# Tech Names as a Service



### Vincent D. Warmerdam koaning.io - fishnets88 - GoDataDriven - imgs @ nounproject



### Today

In this talk I will discuss a hobby project.

The first half of the talk will focus on the machine learning part of what I did and what I can do better.

The second part of the talk will focus on the architecture/cloud part of what I did and what I can do better.

I will make sure to mention lessons learned along the way.

# Who is this guy?

3 years @ GDD koaning.io PyData Amsterdam Rstudio training partner Machine Learning Meetup free open sessions in data 39 endorsements for awesomeness

> I write code, I solve data problems, ASK ME ANYTHING.



### Vincent Warmerdam Pokemon Master at GoDataDriven GoDataDriven • Vrije Universiteit Amsterdam

# my linkedin profile

R, python, javascript, shiny, dplyr, purrr, ditto, ggplot, d3, canvas, spark, sawk, pyspark, sparklyR, lodash, lazy, bootstrap, jupyter, vulpix, git, flask, numpy, pandas, feebas, scikit, pgm, bayes, h2o.ai, sparkling-water, tensorflow, keras, onyx, ekans, hadoop, scala, unity, metapod, gc, c#/c++, krebase, neo4j, hadoop.

# my linkedin profile

R, python, javascript, shiny, dplyr, purrr, ditto, ggplot, d3, canvas, spark, sawk, pyspark, sparklyR, lodash, lazy, bootstrap, jupyter, vulpix, git, flask, numpy, pandas, feebas, scikit, pgm, bayes, h2o.ai, sparkling-water, tensorflow, keras, onyx, ekans, hadoop, scala, unity, metapod, gc, c#/c++, krebase, neo4j, hadoop.

I typically ask recruiters to point out which of these are pokemon.

This whole thing started when I realized that recruiters have a hard time distinguishing pokemon names against names from the coding ecosystem.

I can't blame 'em too much ...





Jolteon 💡 Babel + Electron + React + Browserify + SASS application stack. El... 🚖 999 🍴 38



Klang Clojurescript logging library

🚖 83 🝴 1



Rhydon Save file editor for Pokemon Red/Blue/Yellow. 🚖 24 🍴 3





Gloom

A dark and gloomy

pastel color syntax

theme for Atom.

🚖 83 🝴 10

Shelmet

A JVM Heap dump

viewer - a souped-

up jhat in scala

🚖 22 🝴 1

Electrode Electrode Platform for Universal React + Node application for computing pow... 🚖 801 🖞 86 🚖 922 🝴 71

Geodude

A tiny command-

line utility for

geocoding

addresses.

🚖 65 🍴 3

Natu

Natural units in

Python

🚖 21 || 4

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven



Ditto

Lightweight

Markdown

Documentation

System

🚖 366 || 43

Charmander Charmander Scheduler Lab -Mesos, Docker, InfluxDB, Spark 🚖 57 🝴 12



Elekid Resolver for React's Server Side Render on Module, ReactElement or El...

🖕 18 🍴 0

Lucario

The best flat

theme for Vim,

Atom, Sublime

Text, Jetbrains

Editors, T...

☆ 334 1 57

Binacle

Full-bin indexation

of binary files

🚖 56 🝴 5

Munna Simple wrapper cache for ActiveRecord, Ruby Object 술 18 🍴 4





Castform Async form validation on the client and server. 🚖 16 🍴 1



Golbat Golbat is an Android library that helps you on

🚖 16 🍴 2



Blastoise tiny relational database 🚖 14 🝴 1 working with camera, ...



Vulpix

Automated code

grading system

🚖 46 🍴 9

Kadabra

[DEPRECATED]

Kadabra is my

automatic LFI

Exploiter and

🚖 13 🍴 6

Scanner, writt...

