

Build a Virtual Assistant Using Jarvis, Cloud Sync, and NeMo

NetApp Solutions

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Overview

This section provides detail on the implementation of the virtual retail assistant.

Next: Jarvis Deployment

Jarvis Deployment

You can sign up for Jarvis Early Access program to gain access to Jarvis containers on NVIDIA GPU Cloud (NGC). After receiving credentials from NVIDIA, you can deploy Jarvis using the following steps:

- 1. Sign-on to NGC.
- 2. Set your organization on NGC: ea-2-jarvis.
- Locate Jarvis EA v0.2 assets: Jarvis containers are in Private Registry > Organization Containers.
- 4. Select Jarvis: navigate to Model Scripts and click Jarvis Quick Start
- 5. Verify that all assets are working properly.
- 6. Find the documentation to build your own applications: PDFs can be found in Model Scripts > Jarvis Documentation > File Browser.

Next: Customize States and Flows for Retail Use Case

Customize States and Flows for Retail Use Case

You can customize States and Flows of Dialog Manager for your specific use cases. In our retail example, we have the following four yaml files to direct the conversation according to different intents.

Se the following list of file names and description of each file:

- main_flow.yml: Defines the main conversation flows and states and directs the flow to the other three
 yaml files when necessary.
- retail_flow.yml: Contains states related to retail or points-of-interest questions. The system either provides the information of the nearest store, or the price of a given item.
- weather_flow.yml: Contains states related to weather questions. If the location cannot be determined, the system asks a follow up question to clarify.
- error_flow.yml: Handles cases where user intents do not fall into the above three yaml files. After
 displaying an error message, the system re-routes back to accepting user questions. The following sections
 contain the detailed definitions for these yaml files.

main_flow.yml

name: JarvisRetail
intent transitions:

```
jarvis error: error
  price check: retail price check
  inventory check: retail inventory check
  store location: retail store location
  weather.weather: weather
  weather.temperature: temperature
  weather.sunny: sunny
  weather.cloudy: cloudy
  weather.snow: snow
  weather.rainfall: rain
  weather.snow yes no: snowfall
 weather.rainfall yes no: rainfall
 weather.temperature yes no: tempyesno
 weather.humidity: humidity
 weather.humidity yes no: humidity
 navigation.startnavigationpoi: retail # Transitions should be context
and slot based. Redirecting for now.
 navigation.geteta: retail
 navigation.showdirection: retail
 navigation.showmappoi: idk what you talkin about
  nomatch.none: idk what you talkin about
states:
 init:
   type: message text
   properties:
      text: "Hi, welcome to NARA retail and weather service. How can I
help you?"
 input intent:
   type: input context
   properties:
      nlp type: jarvis
      entities:
        intent: dontcare
# This state is executed if the intent was not understood
 dont get the intent:
    type: message text random
   properties:
      responses:
        - "Sorry I didn't get that! Please come again."
        - "I beg your pardon! Say that again?"
        - "Are we talking about weather? What would you like to know?"
        - "Sorry I know only about the weather"
        - "You can ask me about the weather, the rainfall, the
temperature, I don't know much more"
      delay: 0
    transitions:
```

```
next_state: input_intent
  idk what you talkin about:
    type: message text random
   properties:
      responses:
        - "Sorry I didn't get that! Please come again."
        - "I beg your pardon! Say that again?"
        - "Are we talking about retail or weather? What would you like to
know?"
        - "Sorry I know only about retail and the weather"
        - "You can ask me about retail information or the weather, the
rainfall, the temperature. I don't know much more."
      delay: 0
    transitions:
      next state: input intent
 error:
   type: change context
   properties:
        update keys:
           intent: 'error'
    transitions:
        flow: error flow
 retail inventory check:
   type: change context
   properties:
        update keys:
           intent: 'retail_inventory_check'
    transitions:
        flow: retail flow
 retail price check:
   type: change context
   properties:
        update keys:
           intent: 'check_item_price'
    transitions:
        flow: retail flow
 retail store location:
   type: change_context
   properties:
        update keys:
           intent: 'find the store'
    transitions:
        flow: retail flow
 weather:
    type: change context
   properties:
```

```
update_keys:
         intent: 'weather'
  transitions:
      flow: weather flow
temperature:
  type: change context
 properties:
      update keys:
         intent: 'temperature'
  transitions:
      flow: weather flow
rainfall:
  type: change_context
 properties:
      update keys:
         intent: 'rainfall'
  transitions:
      flow: weather_flow
sunny:
  type: change context
 properties:
      update_keys:
         intent: 'sunny'
  transitions:
      flow: weather flow
cloudy:
  type: change_context
  properties:
      update keys:
         intent: 'cloudy'
  transitions:
      flow: weather flow
snow:
 type: change_context
 properties:
      update keys:
         intent: 'snow'
  transitions:
      flow: weather flow
rain:
  type: change context
  properties:
      update keys:
        intent: 'rain'
  transitions:
      flow: weather flow
```

```
snowfall:
    type: change context
    properties:
        update keys:
           intent: 'snowfall'
    transitions:
        flow: weather flow
tempyesno:
    type: change context
    properties:
        update keys:
           intent: 'tempyesno'
    transitions:
        flow: weather flow
humidity:
    type: change context
    properties:
        update keys:
           intent: 'humidity'
    transitions:
        flow: weather flow
end_state:
  type: reset
  transitions:
    next state: init
```

