#BullhornEngage

ENGAGE

Practical Applications of AI - Real World Examples

Nathan Dickerson

Amir Kurtovic

Lukas Neumann



Nathan Dickerson Senior Developer, Bullhorn



Amir Kurtovic Senior Developer, Bullhorn



Lukas Neumann Chief Architect, Invenias



#BullhornEngage

ENGAGE

Chatbots

Nathan Dickerson

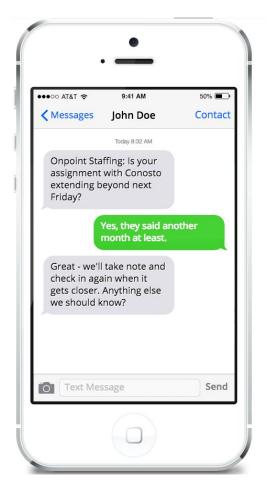
Chatbot Ideation

- Reminders
- Scheduling
- Screening
- Follow Up

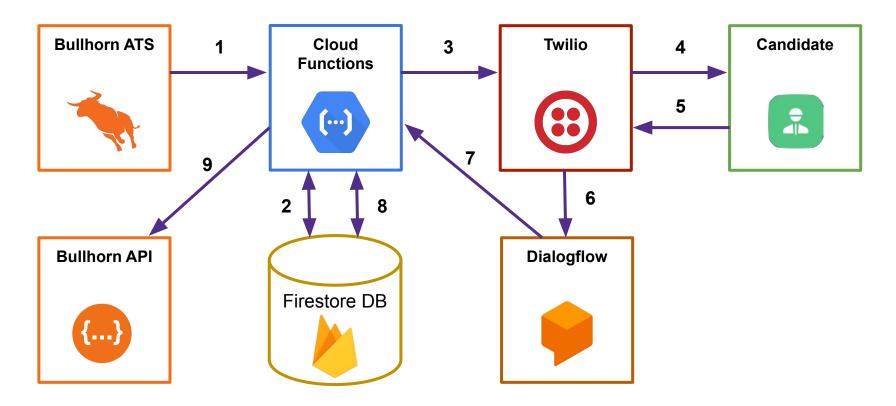
Redeployment Chatbot

Improve redeployment by fixing assignment length accuracy.

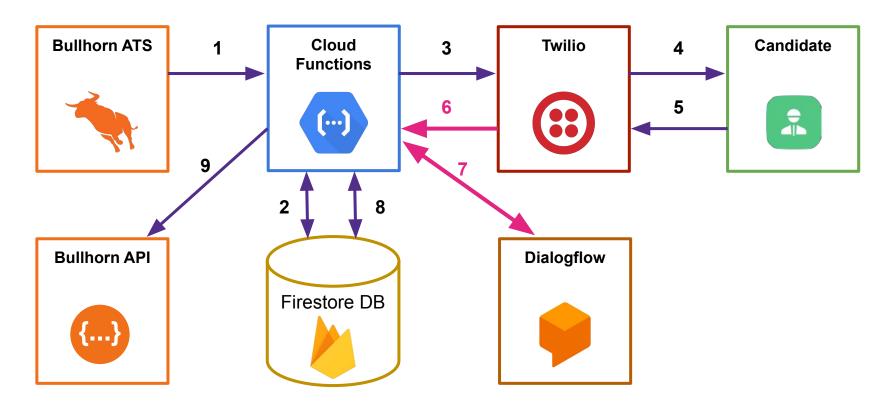
- Problem Statement:
 - Low redeployment correlates with not knowing when assignments end
 - Too high effort to check in with candidates constantly for end dates today
- Features / Benefits:
 - Automatic engagement with candidates and hiring managers to find out when end dates actually are
 - Enables pipeline of redeployment activities thanks to accurate data



How We Built This



How We Would Rebuild It



- Tech stack makes it easy to build out quickly
- 10% model training and 90% customization of business rules
- DialogFlow makes for a nice trainable ML text parser, but needs several layers of business logic before/after to be practical

#BullhornEngage ENGAGE

- Using random number to text from is easy, but ineffective
- Difficulty finding a burning need that customers have that can be filled by chatbots

