

DATASIX

Please Join our WiFi Network
WeWork Guest



Real-time AI

Out of the Lab, Into Prod



Introduction

Hilton Rosenfeld



- Developer / Architect
- Application Modernization
- Digital Transformation
- IT Operations Management
- CI / CD

Build your own NLP text classifier and
expose it as an API



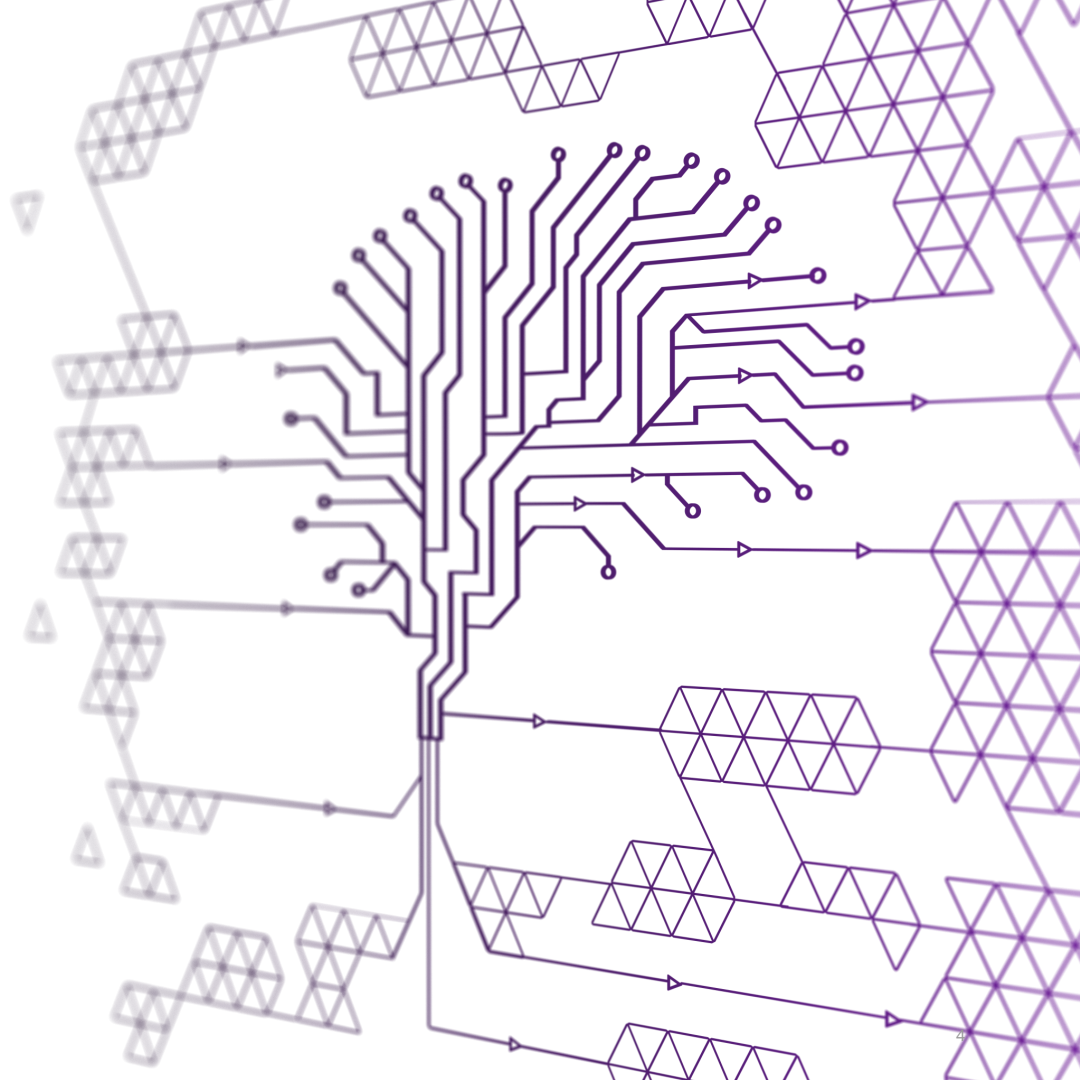


Housekeeping

- Break(s) will be provided.
- Refreshments will be served.
- Ask Questions !
- Scroll through exercises at your own pace.
- Connectivity Issues:
 - disable VPN and/or Firewall