retail_flow.yml

```
name: retail flow
states:
 store location:
   type: conditional exists
   properties:
     key: '{{location}}'
   transitions:
      exists: retail state
      notexists: ask retail location
 retail state:
   type: Retail
   properties:
    transitions:
      next state: output retail
 output retail:
      type: message text
      properties:
```

```
text: '{{retail status}}'
    transitions:
      next state: input intent
ask retail location:
  type: message_text
 properties:
    text: "For which location? I can find the closest store near you."
  transitions:
    next state: input retail location
input retail location:
  type: input user
 properties:
    nlp type: jarvis
   entities:
      slot: location
    require match: true
  transitions:
    match: retail state
    notmatch: check retail jarvis error
output retail acknowledge:
  type: message text random
 properties:
   responses:
      - 'ok in {{location}}'
      - 'the store in {{location}}'
      - 'I always wanted to shop in {{location}}'
    delay: 0
  transitions:
    next state: retail state
output retail notlocation:
 type: message text
 properties:
    text: "I did not understand the location. Can you please repeat?"
  transitions:
    next state: input intent
check rerail jarvis error:
 type: conditional exists
 properties:
    key: '{{jarvis error}}'
  transitions:
    exists: show retail jarvis api error
    notexists: output retail notlocation
show retail jarvis api error:
 type: message text
 properties:
   text: "I am having troubled understanding right now. Come again on
```

```
that?"
    transitions:
    next_state: input_intent
```

weather_flow.yml

```
name: weather flow
states:
 check weather location:
    type: conditional exists
   properties:
     key: '{{location}}'
   transitions:
      exists: weather_state
      notexists: ask weather location
 weather state:
   type: Weather
   properties:
   transitions:
      next_state: output_weather
 output weather:
     type: message text
      properties:
       text: '{{weather status}}'
      transitions:
        next state: input intent
 ask weather location:
   type: message text
   properties:
      text: "For which location?"
    transitions:
      next_state: input_weather_location
 input weather location:
   type: input user
   properties:
     nlp type: jarvis
      entities:
        slot: location
      require match: true
    transitions:
      match: weather state
      notmatch: check jarvis error
 output weather acknowledge:
    type: message text random
   properties:
```

```
responses:
       - 'ok in {{location}}'
       - 'the weather in {{location}}'
       - 'I always wanted to go in {{location}}'
     delay: 0
   transitions:
     next state: weather state
 output weather notlocation:
   type: message text
   properties:
     text: "I did not understand the location, can you please repeat?"
   transitions:
     next state: input intent
 check jarvis_error:
   type: conditional exists
   properties:
     key: '{{jarvis error}}'
   transitions:
     exists: show_jarvis_api_error
     notexists: output weather notlocation
 show jarvis api error:
   type: message_text
   properties:
     text: "I am having troubled understanding right now. Come again on
that, else check jarvis services?"
   transitions:
     next_state: input_intent
```

error_flow.yml

```
name: error flow
states:
 error state:
   type: message text random
   properties:
      responses:
        - "Sorry I didn't get that!"
        - "Are we talking about retail or weather? What would you like to
know?"
        - "Sorry I know only about retail information or the weather"
        - "You can ask me about retail information or the weather, the
rainfall, the temperature. I don't know much more"
        - "Let's talk about retail or the weather!"
      delay: 0
    transitions:
      next state: input intent
```

Next: Connect to Third-Party APIs as Fulfillment Engine

Connect to Third-Party APIs as Fulfillment Engine

We connected the following third-party APIs as a Fulfillment Engine to answer questions:

- WeatherStack API: returns weather, temperature, rainfall, and snow in a given location.
- Yelp Fusion API: returns the nearest store information in a given location.
- eBay Python SDK: returns the price of a given item.

Next: NetApp Retail Assistant Demonstration

NetApp Retail Assistant Demonstration

We recorded a demonstration video of NetApp Retail Assistant (NARA). Click this link to open the following figure and play the video demonstration.



Next: Use NetApp Cloud Sync to Archive Conversation History

Use NetApp Cloud Sync to Archive Conversation History

By dumping conversation history into a CSV file once a day, we can then leverage Cloud Sync to download the log files into local storage. The following figure shows the architecture of having Jarvis deployed on-premises and in public clouds, while using Cloud Sync to send conversation history for NeMo training. Details of NeMo training can be found in the section Expand Intent Models Using NeMo Training.



Next: Expand Intent Models Using NeMo Training

Expand Intent Models Using NeMo Training

NVIDIA NeMo is a toolkit built by NVIDIA for creating conversational AI applications. This toolkit includes collections of pre-trained modules for ASR, NLP, and TTS, enabling researchers and data scientists to easily compose complex neural network architectures and put more focus on designing their own applications.

