

# Online Mental Disorder Analysis

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Improving feature engineering and analysis with  
Elasticsearch and Kibana

# About me



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ML & NLP / Lecturer / Blogger

7+ years: Financial & Graph Data Analysis

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What role does ML play in Search?

What role does Search play in ML?

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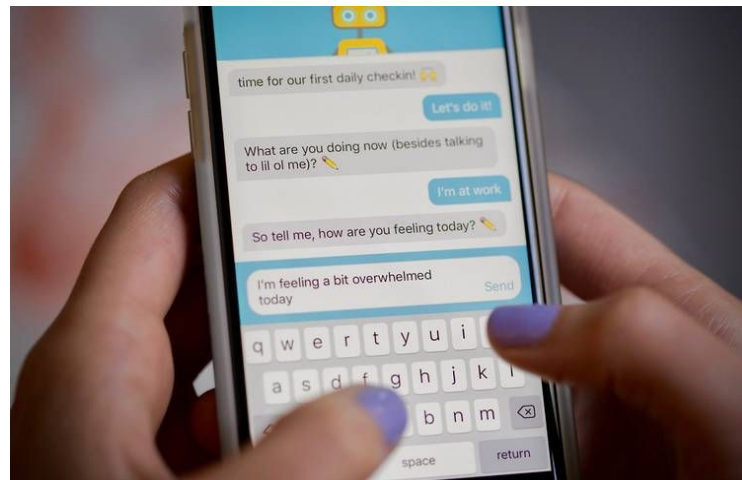
# Journey

- Overview
- Introduce the Data
- Index Mapping
- Data Preprocessing & Ingest Pipeline
- Custom Analyzers
- Demo (Querying and Visualizations)



# Motivation

- Mental disorders impair ability to conduct daily functions
- Leverage *search* and *analytics* to *extract* and *explore* hidden and complex linguistic behaviour from natural language data (e.g., slang, emoticon, stopwords, misspelling, etc.)
- Use *insights* to improve machine learning systems that power chatbots (e.g., monitor and alleviate mood)



*Woebot.io (mood tracker)*

# Overview

## Goals:

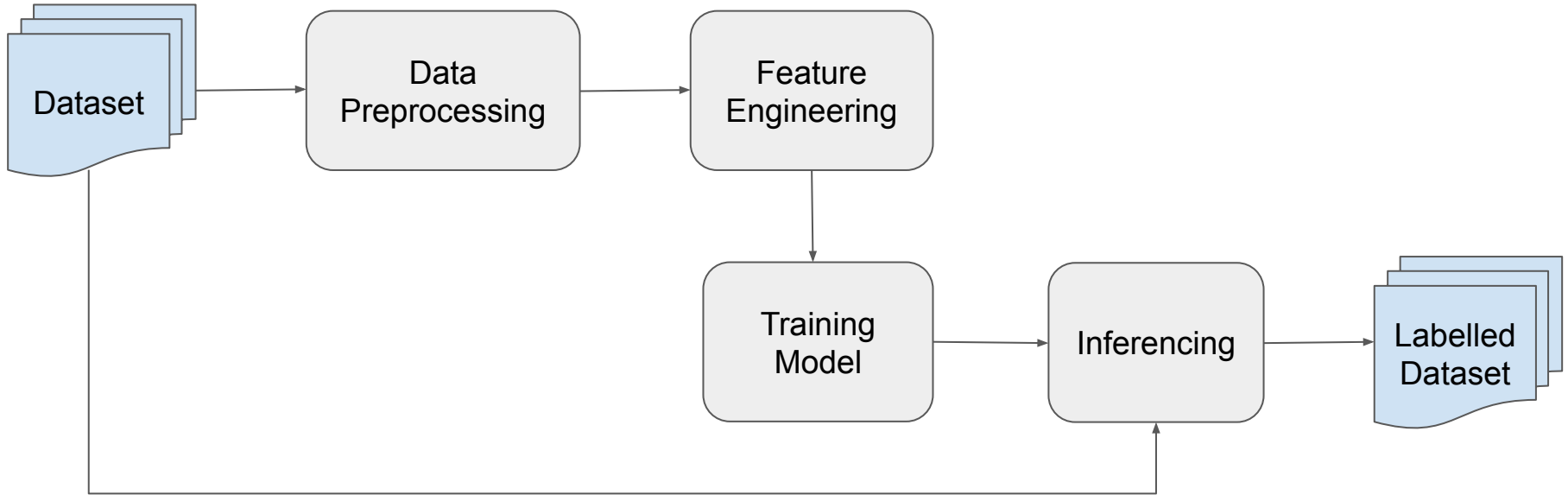
- To broadly demonstrate how to leverage Elasticsearch's *ingest pipeline* and *custom analyzers* for *preprocessing* and *feature engineering*
- To introduce *common best practices* for dealing with natural language data
- To discover *insights* that assist to improve feature engineering and ML models