Gastly

Create

screenshots or

previews of web

pages using



Klink A Simple and Clean Sphinx Docs Theme 🚖 13 🍴 5



Showcase of GitHub repos with Pokémon names

### ★ Star 🛛 238 🛛 😵 Fork 🗧 23 😏 Tweet 🖬 📩 Like 237 Share



Snorlax

The ultimate lazy

library. An

example for being

a good library

citize...

🚖 181 🝴 10

Smeargle

Vim plugin for

colouring the text

background based

on information

min...

🚖 52 🝴 1

Rapidash Rapidly develop your API client

🚖 151 🝴 17

Abra

Assembly Based

ReAligner

🚖 50 🝴 10

Mew Messaging in the Emacs World 🚖 134 🍴 36



Tangrowth Python3.5 async crawler example with aiohttp and

asyncio 🚖 130 🝴 13

Tentacool REST API to manage Linux networking via





Shellder 🐚 Featured zsh/fish shell theme 🚖 129 🍴 22



Slowpoke Rack::Timeout is great. Slowpoke makes it better. 🚖 123 🝴 5



**Eevee** 基于 github page 的在线编辑 blog 平 台 🚖 101 🍴 24



Porygon the unofficial Pokémon Go Plus SDK 🚖 83 🝴 5



Onix A convenient mapping between ruby objects and the ONIX XML specificat... 🚖 38 🍴 34



Treecko A collection of functional and immutable helpers for working with tre... 🚖 29 🍴 1



Dratini Dratini is a neat network abstraction layer. 🚖 29 🍴 4



Scyther The Scyther Tool for the symbolic analysis of security protocols 🚖 27 🝴 12



Klefki Simple substitution cipher module. 🚀 🚖 12 🝴 2



Staryu Open-source 5key Keypad 🚖 12 🍴 3



Hoothoot hoothoot hoothoot hoothoot. Find your path in the forest of nested da... 🚖 12 🍴 1



Diglett Diglett is a cron management system that manage all your cron jobs wi... 🚖 11 🝴 1

7

This whole thing started when I realized that recruiters have a hard time distinguishing pokemon names against names from the coding ecosystem.

I figured making a small web-app that can generate tech names as a service may be a great adventure for fun and profit.



It is a general problem if you think of it. Given a sequence of tokens, can we generate a sequence of tokens that are different but believably similar?



- Things like;
- pokemon •



- pokemon •
- red hot chili pepper lyrics



- pokemon •
- red hot chili pepper lyrics
- ikea furniture names



- pokemon •
- red hot chili pepper lyrics
- ikea furniture names
- notes on a piano •



- pokemon •
- red hot chili pepper lyrics
- ikea furniture names
- notes on a piano •
- anything ipsum



To keep things simple I ventured to make an app that just generates names for now.

You can find the webapp over at tnaas.com.

It is an acronym for Tech Names as a Service.



To keep things simple I ventured to make an app that just generates names for now.

You can find the webapp over at tnaas.com.

It is an acronym for Tech Names as a Service.

The whole point of the app is to find a better name for it.



### **TNaaS: Enterprice Features**

You can select a corpus to generate from. You can select a model to apply. You can click 'go'.



### **TNaaS Bonus Features**

You can select a corpus to generate from. You can select a model to apply. You can click 'go'.

### **Bonus!**

Complete a set of tokens; \*\*\*base, mo\*\*\*, \*\*\*ly

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven



18

# DEMO TIME **Crowd Participation!**



# Part 1 Machine Learning

### **A Sequence of Tokens**

The simplest model you think of;

$$p(t_i|t_{i-1})$$

For every pair of tokens, keep track of how often they occur. You can learn these probabilities given a corpus and then sample;

$$p(t_1) 
ightarrow p(t_2|t_1) 
ightarrow p(t_3|t_2) 
ightarrow .$$



### **A Sequence of Tokens**

A more expressive method might be to not just look at the previous state, but perhaps the state before as well.

$$p(t_i ert t_{i-1}, t_{i-2})$$

Similar method of sampling.

$$p(t_1) o p(t_2|t_1) o p(t_3|t_2,t_1) o p(t_4|t_3)$$

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven



 $(,t_2) 
ightarrow \ldots$ 

### **A sequence of Tokens**

But generally, these chains can have any shape;





### A sequence of Tokens

But generally, these chains can have any shape;



### Who can spot the issue I need to deal with?







### **A Sequence of Tokens**

Markov chains tend to keep on going. They need to stop at some point. Thankfully, bayesians have a solution. We first determine how long the pokemon name needs to be, then we sample tokens.

