

General Outline

- Web app
 - Frontend in Elm
 - Backend in Node.js
 - Through Websockets
 - Al part in Python
 - Used by Node.js when replying to messages.
- Intelligent application in
 - Natural Language Processing

Representation of holiday locations

- Each location contains
 - The name of the location
 - Generic data about the location
 - And, relevant for the Intelligent part
 - Sets of tags with a corresponding degree of certainty.
 - The degree can be thought of as
 - "We've asked 1000 tourists if they consider the location to be X."
 - X is a trait, such as "noisy", "expensive", "near the sea".

Process

- Obtain the user's sentence
- Somehow parse information out of it
- Use the obtained information to make suggestions

Agent 001

- Has three extraction approaches, each imposing more rigidity but allowing more precise suggestions
- Relies on
 - Natural Language Toolkit (https://www.nltk.org/)
- Stanford CoreNLP
 (https://stanfordnlp.github.io/CoreNLP/)

Approach 1 - flat tags

- Segment input into words
- POS tag
- Lemmatize
- Extract words of interest

Approach 1 - Flat tags

- Upsides
 - Conveys information even if input is chaotic
- Downsides
 - Hard to express very specific preferences

Approach 2 - ranked tags

- Parse into sentence tree
- POS-tag
- Lemmatize
- Extract rules from a supposed sentence structure

Approach 2 - ranked tags

- Upsides
 - Very easy to express complex rules
 - Can be really specific about holiday locations
- Downsides
 - Very opinionated about sentence structure
 - Incomprehensible sentence parts yield useless rules

Approach 3 tags with dependency graph

- Parse sentence into dependency graph
- Traverse the nodes of the graph
 - For each node, remember the lemmatized form of the word
 - If the word is modified by a negation, negate the extracted rank.

Approach 3 tags with dependency graph

- Upsides
 - Conveys information even if input is very chaotic
 - Can express negation of holiday characteristics
- Downsides
 - Cannot express more complex rules

Agent 002

- Classifies a query based on the intent it represents
- When the intent is confirmed a predefined message can be sent

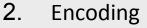






Process

- 1. Data preparation
 - 1.1. Data cleaning
 - 1.2. Lemmatization



- 2.1. Input Encoding
- 2.2 Output Encoding
- 3. Train and validation set
- 4. Results







1. Data preparation

Simplification of data by:

- Building a set of unique intents
- Tagging sentences with their corresponding intents
- Building dataframes for sentences and intents

| Sentences | Intents |
|---------------------------|-----------------|
| Yo, dawg, what's up | Greet |
| I want to go on a trip | Vacation_Search |
| My last vacation was shit | Vacation_Review |
| Show me what you got | Vacation_List |

1.1. Data cleaning

- Remove punctuations and special characters
- Break sentences into words
- Lowercase words and apply lemmatization

Help me with my vacation plan, please.



me

with

my

vacation

please

1.2. Lemmatization

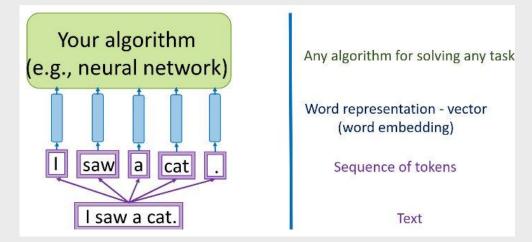
NLTK lemmatizer

- easy to use python library
- large, freely and publicly available english lexical database
- based on WordNet's built-in morphy function

| Pos | Suffix | Ending |
|-----------|--------|--------|
| Noun | "s" | nn |
| Adjective | "est" | "e" |
| Verb | "ies" | "y" |

2. Encoding

- Process of turning a set of categorical features in raw (or preprocessed) text to a series of vectors
- 2 phases: Input and Output Encoding

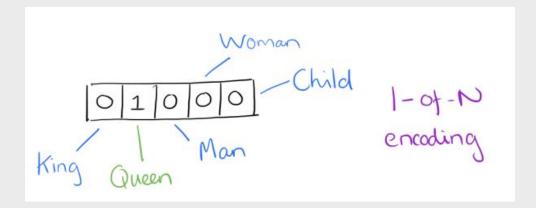


2.1 Input encoding

- Cleaned data is passed to the Keras tokenizer
- The use of padding to make tokenized words the same length

2.2 Output encoding

- Apply the same keras tokenize class to the output intents
- One hot encoding of the intents



3. Train and validation set



- Split the dataset into training and validation set (80% training and 20% validation set)
- Build the model and train it on the data
- Bidirectional Gated Recurrent Unit neural network

120 epochs, batch size of 32



The food on my last vacation trip has been amazing

4. Results

- Achieved 75% training accuracy and 70% validation accuracy
- Some drawbacks: lack of data, hyperparameters not fully optimized
- Confusion due to similarity of intents between vacation_search and vacation_list

Agent 001 + Agent 002

- We first find the intent of the user's sentence (agent 002)
- We then extract the tags (agent 001)
- If the user's intent is to search vacation we use the tags to recommend a place from the database

Demo time!