**Target Audience:** Data Scientists / Data Engineers

**Prerequisites:** Assumes basic knowledge of Elasticsearch, Kibana, and Python

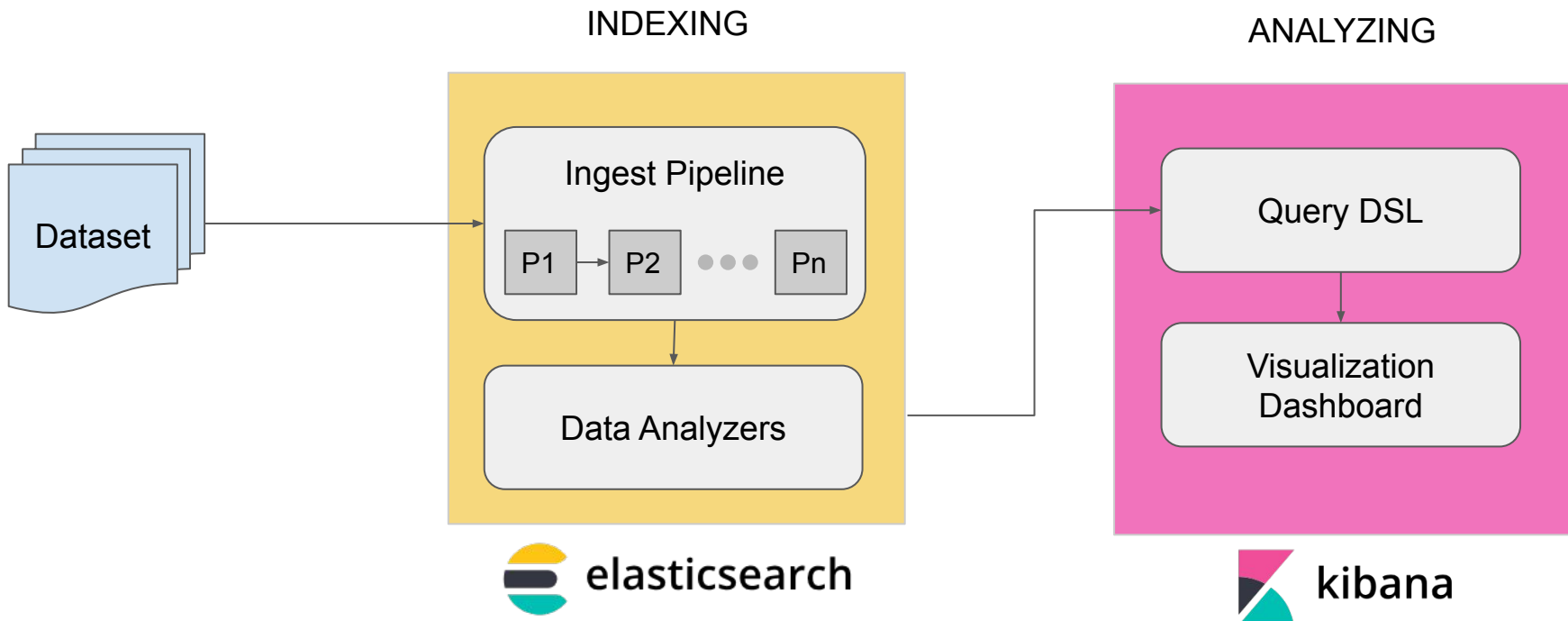
**Duration:** 30 minutes (15 minute demo included)