 $p(T) = p(T|\lambda)p(\lambda)$ 





### Intermediate Results $p(t_i | t_{i-1})$

### **POKEMON** results

lydo

keen

wqool

ryrys

poole

utcala

youtail

olma

elttyp

### Intermediate Results $p(t_i | t_{i-1})$

### **IKEA** results

anapa frodok pasig ripe latrank vis gsoo yirbs ilosseln

### **Intermediate Results** $p(t_i | t_{i-1})$

### **RHCP** Lyrics

Can you believe. Hold me please. By the way I wonder what the wave meant. White heat is screaming in the nearest bin. When I was fortunate I know you must be fat this year. And eat the sun and a Bottom Dollar. Fox hole love Pie in your house now let me spin Feather light but you cant move this

### Ok ... but.

The results are fun and all but not good enough.

- 1. The model is a bit naive and the results could be better.
- 2. Perhaps the end user does not want a name from scratch. Perhaps the user can supply me with some given tokens that I need to finish. Whatever method I come up with should be able to do this too.

Ensembles, anyone?





### Add lexicons manually?





### Or take different datasets



Map the tokens to another space? Phonetics?



Adverserial judges?



# Sample 100 samples, have another algorithm sort them on believability and pick the best performing one.

### Heuristics?

Levenstein-ish approach

Gensim -> word 2 vec? tohen 2 vec 2

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven

35

### Distance measure suggestion. Char2Vec.


## Improving the model: idea 7

Why not deep models?



# **Deep Model Failure**

The standard LSTM trick will not work for my **ultimate** usecase.

Let's take as input '\*u\*a' and 'ou\*a'.

What is the most likely 3rd character? I hope you agree that we cannot 'just sample forward'. It part algorithm problem but part data structure problem too.

# **Completing Tokens is Harder**

Completing tokens may be a better features for end users and it entropy wise it should reduce the search space considerably.

This feature changes the ML task, but may be a better feature.

# **Completing Tokens is Harder**

Completing tokens may be a better features for end users and it entropy wise it should reduce the search space considerably.

## This feature changes the ML task, but may be a better feature.

We cannot just count and estimate probabilities and just sample from them. If we know some part of the sequence then this information needs to propagate forwards and backwards.

How to deal with this?

## **Algorithm Intuition**



## **Algorithm Intuition**



## **Algorithm Intuition**



## **Same Intuition, Advanced Mode**











## Implementation





Every arc is a discrete probability distribution. If you have those you have a defined system, you merely need a tool to sample from it.

# **Towards Implementation**

For anything simple I've used markovify it is a small python library that does nothing more but sampling forward markov chains.

A more proper inference tool for the missing characters situation is pomegrenate. It is a very likeable library for more bayesian-type of algorithms. It is written in cython by a cool guy named Jacob who uses it in the field of genome science.

Might make a cool NumFocus project someday. Keep an eye!

## Implementation

Bulk of the training is done via;

import pomegranate as pg
chain = pg.MarkovChain.from\_samples(names, k=lookback)
dists = [chain.distributions[i] for i in range(lookback)]

These distributions are basically nodes in the probibalistic network. This bit will train the transition probabilities.

kback) bokback)] bibalistic ties.

# Implementatio

## Bulk of the inference;

```
network = pg.BayesianNetwork("name-problem")
network.add_states(*states)
network.bake()
network.predict_proba(givens)
```

We need to load the trained probabilities, create a new network that is as long as it needs to be and then give it the givens to get a probability to sample from.