What we are
building today

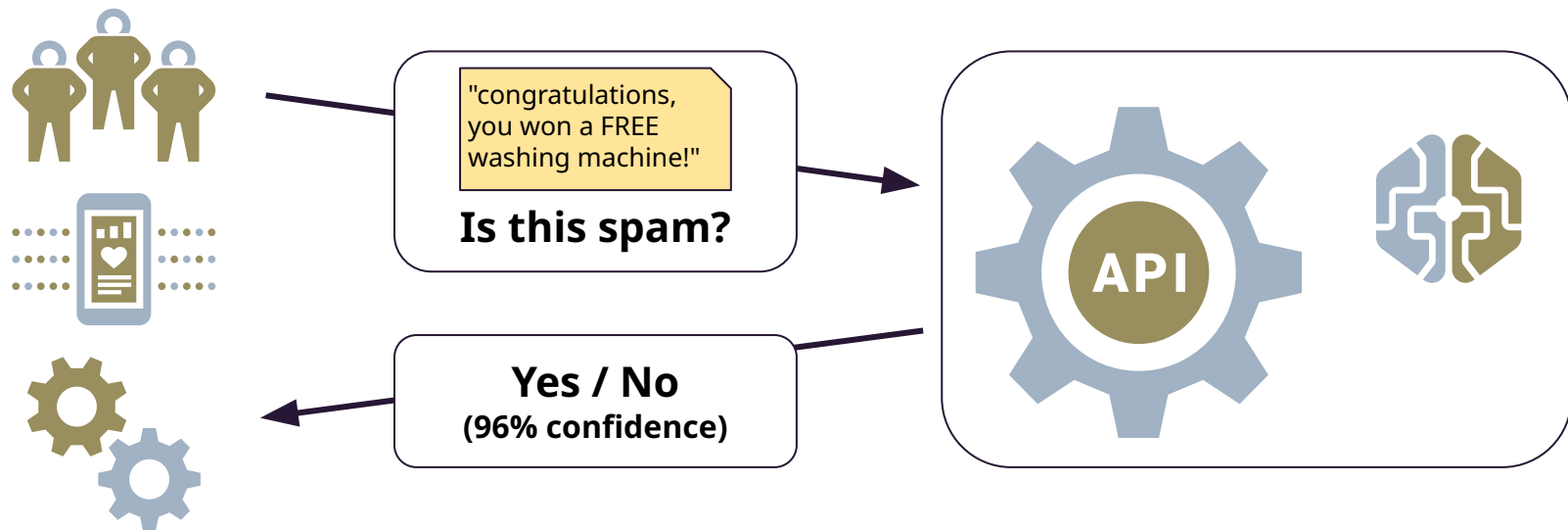


» Spam

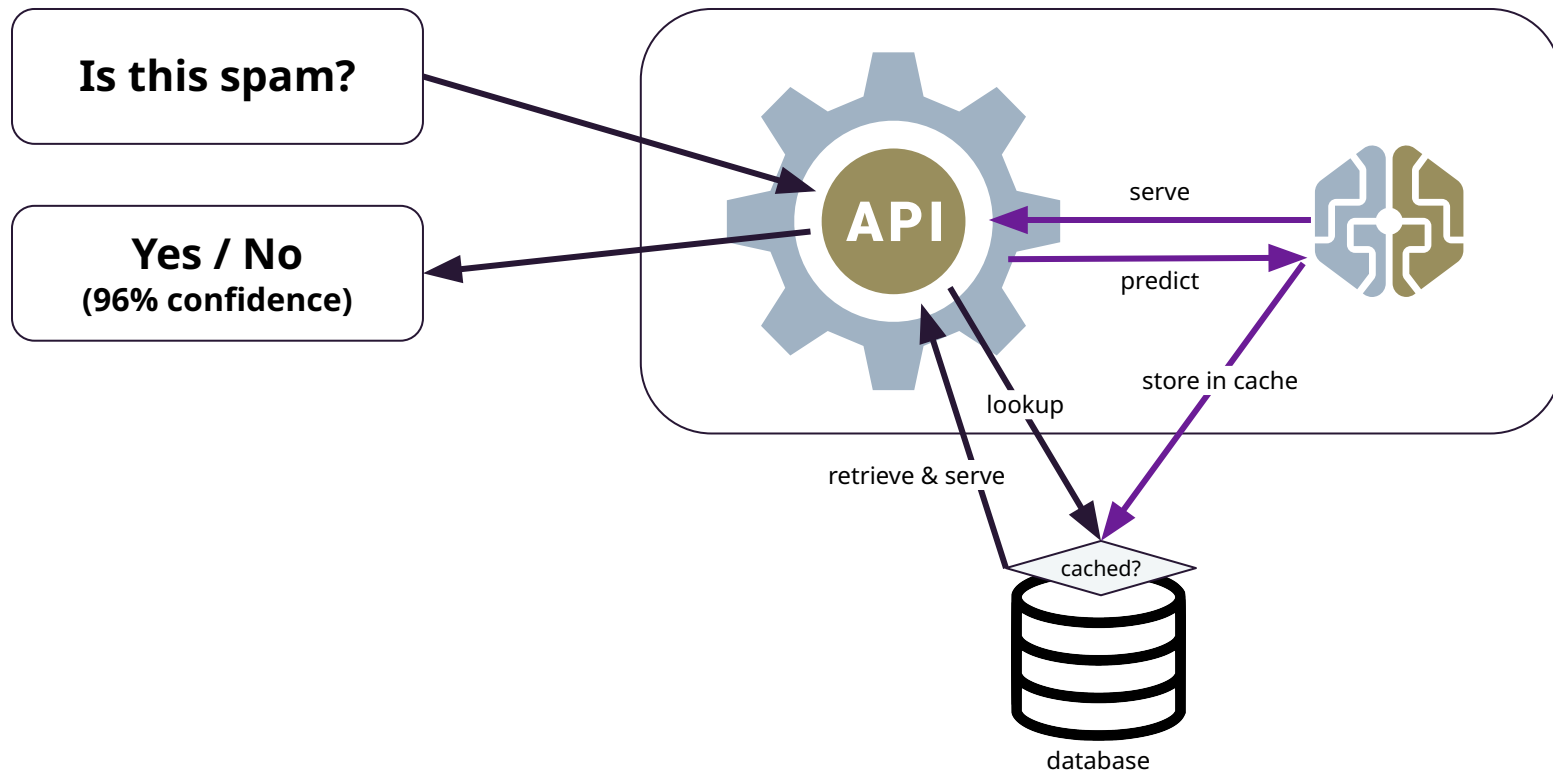
- 85% of emails are spam:
122, 330, 000, 000 daily
 - Advertising - 36%
 - Adult-related content - 31.7%
 - Financial matters - 26.5%
 - Scams and fraud - 2.5%
- For every 12,500,000 emails sent, spammers receive one reply.
- Email spam costs businesses **\$20.5 billion** annually.



