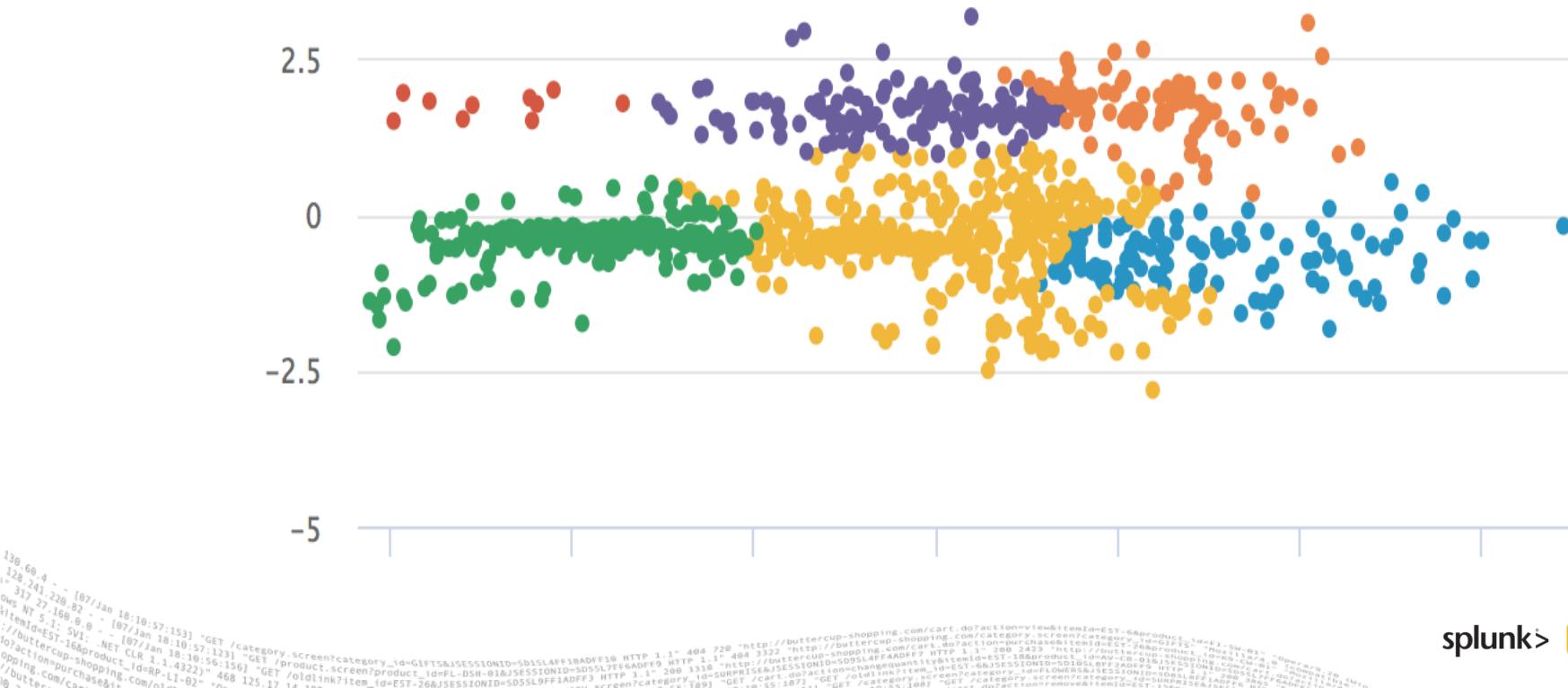


Deep Dive: Relevant Query Recommendation



How Do We Recommend Relevant Queries?

- ▶ Clustering (Unsupervised ML Technique)



Example of Clustered Queries

Show me sales last week



Trained Model

Show me sales in SF last week
Show me sales in California last year
Show me sales in 2017 vs 2016 in California

Architecture

Relevant Query Recommendation REST API

Model Builder

Predictor

Machine Learning Components

Compute (Spark Cluster)

Model and State Persistence (HDFS)

Demo

Where next?

1. More user interface driven ML!
2. Sequence to sequence prediction - predict Next Query for Dialog based interaction
3. Federated learning - sharing trained models (not data) to help everyone

Key Takeaway

Machine learning can be used to solve a myriad of problems.

We have shown you production examples of machine learning used to improve the usability of an application.

You can use the same techniques in your daily work to accomplish similar results.

Next steps

- ▶ Get Project Natural Language Search
 - Dipock Das - dipock@splunk.com
 - Melissa Gannes - mgannes@splunk.com
- ▶ Get the MLTK
- ▶ Get on the ML Advisory program
 - Work with your Rep
 - https://www.splunk.com/en_us/software/splunk-enterprise/machine-learning.html#MLTK-FAQ

Other sessions

Machine Learning & Natural Language Processing at BMW (FN1199)

11:30 yesterday

Spreading the Word: How Chat and Voice Is Transforming Splunk in Retail AI Ops (FN1572)

4:30 yesterday

Ask Splunk! Using natural language, voice and chat with Project Natural Language Search (FN1615)

12:15 today

Thank You

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