# FLY ME

# FLIGHTS BOOKING CHATBOT

#### Challenge

**Fly Me** is a travel agency offering *flight booking services* for private or professional customers.

In order to make the booking process easier, we want to automate it with a **chatbot**. To be efficient, the bot must be able to :

- interact with the user in a natural way
- recognize the user's **intent** (book a flight, ...)
- recognize the desired caracteristics of the trip (origin and destination cities, ...)
- ask for more information if necessary
- allow the user to cancel the booking process
- raise an **alert** if the booking process fails too often

#### Goals

In this project, we are going to:

- deploy a chatbot in the cloud
- train a language understanding model
  - o in order to *infer* the user's intent and trip caracteristics from the user's input
- implement unit and integration tests
  - to ensure that the chatbot is working as expected
- monitor the chatbot's performance
  - raise an *alert* in case of recurring issues

## **Exploratory Data Analysis**

For this MVP, we used the Azure Frames Dataset:

- 1369 dialogues between a human and a bot, composed of ~15 turns on average
- the bot can:
  - o ask the user for information
  - suggest a trip to the user
  - ask the user to confirm
- the user can:
  - ask the bot for suggestions
  - inform the bot about relevant information
  - cancel or confirm the trip

#### Dialog Example

```
0 - user says :
"Can you get me to Kyoto"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto'}
1 - wizard says :
"Ok! From where?"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto'}
2 - user says :
"I need to be there for at least four days"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4'}
3 - wizard says :
"Have you a budget?"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4'}
4 - user says :
"I'm on the road so I can head there from any origin point. Budget is 3500"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4', 'or_city': '-1', 'budget': '3500.0'}
5 - wizard says :
"Travelling alone?"
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4', 'or_city': '-1', 'budget': '3500.0'}
6 - user says :
"two adults. oh and please find me a place near a park"
Known facts:
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4', 'or_city': '-1', 'budget': '3500.0', 'park': True, 'n_adults': '2'}
7 - wizard says :
"I can get you 4 days in Kyoto for 1857.63USD if you leave from Sapporo."
Known facts :
{'intent': 'book', 'dst_city': 'Kyoto', 'min_duration': '4', 'or_city': 'Sapporo', 'budget': '3500.0', 'park': True, 'n_adults': '2', 'duration': '4', 'price': '1857.63'}
```

# Step 1: LUIS Model Training And Deployment

The first step in this project was to train and deploy a **language understanding** model using the LUIS *natural language service*:

- create an Azure LUIS resource
  - add the intents and entities to the LUIS model
- format the raw data to be compatible with Azure LUIS
  - add the examples to the LUIS model
- run the LUIS model training
- deploy the model to the *cloud*

## Step 2: Chatbot Development And Deployment

The second step in this project was to develop and deploy a **Chatbot**:

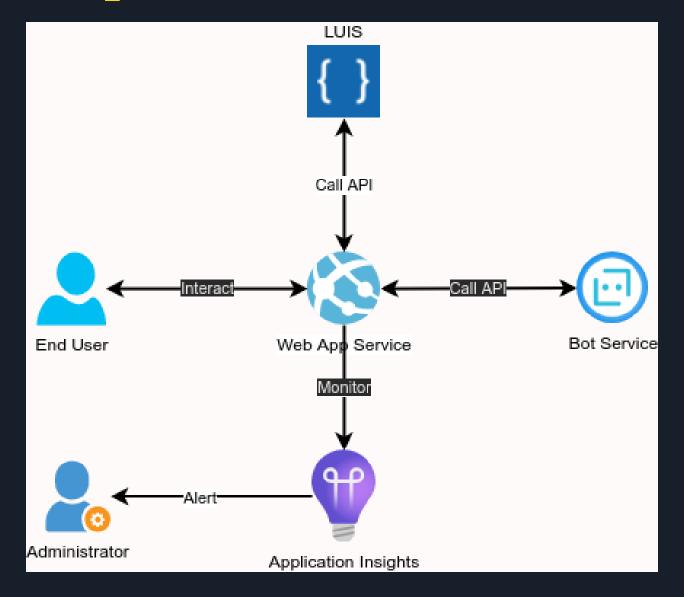
- create a **Bot Service** resource
- implement the bot with the Bot Framework SDK
  - o implement dialogs: welcome, gather intention, ask informations, confirm, cancel, ...
  - integrate the LUIS service
  - add unit and integration tests
    - test the LUIS service
    - test the LUIS service integration
    - test a standard dialog
- deploy the bot to the Azure Web App

# Step 3: Chatbot Monitoring

The final step in this project was to monitor the **chatbot's performance**:

- create a Application Insights resource
- log a Telemetry trace every time the bot is used (booking\_accepted or booking\_refused)
- add an Alert if the bot fails too often:
  - if More than 5 Booking Refused last 5 minutes
  - then Send Email and SMS to admins

# **Curent MVP System Architecture**

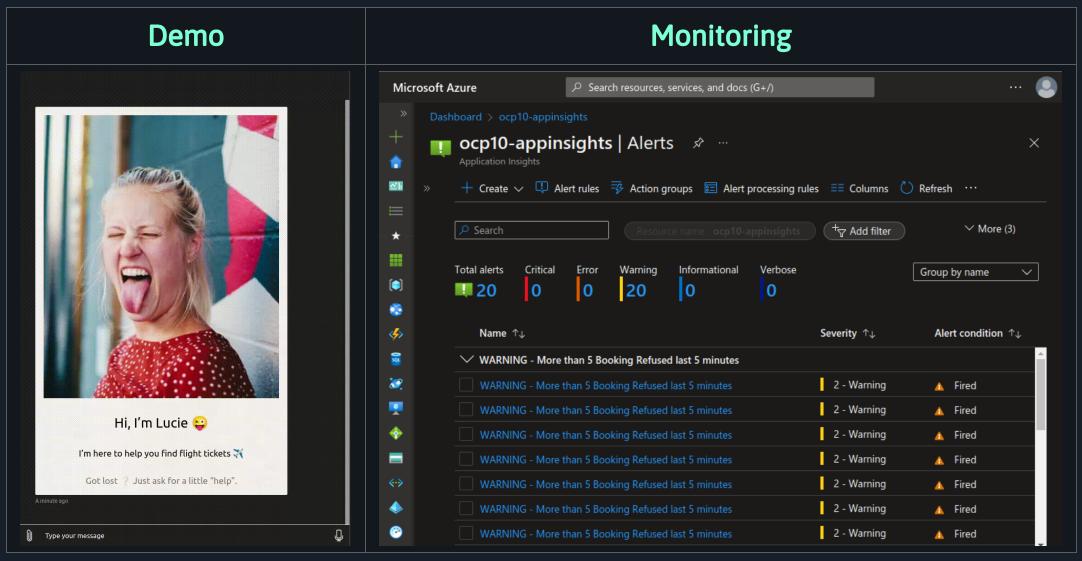


# **Resources Inventory**

| LUIS                         | Chatbot          | Monitoring           |
|------------------------------|------------------|----------------------|
| Language understanding model | Bot service      | Application Insights |
| Authoring resource           | App Service Plan | Alert Rule           |
| Prediction resource          | App Service      | Monitoring Dashboard |



#### Demo

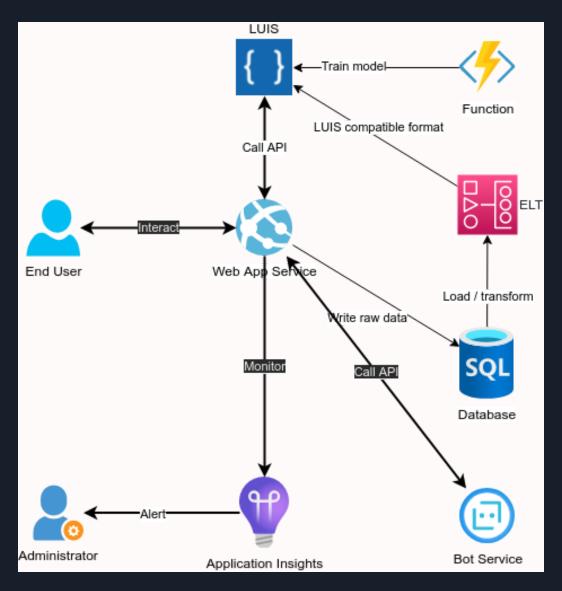


### Target Performance Management Policy

In order to achieve the target performance, the following policy must be implemented:

- **store** each dialogs and inferred *intents* and *entities* in a *database*
- transform and load (ETL) the raw data in a datawarehouse in a format compatible with the LUIS service
- daily re-train the language understanding model with the new intents and entities of the successful bookings

# **Target Production Architecture**



## **Next Steps**

- integrate the bot with multiple *Channels* (Website, Discord, Teams, Slack, ...)
- **improve** the bot capacity to handle more *Intentions* and *Entities*
- connect the bot to an actual *Flight booking* system
- monitor more precisely the bot's performance : errors, performance, availability, ...
- implement the model continuous training and deployment

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