

# FAQ CHATBOT

MIDSEM PRESENTATION

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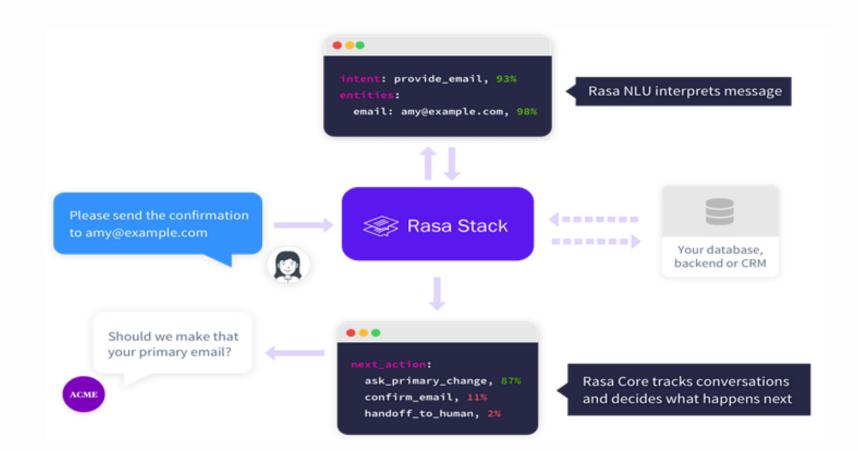
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## ABOUT RASA

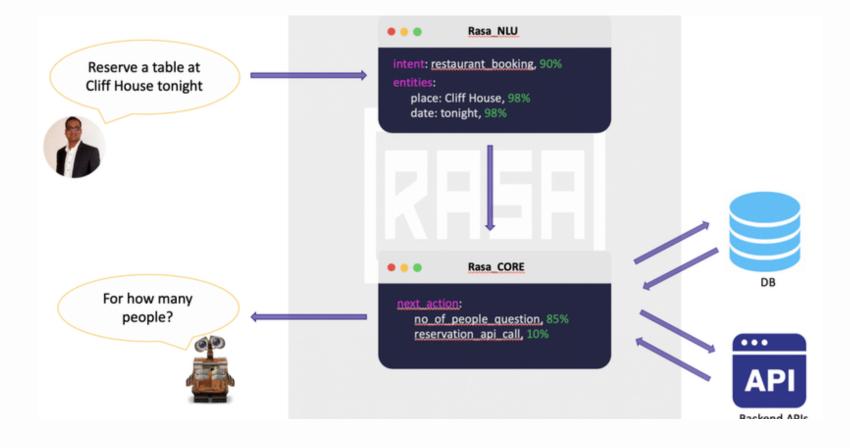


Rasa is a framework for bootstrapping coversational chatbots.

It has two components:

- Rasa NLU
- Rasa Core

### RASA NLU

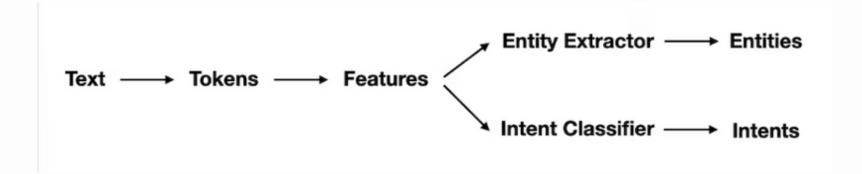


It is a natural language understanding tool for intent classification and entity extraction in chatbots.

The NLU pipeline is defined in the `config.yml` file in Rasa. This file describes all the steps in the pipeline that will be used by Rasa to detect intents and entities.

There are different types of components that you can expect to find in a pipeline. The main ones are:

- Tokenizers
- Featurizers
- Intent Classifiers
- Entity Extractors



### CONFIG.YML

```
language: "en_core_web_md"
pipeline:
 name: "SpacyNLP"
                                     # loads the spacy language model
       "SpacyTokenizer"
                                     # splits the sentence into tokens
 name: "CRFEntityExtractor"
                                     # uses the pretrained spacy NER model
 name: "SpacyFeaturizer"
                              # transform the sentence into a vector representation
 name: "SklearnIntentClassifier"
                                    # uses the vector representation to classify using SVM
 name: "EntitySynonymMapper"
                                             # trains the synonyms
policies:
 name: MemoizationPolicy
 name: TEDPolicy
 max history: 5
 epochs: 10
```

### **SpacyTokenizer**

 The tokenizer splits each individual word in the utterance into a separate token, and commonly the output of the tokenizer is a list of words.