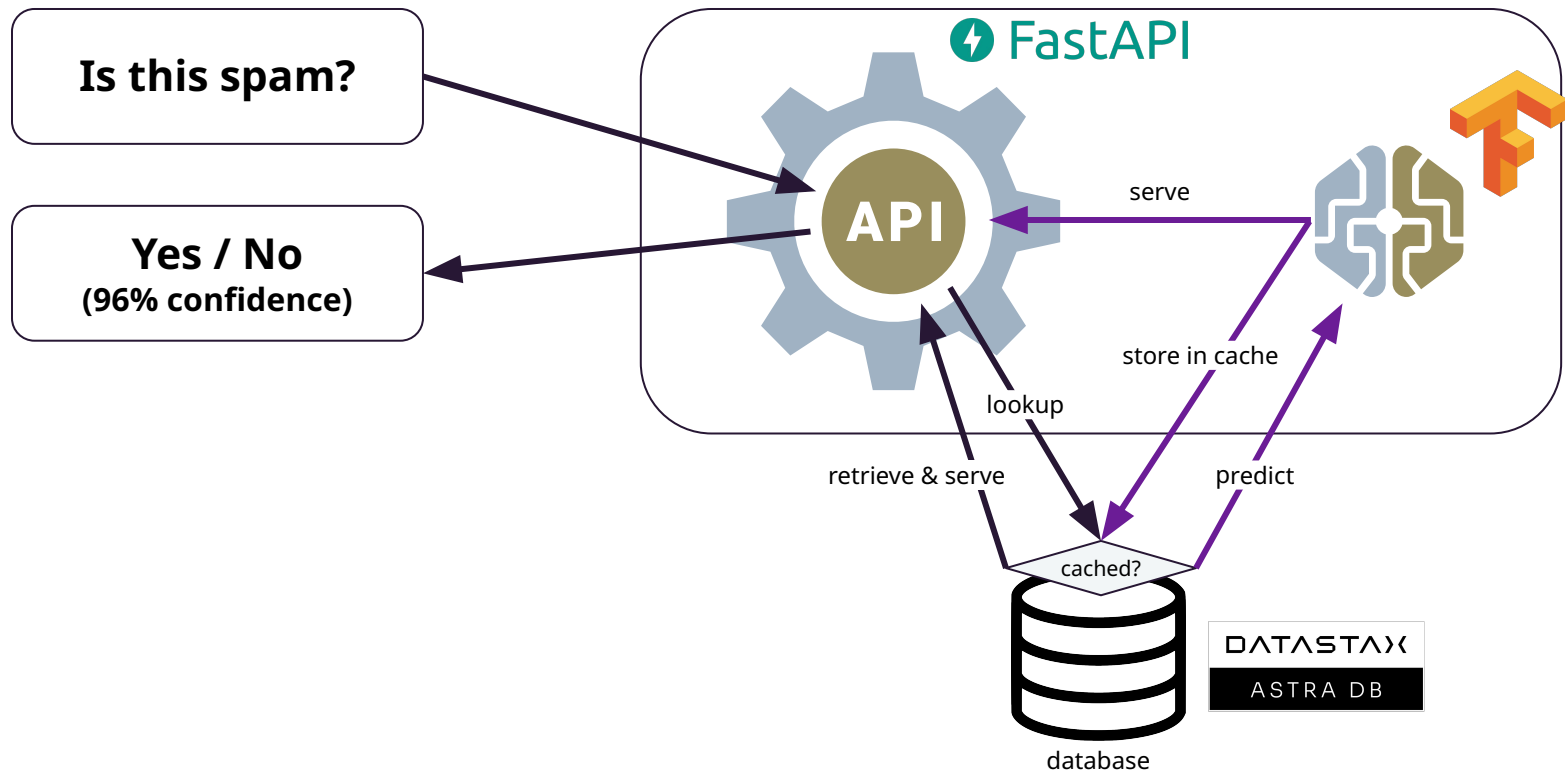
› What we want



› Architecture sketch



» Tech stack

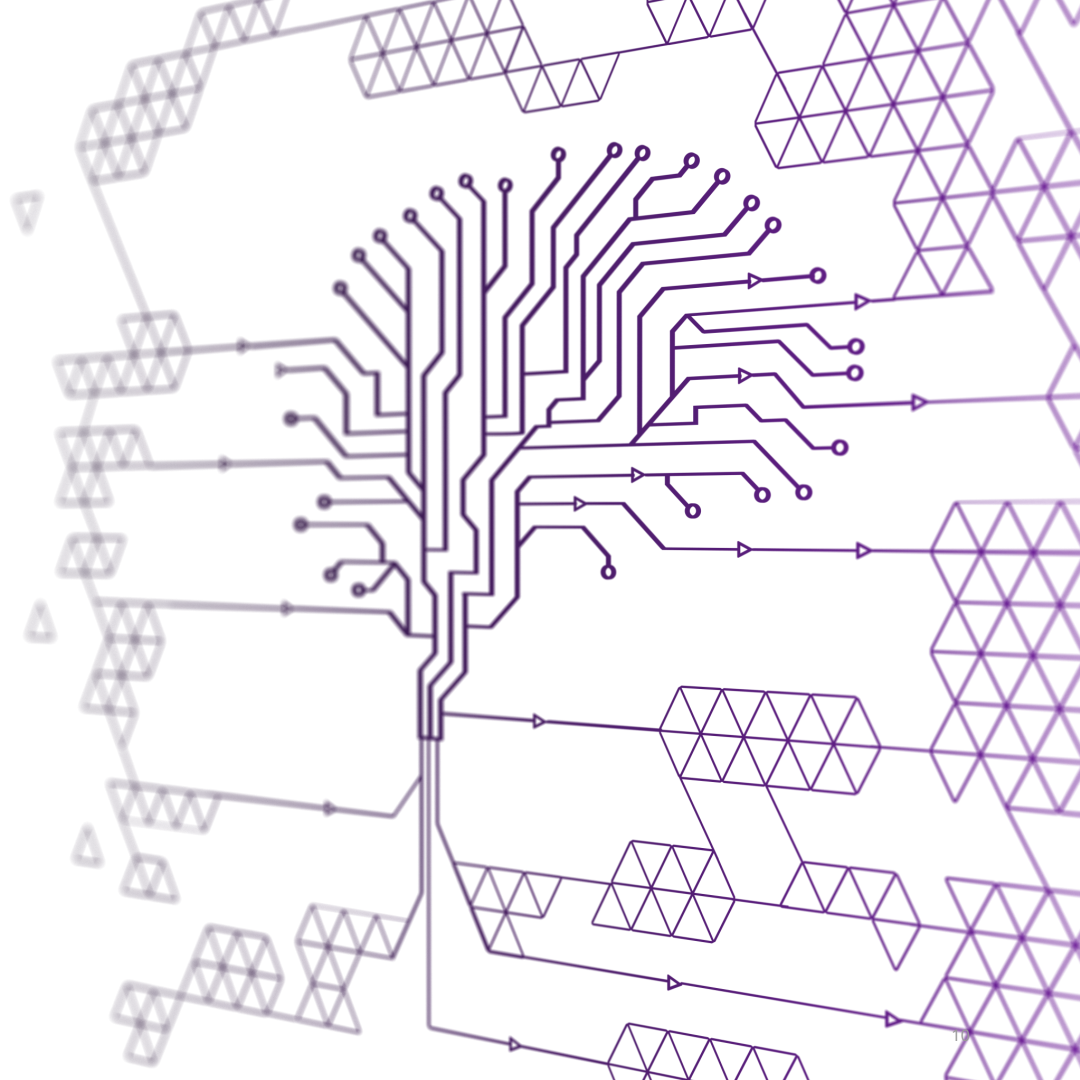


» Applying Machine Learning to the problem of Spam filtering

1. Create a deep learning model from multiple spam datasets
2. Make that available as a REST API using Jupyter, TensorFlow, FastAPI and DataStax Astra DB.

- Combining 2 open source spam datasets curated by The University of California, Irvine (UCI).
- Preparing a new dataset to be fully ready for training a model.
- Building and training an LSTM model leveraging Keras and TensorFlow.