As shown in the previous example, NARA can only handle a limited type of question. This is because the pretrained NLP model only trains on these types of questions. If we want to enable NARA to handle a broader range of questions, we need to retrain it with our own datasets. Thus, here, we demonstrate how we can use NeMo to extend the NLP model to satisfy the requirements. We start by converting the log collected from NARA into the format for NeMo, and then train with the dataset to enhance the NLP model.

Model

Our goal is to enable NARA to sort the items based on user preferences. For instance, we might ask NARA to suggest the highest-rated sushi restaurant or might want NARA to look up the jeans with the lowest price. To this end, we use the intent detection and slot filling model provided in NeMo as our training model. This model allows NARA to understand the intent of searching preference.

Data Preparation

To train the model, we collect the dataset for this type of question, and convert it to the NeMo format. Here, we listed the files we use to train the model.

dict.intents.csv

This file lists all the intents we want the NeMo to understand. Here, we have two primary intents and one intent only used to categorize the questions that do not fit into any of the primary intents.

```
price_check
find_the_store
unknown
```

dict.slots.csv

This file lists all the slots we can label on our training questions.

```
B-store.type
B-store.name
B-store.status
B-store.hour.start
B-store.hour.end
B-store.hour.day
B-item.type
B-item.name
B-item.color
B-item.size
B-item.quantity
B-location
```

```
B-cost.high
B-cost.average
B-cost.low
B-time.period of time
B-rating.high
B-rating.average
B-rating.low
B-interrogative.location
B-interrogative.manner
B-interrogative.time
B-interrogative.personal
B-interrogative
B-verb
B-article
I-store.type
I-store.name
I-store.status
I-store.hour.start
I-store.hour.end
I-store.hour.day
I-item.type
I-item.name
I-item.color
I-item.size
I-item.quantity
I-location
I-cost.high
I-cost.average
I-cost.low
I-time.period of time
I-rating.high
I-rating.average
I-rating.low
I-interrogative.location
I-interrogative.manner
I-interrogative.time
I-interrogative.personal
I-interrogative
I-verb
I-article
0
```

train.tsv

This is the main training dataset. Each line starts with the question following the intent category listing in the file dict.intent.csv. The label is enumerated starting from zero.

train_slots.tsv

```
20 46 24 25 6 32 6
52 52 24 6
23 52 14 40 52 25 6 32 6
...
```

Train the Model

```
docker pull nvcr.io/nvidia/nemo:v0.10
```

We then use the following command to launch the container. In this command, we limit the container to use a single GPU (GPU ID = 1) since this is a lightweight training exercise. We also map our local workspace /workspace/nemo/ to the folder inside container /nemo.

Inside the container, if we want to start from the original pre-trained BERT model, we can use the following command to start the training procedure. data_dir is the argument to set up the path of the training data. work dir allows you to configure where you want to store the checkpoint files.

```
cd examples/nlp/intent_detection_slot_tagging/
python joint_intent_slot_with_bert.py \
    --data_dir /nemo/training_data\
    --work_dir /nemo/log
```

If we have new training datasets and want to improve the previous model, we can use the following command to continue from the point we stopped. checkpoint dir takes the path to the previous checkpoints folder.

```
cd examples/nlp/intent_detection_slot_tagging/
python joint_intent_slot_infer.py \
    --data_dir /nemo/training_data \
    --checkpoint_dir /nemo/log/2020-05-04_18-34-20/checkpoints/ \
    --eval_file_prefix test
```

Inference the Model

We need to validate the performance of the trained model after a certain number of epochs. The following command allows us to test the query one-by-one. For instance, in this command, we want to check if our

model can properly identify the intention of the query where can I get the best pasta.

```
cd examples/nlp/intent_detection_slot_tagging/
python joint_intent_slot_infer_b1.py \
--checkpoint_dir /nemo/log/2020-05-29_23-50-58/checkpoints/ \
--query "where can i get the best pasta" \
--data_dir /nemo/training_data/ \
--num_epochs=50
```

Then, the following is the output from the inference. In the output, we can see that our trained model can properly predict the intention find_the_store, and return the keywords we are interested in. With these keywords, we enable the NARA to search for what users want and do a more precise search.

```
[NeMo I 2020-05-30 00:06:54 actions:728] Evaluating batch 0 out of 1
[NeMo I 2020-05-30 00:06:55 inference utils:34] Query: where can i get the
best pasta
[NeMo I 2020-05-30 00:06:55 inference utils:36] Predicted intent:
                                                                        1
find the store
[NeMo I 2020-05-30 00:06:55 inference utils:50] where
                                                        B-
interrogative.location
[NeMo I 2020-05-30 00:06:55 inference utils:50] can
                                                        0
[NeMo I 2020-05-30 00:06:55 inference utils:50] i
                                                        0
[NeMo I 2020-05-30 00:06:55 inference utils:50] get
                                                        B-verb
[NeMo I 2020-05-30 00:06:55 inference utils:50] the
                                                        B-article
[NeMo I 2020-05-30 00:06:55 inference utils:50] best
                                                       B-rating.high
[NeMo I 2020-05-30 00:06:55 inference utils:50] pasta
                                                       B-item.type
```

Next: Conclusion

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