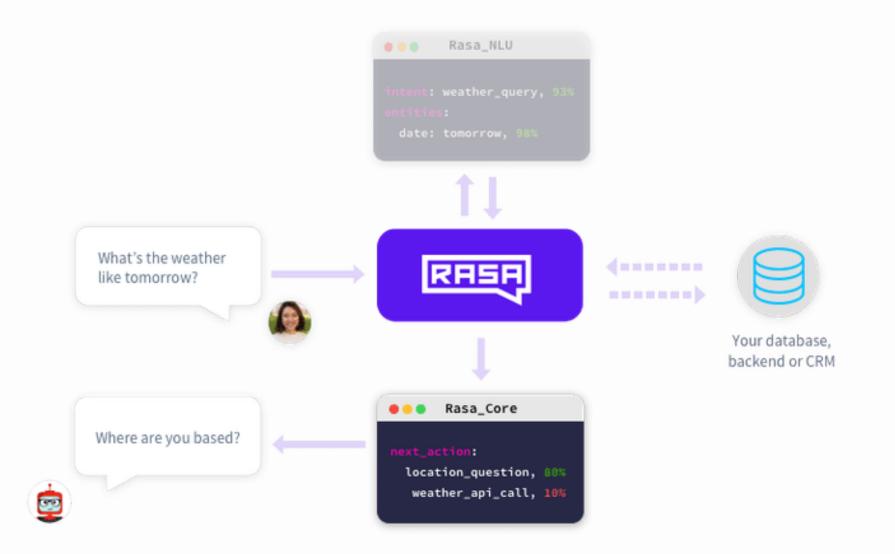
#### **SpacyFeaturizer**

- Featurizers generate numeric features for machine learning models.
- There are two types of features:
  - Sparse Features
  - Dense Features

#### SklearnIntentClassifier

- Once we've generated features for all of the tokens and for the entire sentence, we can pass it to an intent classification model.
- DIET model is used which can handle both intent classification and entity extraction.

## RASA CORE



Rasa Core's job is to decide what should happen next during the course of a dialogue

it uses the files,

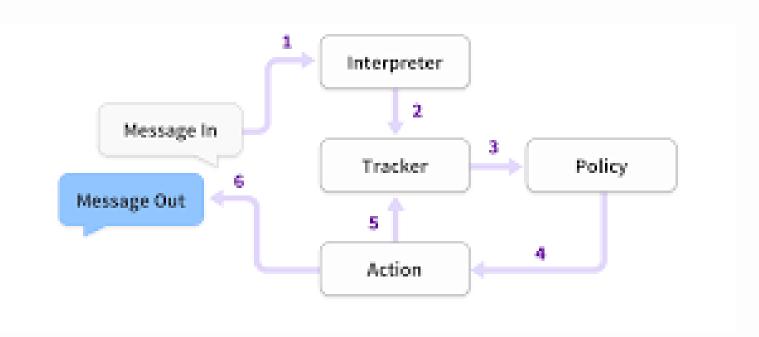
- stories
- domain.yml
- dialogue model

## RASA\_CORE POLICIES

RASA uses policies to decide which action to take at each step in a conversation. There are machine-learning and rule-based policies that your assistant can use in tandem

- MemoizationPolicy
- TEDPolicy
- RulePolicy

AT every turn, each policy defined in your configuration will predict a next action with a certain confidence level. Prediction having max confidence and high priority gets selected



### MemoizationPolicy

- remembers the stories from your training data.
- if the current conversation matches the stories in your stories.yml file.
- it will predict the next action from the matching stories of your training data with a confidence of 1.0

# **TED** (transformer embedding dialogue policy)

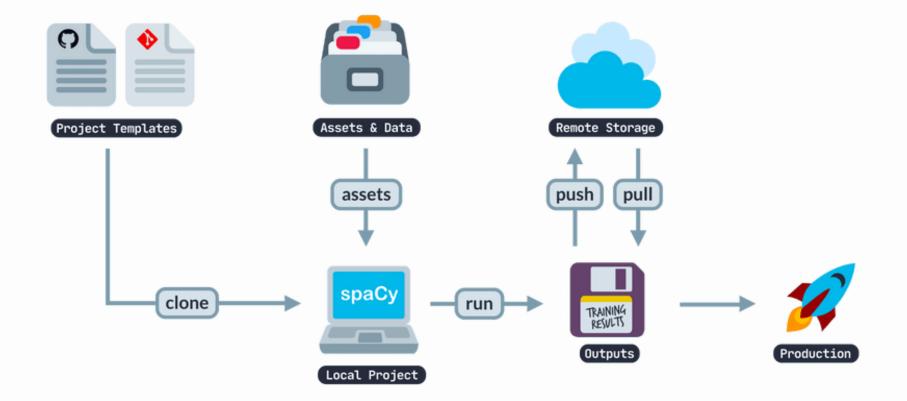
It is a multi-task architecture for next action prediction and entity recognition. The architecture consists of several transformer encoders which are shared for both tasks.