- Creating an API to expose the model for use leveraging FastAPI
- Caching predictions and inference data

➤ The AI



» AI and ML

"ML: LSTM RNN for NLP"

Machine Learning = *algorithms that improve by being fed data, without explicit instructions what to do.*

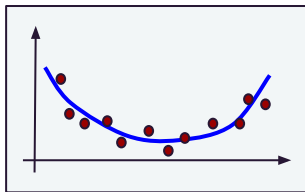
It's essentially statistical inference (*with superpowers*).

Lots of math involved (*linear algebra, calculus, probability/statistics*).

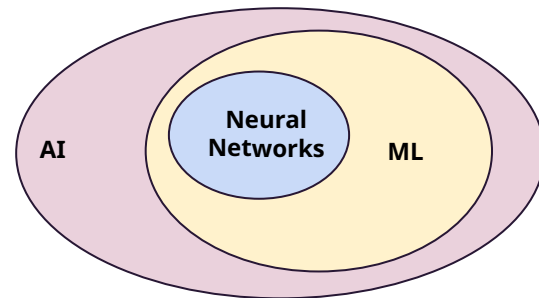
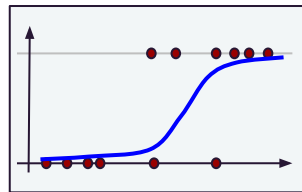
Nowadays accessible as neatly-packaged tools (good for us!)

Simple examples of ML:

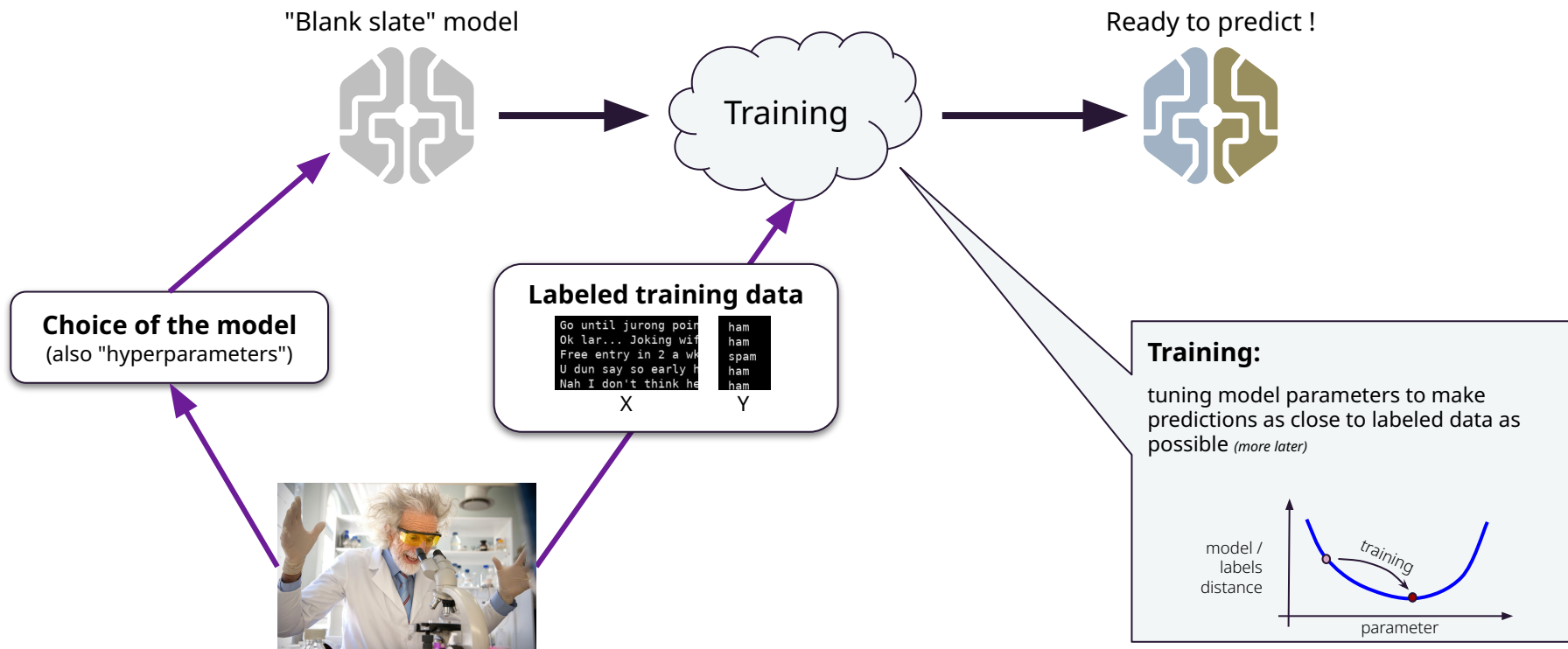
Least-Squares Fits



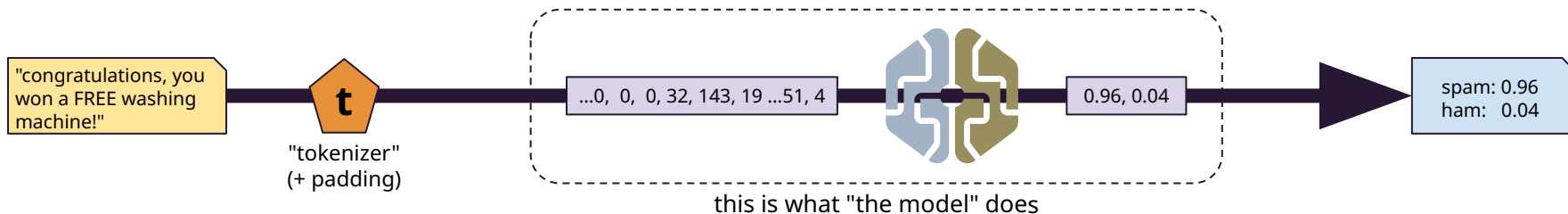
Logistic Regressions



› Supervised learning



➤ A closer look: numeric encoding



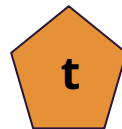
Prepare the dataset before training

Labeled training data

| | |
|-----------------------|------|
| Go until jurong point | ham |
| Ok lar... Joking wif | ham |
| Free entry in 2 a wk | spam |
| U dun say so early h | ham |
| Nah I don't think he | ham |
| FreeMsg Hey there da | spam |

X Y

...0, 0, 0, 32, 143, 19 ... 51, 4
...0, 0, 0, 0, 41, 4 ... 65, 10
...0, 0, 0, 0, 0, 21 ... 84, 5
...0, 0, 0, 0, 81, 205 ... 3, 51
...0, 0, 0, 7, 56, 4 ... 15, 75
...0, 0, 0, 0, 0, 194 ... 2, 65



1, 0
1, 0
0, 1
1, 0
1, 0
0, 1

Split "train" / "test"

train

test

...0, 0, 0, 32, 143, 19 ... 51, 4
...0, 0, 0, 0, 41, 4 ... 65, 10
...0, 0, 0, 0, 0, 21 ... 84, 5
...0, 0, 0, 0, 81, 205 ... 3, 51
...0, 0, 0, 7, 56, 4 ... 15, 75
...0, 0, 0, 0, 0, 194 ... 2, 65
... ..