# Scenario - Typical Machine Learning Pipeline





# Framework

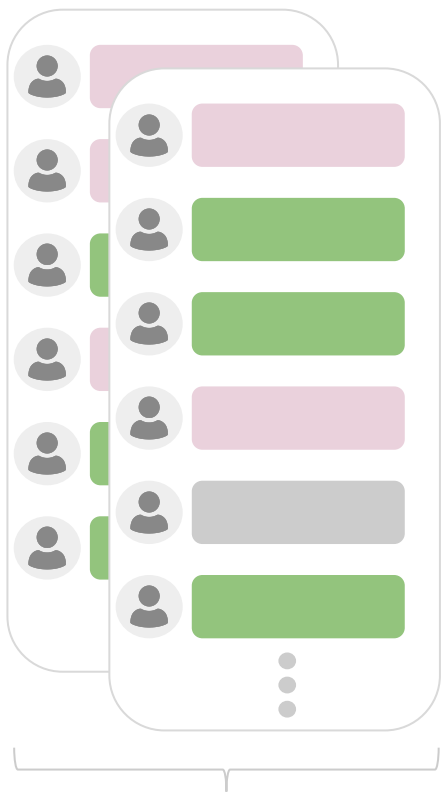


# Data Collection

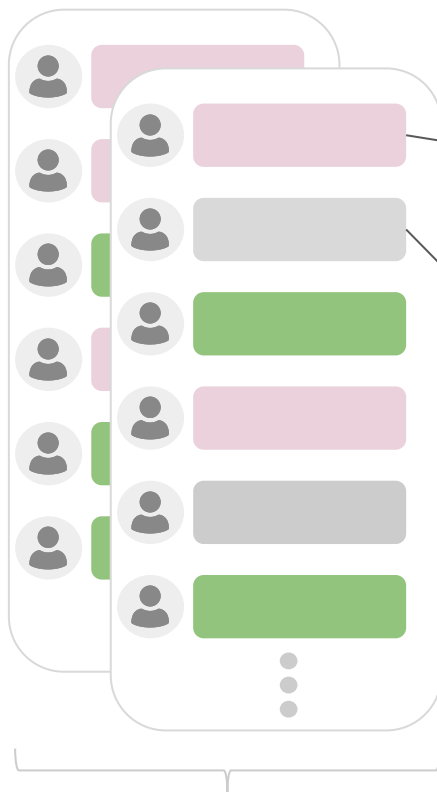
- Online self-reported, mental disorder cases (via “I am diagnosed with X”) :
  - Bipolar disorder: *periods of depression and abnormally elevated mood* (278)
  - Border personality disorder: *longstanding mood swings* (203)
- Normal user profiles (548)

text	polarity	emotion	emotion_2	ambiguous	dt	date	user_id	user_type
@DerekActual hehe Yeah it's definitely 1 that ...	1	joy	anger	True	1.533333	2014-09-30 23:26:47	1	bipolar
@DestinyTheGame Omg plz bring it out for pc.	0	anticipation	joy	False	4.050000	2014-09-30 23:28:19	1	bipolar
@Redtippertruck with great pleasure. Xxx	1	joy	0	False	1.316667	2014-09-30 23:32:22	1	bipolar
@TherapyAfterCSA every day. Xxx	0	joy	trust	False	1.650000	2014-09-30 23:33:41	1	bipolar
@Redtippertruck Hehe I signed it lol. Also ask...	1	sadness	joy	True	7.033333	2014-09-30 23:35:20	1	bipolar

# Data - User Timeline



Control group



Diagnosed group

```
{
  "text": "@DerekActual hehe
  Yeah it's definitely 1 that
  defies logic and explanation
  . Stranger things exist in
  heaven & earth..",
  "polarity": 1,
  "emotion": "joy",
  "emotion_2": "anger",
  "ambiguous": True,
  "dt": 1.5333333333,
  "date": "2014-09-30T23:26:47
  .000Z",
  "user_id": 1,
  "user_type": "bipolar"
},
{
  "text": "@DestinyTheGame Omg
  plz bring it out for pc.",
  "polarity": 0,
  "emotion": "anticipation",
  "emotion_2": "joy",
  "ambiguous": False,
  "dt": 4.05,
  "date": "2014-09-30T23:28:19
  .000Z",
  "user_id": 1,
  "user_type": "bipolar"
}
```

