Inside our Al Pipeline

Putting the Natural Back Into Artificial Intelligence to Move at Light Speed

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about me











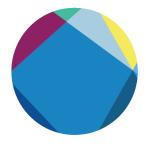




to move fast experiment



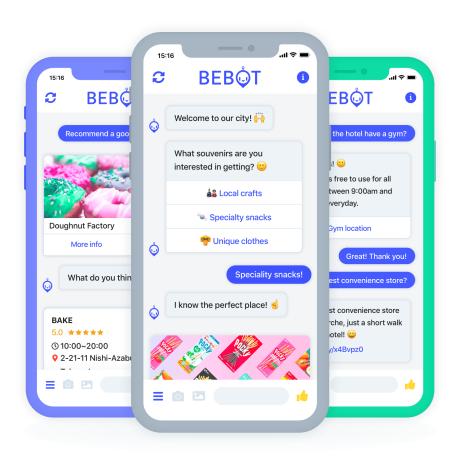
Leverage humans before coding effort

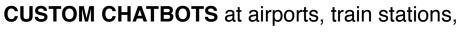


CODE HOOKS

Leverage bad code before more coding effort

Bebot by Bespoke





hotels, and cities across Japan

FAQS, RESTAURANTS, EMERGENCY INFORMATION, AND MORE

When automation isn't available, human operators respond

> 30,000 USERS EVERY DAY getting help as a tourist

Humans define example questions and answers.

Answer:

"Check out is at 9AM"

Question phrasings:

"when is check out?"

"When can I check-out?"

"How late is checkout?"





Train an ML model on the question phrasings.

Label, aka Class:

WHEN_IS_CHECK_OUT
WHERE_IS_THE_HOTEL

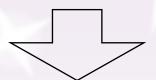
. . .

Data:

(Example questions for each label)

When a new user query arrives, predict which question it represents.

"ohi when could I
maybe check out?
Thanks :)"



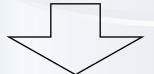
WHEN_IS_CHECK_OUT

03



Look up the pre-defined answer for that question.

WHEN_IS_CHECK_OUT



"Check out is at 9AM"

Train a machine learning model on the question phrasings.

```
Define
example
  and
```

BESPOKE

```
'Where is my hotel?',
question
                    'Hotel location?'
                  when is checkout time : [
answers
                    'When is check-out time?',
                    'When do I need to check out?'
                                            Which class is most similar?
                  where_is_the_hotel : [
                    'hotel' : 2,
                    'location': 1,
                    'when': 0
                  when is checkout time : [
                    'when': 2,
                    'check' : 2,
                    'location': 0
```

where is the hotel : [

"ohi how get 2 hotel kthx _(ツ)_/"



```
'ohi' : 1,
'how' : 1,
' - \_(ツ)_/ - ': 1
```

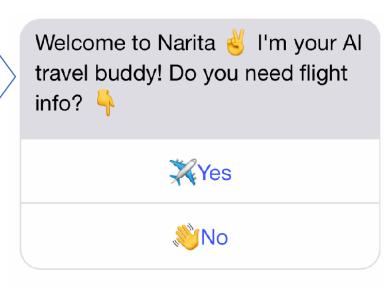
where is the hotel: 85% when is checkout time: 32%

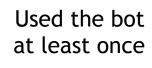
When a new user query arrives, predict which question it represents.

Look up a pre-defined answer

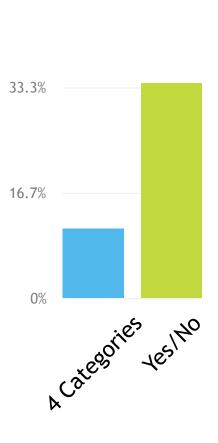
Experiment
Quickly test out an idea and see if it actually improves your product.







50%



humans



DESIGN CONVERSATIONS

Humans are good at sensical
conversation flows.



PERFORM TASKS NOT YET AUTOMATED
When considering a new feature, have
humans chat with users first.



DEFINE EXPERIMENTS

Make hypotheses about human behavior and test them.



EVALUATE AUTOMATED REPLIES

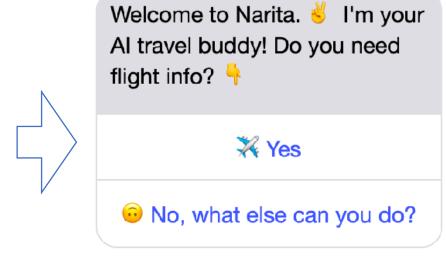
Operators give you immediate

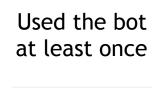
feedback on algorithm performance.

Welcome to Narita. I'm your
Al travel buddy! Do you need
flight info?

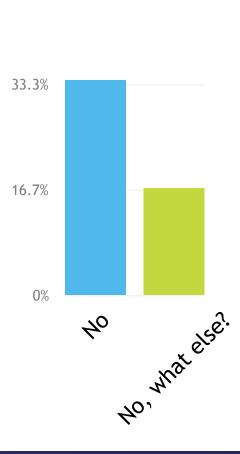
Yes

No





50%



Code hook

An integration point that allows you to insert a different code path, without altering the rest of the application.

"i want to go to a pokèmon centre"

Accuracy improvement 20%

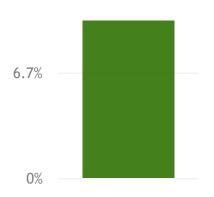
pokèmon pokemon

ok <a>\$\mathbb{G}\$ thanks!



(convert to ASCII)

ok thanks!



13.3%

BESPOKE

```
class MyApp {
   function getAnswer(String question) {
       // normal algorithm
       result = new NaiveBayesAlgorithm().predict(question);
       return result;
```

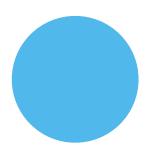
```
class MyApp {
   function getAnswer(String question) {
                                                    AWS Lambda
       if (NEW_ALGORITHM_ENABLED) {
          return new NewAlgorithm().predict(question);
      // normal algorithm
       result = new NaiveBayesAlgorithm().predict(question);
       return result;
```

code hooks



RUN EXPERIMENTAL ALGORITHMS

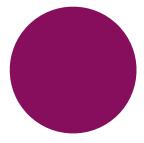
After benchmark test sets, enable in production for a short period of time.



RUN UNOPTIMIZED CODE

Ugly code? Memory-intensive code?

See if it works before improving it.



COMPARE ALGORITHMS

Log the results of both the old and the new algorithms.



USE NEW TECHNOLOGIES

Code hooks can run independently of your existing stack.

ML pipeline + humans and code hooks

To support experimentation and bring guests the most natural user experience.

136

Experiments run in 2019H1

1.13

Average experiments run per day

3

Average days between a new idea and trying it in prod



Hotel guests make reservations (via human operators)

Thanks!

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- Chatbot R&D Lead (ML/NLP) Engineers · ...and more
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