# Part 3 Getting this live

## **Machine Learning Models**

In general, machine learning models are lambda functions. They are mostly stateless and merely predict an output based on an input.

Since I am sampling my function won't be a pure function since I need some form of state to use a random number generator. If we omit this detail then we can regard my sampling function as pure enough.

## **General ML in Productio**



# **Machine Learning Model**

Once you realize this, you can already recognize that some serverless setup with just functions may be good enough for what I want to do.

## ML $\propto \lambda$

This applies to most trained machine learning models but not all. Some machine learning models learn via streaming data in, these would be an acception.



## **Super Naive Implementatio**

- Python to the rescue.
- import markovify

poke\_names = [<all\_pokemon\_names>]

def lambda\_handler(event, context): poke\_names = [list(\_) for \_ in poke\_names] text\_model = markovify.Chain(poke\_names, state\_size=3) return ''.join(text\_model.walk())

## **Super Naive Implementatio**

This is a **very** naive implementation.

In the function call I am both training and then taking a single sample. This is inefficient but it seemed fine to try with a corpus that is only 750 words long.

## **Super Naive Implementatio**

Implementing this in AWS lambda is super easy. You just need to make sure that all dependencies are installed and that you zip everything together.

I have only one dependency, so;

pip install markovify -t /some/path/markov

Then I place the aforementioned code into a file named lambda\_function.py. lthen zip up /some/path/markov

55



# Button Clicking First in AWS Lambd





## Configure function

A Lambda function consists of the custom code you want to execute. Learn more about Lambda functions.

Name*	pokemon-python
Description	super-useful-task-pokemon
Runtime*	Python 2.7

### Lambda function code

Provide the code for your function. Use the editor if your code does not require custom libraries (other than boto3). If you need custom libraries, you can upload your code and libraries as a .ZIP file.

Code entry type	Upload a .ZIP file
Function package*	1 Upload
	For files larger than 10 MB, consider uploading via S3.



# Button Clicking Now in API Gateway





Amazon API Gateway



### Method Request

Auth: NONE

ARN: arn:aws:execute-api:eu-west-

1:829902169525:dyccxmwpz9/\*/GET/j

Method Response

HTTP Status: 200

**Models:** application/json => Empty

## What do I have?

- Supposedly 1 million requests for free.
- Latency of 200ms when hot. More like 1s when cold.
- Only 20 cents for every next million.
- No real scaling concerns
- Easy to update, but immature architecture. •
- Modest cloud lockin.

## **Quick cost analysis**

## \$10 flask server $\rightarrow$ 100 rq/s

Let's compare this to AWS lambda.



I'm not so sure how much I like the logging/live debugging. This is still an issue that requires some engineering. Probably you want to push to elasticsearch/kibana and probably you'll still want something like sentry to keep an eye on the front end.

It is a cloud service it can go down.

S3 was briefly down a few months ago. It pulled down my blog but also other websites like http://isitdowntoday.com/.

S3 was briefly down a few months ago. It pulled down my blog but also other websites like http://isitdowntoday.com/.

Internet responded via http://isisitdowntodaydowntoday.com.

S3 was briefly down a few months ago. It pulled down my blog but also other websites like http://isitdowntoday.com/.

Internet responded via http://isisitdowntodaydowntoday.com.

This website is now down.

So you've put your function on the internet? **THINK SECURITY!** 

You still need to worry.

You need to *remember* to set CORS headers. Otherwise you might be able to reach the endpoint from outside of your website.

You need to still think about auth.

It may be a good idea to add cloudformation.

Suppose you get a DDOS attack.

In a server situation your server would go down.

In a lambda situation amazon will horizontally scale on your behalf.

Suppose you get a DDOS attack.

In a server situation your server would go down.

In a lambda situation amazon will horizontally scale on your behalf.

Monotoring makes sense but you may also want to look at throttling some traffic. There are features for this on AWS but you need to do read up on this.

## **Quick DDOS cost analysis**

When am I powned for \$1000?