› The antispam model architecture

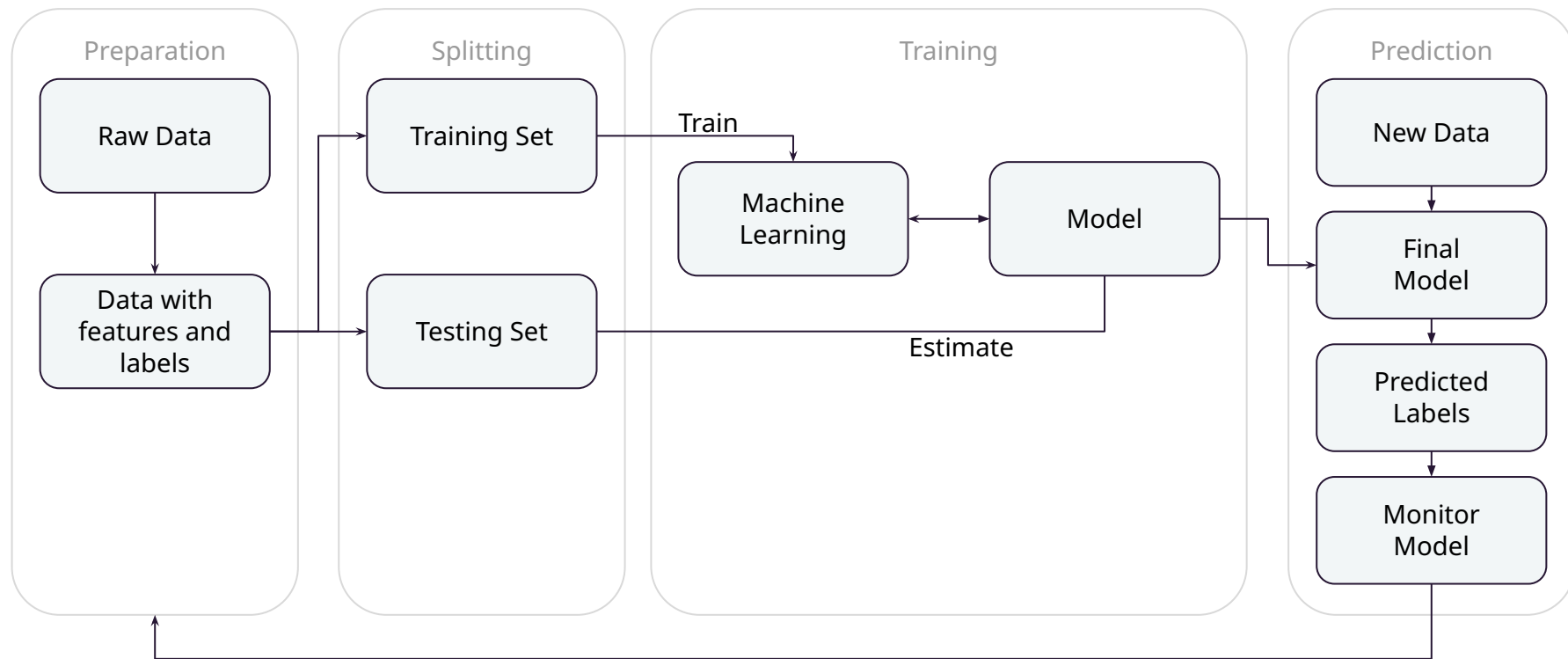


A stack of layers:

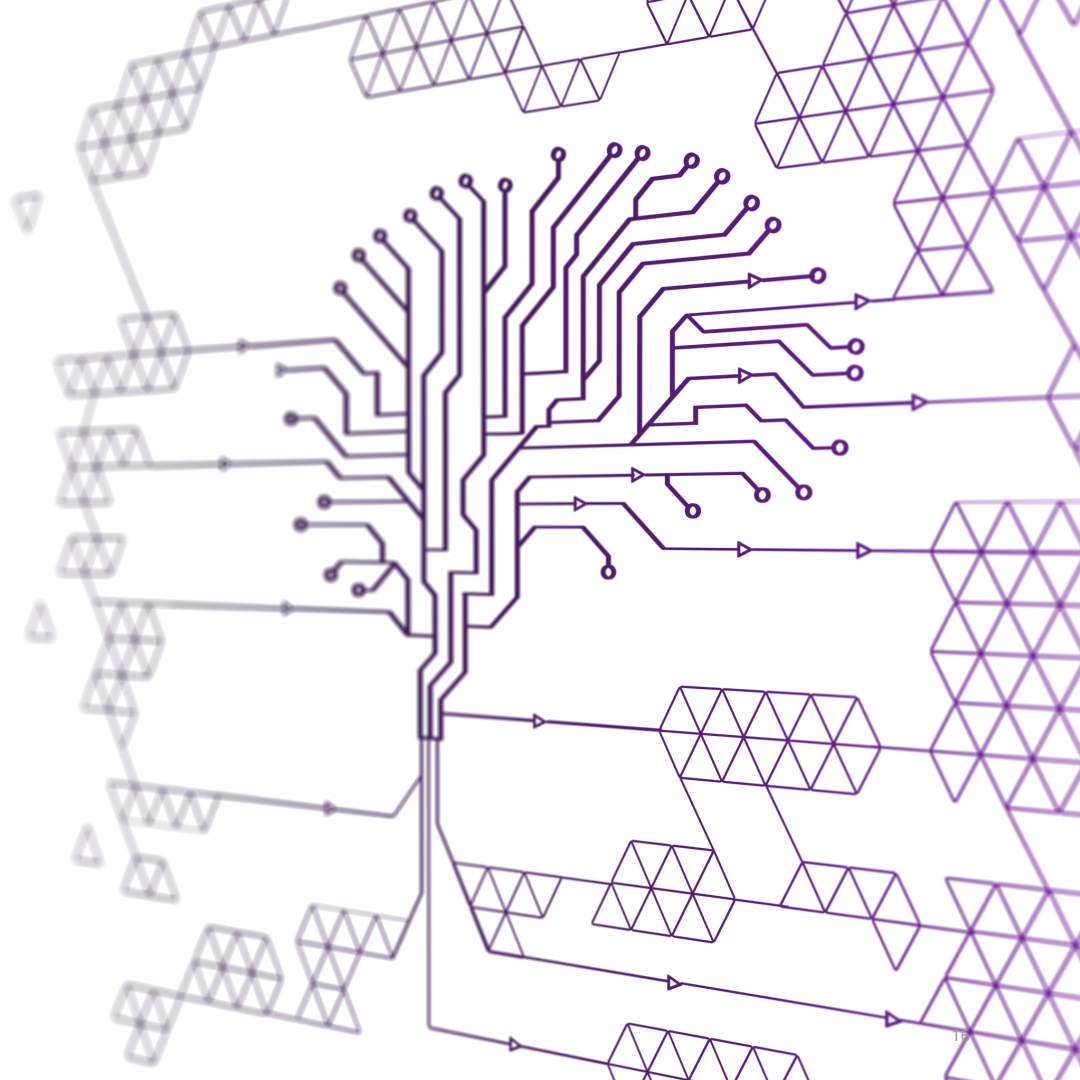
- translate an input number into a vector
- randomly disables pieces of network to enhance training ("dropout")
- the LSTM (recurrent within the layer)
- reducing to a 2-option output (spam/ham)
- "finalize" the model: ready to train!

```
model = Sequential(  
    model.add(Embedding(maxNumWords, embedDim, input_length=1))  
    model.add(SpatialDropout1D(0.4))  
    model.add(LSTM(LstmOut, dropout=0.3, recurrent_dropout=0.3))  
    model.add(Dense(2, activation='softmax'))  
    model.compile(loss='categorical_crossentropy', optimizer='adam')
```