# Storing Data in Elasticsearch

## Considerations before indexing data:

- How to transfer & index data?
  - Logstash / API client (python)
- What scheme or mapping should the data follow?
  - Fields, types, index mapping, preprocessing, etc.
- Any data transformations?
  - Ingest, Analyzers, etc.

```
{
  "text": "@DerekActual hehe
  Yeah it's definitely 1 that
  defies logic and explanation
  . Stranger things exist in
  heaven & earth..",
  "polarity": 1,
  "emotion": "joy",
  "emotion_2": "anger",
  "ambiguous": True,
  "dt": 1.5333333333,
  "date": "2014-09-30T23:26:47
  .000Z",
  "user_id": 1,
  "user_type": "bipolar"
},
{
  "text": "@DestinyTheGame Omg
  plz bring it out for pc.",
  "polarity": 0,
  "emotion": "anticipation",
  "emotion_2": "joy",
  "ambiguous": False,
  "dt": 4.05,
  "date": "2014-09-30T23:28:19
  .000Z",
  "user_id": 1,
  "user_type": "bipolar"
}
```

# Indexing

## How to transfer index data?

- API client (Python library)
- Data is available in dataframe format
- Convert data to JSON
- Bulk insert data with Python library
  - Fast / Efficient
  - Flexibility in fields to include
  - Perform any transformations
  - ([link to notebook](#))

```
# example code of how to convert one user into json
bipolar.group[1]["date"] = bipolar.group[1].index
bipolar.group[1]["user_id"] = 1
bipolar.group[1]["user_type"] = "bipolar"
bipolar.group[1].to_json(orient="records", date_format="iso",
                        path_or_buf="data/user_json/user.json",
                        index=True)

converted = json.load(open("data/user_json/user.json"))
converted[0:2]
```

```
[{'text': '@DerekActual hehe Yeah it's definitely 1 that defies logic
and explanation. Stranger things exist in heaven & earth..',
  'polarity': 1,
  'emotion': 'joy',
  'emotion_2': 'anger',
  'ambiguous': True,
  'dt': 1.5333333333,
  'date': '2014-09-30T23:26:47.000Z',
  'user_id': 1,
  'user_type': 'bipolar'},
```

# Index Mapping

Index mapping provides a way of formatting or schematizing data:

- Configure default pipeline of processors
- Declare field types
- Configure custom analyzers
- ...

```
}
},
"mappings": {
  "_doc": {
    "properties": {
      "date": {"type": "date"},
      "text": {
        "type": "text",
        "fields": {
          "ttokens": {type: "text"},
          "stopwords": {type: "text"},
          "positive_emoticons": {type: "text"},
          "negative_emoticons": {type: "text"}
        }
      },
      "emotion": {"type": "keyword"},
      "emotion_2": {"type": "keyword"},
      "ambiguous": {"type": "boolean"},
      "dt": {"type": "float"},
      "user_id": {
        "type": "text",
        "fields": {
          "keyword": {"type": "keyword"}
        }
      }
    }
  }
}
}
```

# Ingest Pipeline

- Provides a mechanism to preprocess data before indexing it
- An ingest pipeline is made of **processors**:
  - Convert labels with 'set'
  - Lowercase with 'lowercase'
  - Extracts structured field with regex using 'grok'
  - Replace text with regex using 'gsub'



# Analyzers

- Analyzers provide a way to improve search and conduct special analyses on data
- We will use analyzers to **discover linguistic phenomena**:
  - Twitter special tokenizer
  - Extract stopwords from predefined list
  - Obtain positive and negative emoticons

<MENTION> thanks so much for following!  
god bless! :-> <HASHTAG>

[ <MENTION>, thanks, so, much, for, following,  
!, god, bless, !, <HASHTAG> ]

[ so, for, !, ! ]

[ :-> ]



# Future Ideas

- Build and train ML model based on processed text and features
- Store ML model and use Logstash to ingest real-time profiles of online mental disorder cases via “I am diagnosed with X” filter
- What can we learn from natural language that generalizes to logs, metrics, etc.)? `55.3.244.1 GET /index.html 15824 0.043`
- Generalize pipeline to different conversations (chatbot, reviews, language etc.)



# References

- [Elasticsearch 6.6 Reference](#)
- [Elastic Resources and Training](#)
- [Clinical NLP with Elasticsearch](#)
- [OpenNLP with the Elastic stack](#)
- [MIDAS: Mental illness detection and analysis via social media](#)

# Q&A

**Demo**