When n = 3800 you'll get near. If you have an enemy out there, this is double with a few servers.



# Part 3 **Mature ML/AWS Lambda Flow**


## Now to improve

- Think about the future
- Allow for better python models
- Automate this workflow

The simple function that is live now is doing training per request! You as a user do not notice but it is suboptimal.

Also, current this is a request that has no input.

### **Hot Cold**

It may be smart to add more resources to the function.

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven

.

### JFF!

### ς, iN

### Api Design

Vincent D. Warmerdam - @fishnets88 - koaning.io - GoDataDriven

### \* base

⇒ resp.

⇒ resp

→ resp

### Flow



.





### **Steps**



### **Steps**

You'll also need to integrate this with a machine learning pipeline. Thusfar I've found that automating this with a custom command line app works the best for me.

> tnaas-learn config

project root path: /Users/code/Development/tnaas-learn datasets folder: /Users/code/Development/tnaas-learn/datasets model output folder: /Users/code/Development/tnaas-learn/model-json

### I like testable automation.

> tnaas-learn generate-all-models GENERATE ALL overwrite=True will remove old files in /Users/code/Development/tnaas-learn/model-json MARKOV-2 pokemon : started! MARKOV-2 pokemon : model file will be /path/model-json/markov2/pokemon.json MARKOV-2 pokemon : learning from 798 names MARKOV-2 pokemon : succes new file has st\_size 62.024K MARKOV-3 pokemon : started! MARKOV-3 pokemon : model file will be /path/model-json/markov3/pokemon.json MARKOV-3 pokemon : learning from 798 names MARKOV-3 pokemon : succes new file has st\_size 2195.031K

I like testable automation.

> tnaas-learn deploy-all-models MARKOV-2 pokemon deployed! MARKOV-3 pokemon deployed! you have deployed!

This can all go to **git** and is as vendor independant as I can get it.



Pomegrenate has a bunch of dependencies that are nasty, the main culprit being numpy. It has BLAS and all sorts of low level stuff that can be machine specific and AWS runs functions on their AMI.

Quick trick; prepare the zip file via docker!

See https://github.com/ryansb/sklearn-build-lambda





82

Step one; edit the 'build.sh' file.

```
do_pip () {
    pip install --upgrade pip wheel
    pip install --use-wheel --no-binary numpy numpy
    pip install --use-wheel --no-binary cython cython
    pip install --use-wheel --no-binary scipy scipy
    pip install --use-wheel networkx
    pip install --use-wheel joblib
    pip install --use-wheel pomegranate
}
```



Step two; generate zip file.

\$ docker pull amazonlinux:2016.09

\$ docker run -v \$(pwd):/outputs -it amazonlinux:2016.09 \ /bin/bash /outputs/build.sh



### Automate!

### .yml all the things with cloudformation!

aws cloudformation package --template-file poke-mod-adv.yml --s3-bucket lambda-intermediate --output-template-file poke-template.yml

aws cloudformation deploy
 --template-file /somepath/poke-template.yml
 --stack-name pokemon-tut

All this clicking stinks and cloudformation can be tricky. There are easier methods.

There are other methods of automating.

- codestar
- chalice •
- zappa

There are other cloud services out there.

- google. later!
- azure, meh.

I have a personal history of **megahard** issues with **microsoft**. I applaud the more open source direction they are taking now but I still remember their attitude in the 90s. It will take a lot of time and effort to convince me.

### Conclusion

These lambda cloud services seem cool. Even when you're not a devops/cloud kind of person. I may also start using AWS lambda as an endpoint for user tracking instead of sending it to google analytics. Requires thought.

There is a bit of lock in but I imagine every cloud vendor to have an offering that does something similar. My gut tells me to go to either AWS or wait for google cloud.

### Conclusion

You could also just provide a server. Just sayin'.

Probibalistic graphical models rock though! Please try tangible models and heuristics before you do deep learning.

They are easier to design, debug and deploy.

These are things that I care for.