» The AI



➤ Database



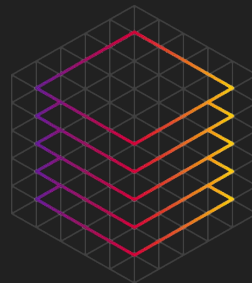
› Database-as-a-Service powered by Apache Cassandra

DATASTAX

ASTRA DB

Sign up for free

www.datastax.com/products/datastax-astra



Build in any language



Global Scale



Zero Operations



Zero lock-in

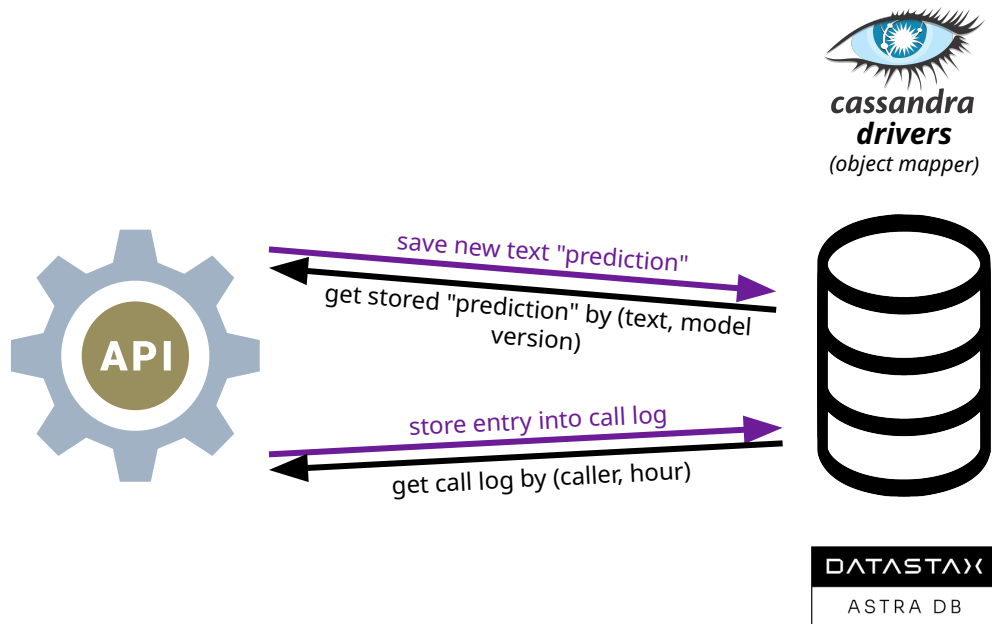


Enterprise security



Extensibility

› DB Query patterns

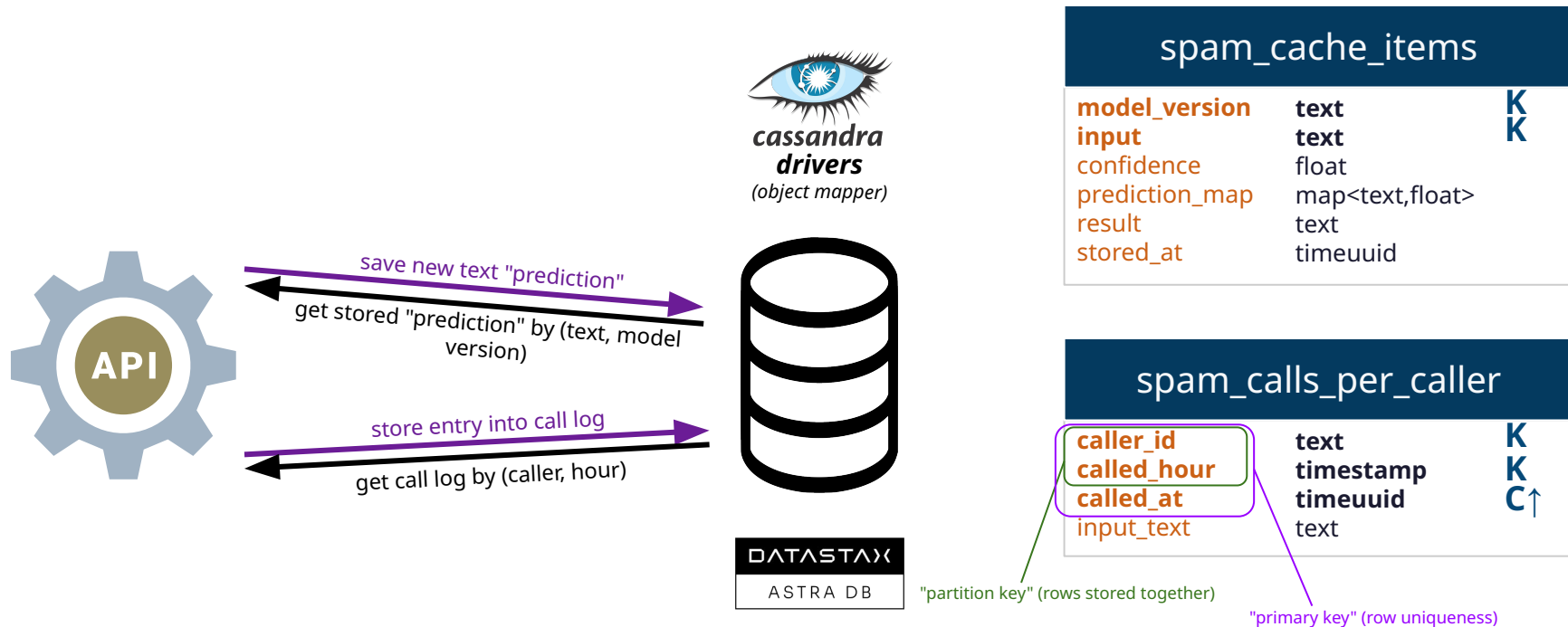


Data modeling, the Astra way:

"design the table after the query"

(also: tables are partitioned!)

› DB Query patterns



› Real-Time Data Processing (Feature Engineering)

Table Row Examples

| version | input | confidence | prediction_map | result | stored_at |
|---------|--------------------------------------------------|------------|---------------------------------------|--------|----------------|
| v1 | Combing the white hair of the waves blown back | 0.955171 | { 'ham': 0.955171, 'spam': 0.044829 } | ham | 08f849b6-f92a- |
| v1 | We have lingered in the chambers of the sea | 0.992575 | { 'ham': 0.992575, 'spam': 0.007425 } | ham | 090de000-f92a- |
| v1 | Till human voices wake us, and we drown. | 0.988767 | { 'ham': 0.988767, 'spam': 0.011233 } | ham | 092371b8-f92a- |
| v1 | Click TO WIN a FREE CAR | 0.739895 | { 'ham': 0.260105, 'spam': 0.739895 } | spam | 5d957634-f929- |
| v1 | By sea-girls wreathed with seaweed red and brown | 0.912248 | { 'ham': 0.912248, 'spam': 0.087752 } | ham | e9424acc-f929- |
| v1 | When the wind blows the water white and black. | 0.979708 | { 'ham': 0.979708, 'spam': 0.020292 } | ham | d4647bca-f929- |
| v1 | I have seen them riding seaward on the waves | 0.977917 | { 'ham': 0.977917, 'spam': 0.022083 } | ham | d44dafda-f929- |

| spam_cache_items | | |
|------------------|-----------------|---|
| model_version | text | K |
| input | text | K |
| confidence | float | |
| prediction_map | map<text,float> | |
| result | text | |
| stored_at | timeuuid | |

| caller_id | called_hour | called_at | input |
|----------------|---------------------------------|--------------------------------------|----------------------------------------|