### RulePolicy

The RulePolicy is a policy that handles conversation parts that follow a fixed behavior (e.g. business logic).

It makes predictions based on any rules you have in your training data

### TRAINING DATA



Our chatbot utilizes a pre-trained model extracted from spaCy library. The language module for the training is "en\_core\_web\_md" which helps in efficiently determining entities and intents using Natural Language Understanding.

The model is fine-tuned using custom training data which is defined in "nlu.md" and "domain.yml" files.

### NLU.MD

```
## intent: info_timings
   - what are the timings of the temple?
   - when can we go to the temple?
   - when will the temple be open?
   - when should I visit the temple?
   - when to visit the temple?
   - when to plan visit to temple?
   - when does the temple close?
   - what is closing time of temple?
## intent: significance
 - what is significance?
 - what does the temple signify?
 - what is special about this temple?
 - why was the temple built?
 - what does the temple stand for?
## intent: reachability
 - how to reach this temple
 - what is nearest airport?
 - what is nearest railway station?
 - how do I reach the temple
 - how to plan logistics to reach the temple
 - how to travel to this temple
  - best route to reach this temple
```

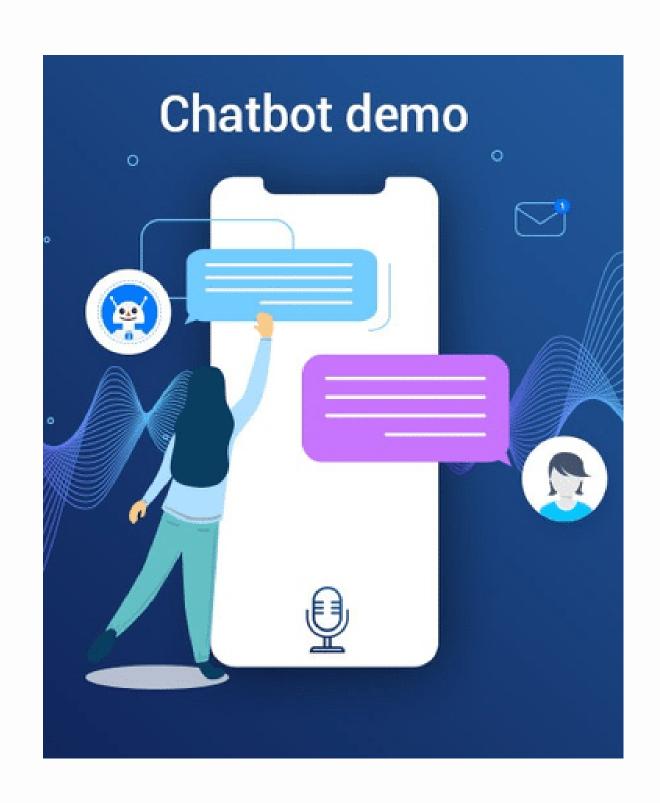
We have fine-tuned our model using the most common intents for our specific use case

- intents : corresponds to respective purpose which user is trying to execute, identified by our model
- followed by the training examples we provided

### DOMAIN.YML

```
templates:
 utter_greet:
 - text: "Hey! How are you?"
 - text: "Hey there! How can I help you, today?"
 - text: "Hello! I am here to answer your questions regarding Saraswati Temple."
 - text: "Hola! How can I help?"
 utter did that help:
 - text: "Did that help you?"
 utter goodbye:
 - text: "Bye"
 - text: "Goodbye!"
 - text: "I hope I solved your queries. Adios!"
 - text: "See ya there!"
 utter iamabot:
 - text: "I am a bot, powered by Rasa."
 utter location:
 - text: "The temple is located inside Bits Pilani premises infront of T-lawns."
 utter reachability:
 - text: "Choose from where you want to visit: Jaipur and Delhi"
 utter_info_timings:
 - text: "The temple is open everyday from morning 8am to evening 7pm (except for special holidays). "
```

Domain.yml file contains the declarations and responses corresponding to every intent defined in the nlu.md file. This mapping is used to return the correct response from the chatbot after it recognizes the intent of the user input.



### DEMONSTRATION

basic chatbot link:

<u>https://colab.research.google.com/drive/1pTuJysl3lXaeKJHxND</u>
<u>ULzCoCuE2CE-7b#scrollTo=PyQVtGjWUxdK</u>

multilingual chatbot link:

https://colab.research.google.com/drive/1UBQ4IjMReBp0jJIKuA9 62XBRItV\_P1Nw#scrollTo=GfWnPccZdtC7

# QUESTIONS