| 127.0.0.1 | 2023-05-25 12:00:00.000000+0000 | 77049492-fafa-11ed-8a10-a2ab11a8d5af | Click TO WIN |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | d42292e6-f929-11ed-a192-1a3838142467 | I have seen them riding seaward on |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | d438241c-f929-11ed-a192-1a3838142467 | When the wind blows the water white |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | e901c97a-f929-11ed-a192-1a3838142467 | I have seen them riding seaward on |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | e9174da4-f929-11ed-a192-1a3838142467 | When the wind blows the water white |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | e92cc8dc-f929-11ed-a192-1a3838142467 | By sea-girls wreathed with seaweed red |
| 192.168.150.76 | 2023-05-23 05:00:00.000000+0000 | 08771a08-f92a-11ed-a192-1a3838142467 | I have seen them riding seaward on |

| spam_calls_per_caller | | |
|-----------------------|-----------|----|
| caller_id | text | K |
| called_hour | timestamp | K |
| called_at | timeuuid | C↑ |
| input_text | text | |

"partition key" (rows stored together)

"primary key" (row uniqueness)



Hands On Time



› Tools

- Nothing to Install!
- GitHub repository:
bit.ly/irt-ai-as-an-api



Source code + Exercises + Slides



Gitpod

Cloud Development Environment



Database



FastAPI

API

› Know your tools

- Python
- Jupyter
- TensorFlow
- Keras
- FastAPI
- Astra DB



» Lab Steps

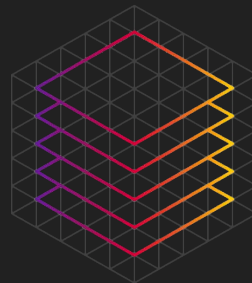
1. Initialise GitPod
2. Create a database in Astra DB
3. Inspect the Dataset
4. Train the Model in Jupyter
5. Expose the Model as an API
6. Use the API
7. Inspect the database

DATASTAX

ASTRA DB

Sign up for free

www.datastax.com/products/datastax-astra



Build in any language



Global Scale



Zero Operations



Zero lock-in



Enterprise security



Extensibility

Available from  **aws marketplace**



DATASTAX

HANDS-ON WORKSHOP | MELBOURNE, AU

› IRT AI/ML – Out of the Lab, Into Production

20 July, Thursday | 5pm - 7pm | WeWork Office, Level 22, 120 Spencer St



Hilton Rosenfeld

Data Architect, DataStax

REGISTER NOW



DATASTAX



Thank You

DATASTAX