#BullhornEngage

ENGAGE

Entity Relationship Mapping

Amir Kurtovic

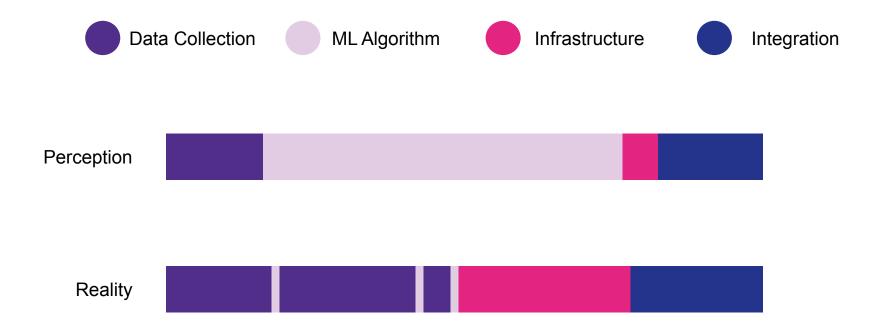
Problem Definition

 Create a machine learning-driven system capable of automatically identifying entity relationships based on historical data

DEMO

ENGAGE

ML Project Phases



Problems We Faced





ML Algorithm



Infrastructure



Integration

No existing labeled dataset

Optimizing for business objective

New API for interacting with hosted models

ETL Pipeline

Cloud Infrastructure configuration

New UI Components

Client onboarding process

Cloud functions

Data Quality

Data Quality

Preprocessing pipelines

Data warehouse exports

Storage

Balancing dataset distributions

Organizational Challenges

 Hard to integrate ML projects into Agile/Lean development processes

- Accepting less than perfect performance
- Breaking down silos

#BullhornEngage

ENGAGE

Automated Invoice Parsing

Lukas Neumann

Automated Invoice Parsing

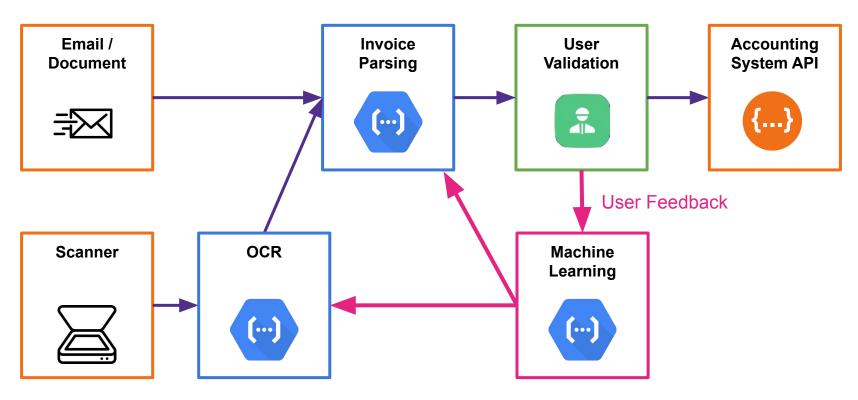
- GOAL: Eliminate manual data entry of incoming invoices
- We built a fully-customizable AI engine which automates document ingestion
 - Users define which fields they want to extract
 - They mark the fields in sample documents to generate training data

ENGAGE

Automated Invoice Parsing

- The engine learns to automatically extract fields as specified by the user
- If the system makes a mistake, user can provide instant feedback
 - We get more training data
 - Constant improvement in accuracy

System Overview





číslo: 17-024-00227 INVOICE Variabilní symbol: 0231001128 Konstantní symbol: IČO: 18838014 DIČ: CZ6206251689 Doručovací adresa: Ing. Jan Svatoň Dodavatel: Na Náhonu č.p. 146 IČO: 00255416 DIČ:CZ00255416 563 01 Lanškroun Obecní úřad Dlouhá Ves Česká republika 155 342 01 Dlouhá Ves Datum splatnosti: 31.12.2017 Datum vystavení dokladu: 19.12.2017 3328351/0100 Bankovní spojení: 18.12.2017 Datum uskut. zdaň. plnění Komerční banka, a.s. Specifický symbol: SWIFT: IBAN: Číslo zákaznického účtu: Příjemce: Evidenční číslo OM: Ing. Jan Svatoň OM číslo: Na Náhonu č.p. 146 Číslo smlouvy: 563 01 Lanškroun Česká republika Adresa OM: kotelna Dlouhá Ves čp. 178

Název položky	množství	jednotka	Cena celkem (Kč bez DPH)
vodné	25	m3	413,00 Kč

 Vyúčtování za období:
 od 1.1.2017
 do 18.12.2017

 Celkem bez DPH
 413,00 Kč

 Celkem s DPH
 474,95 Kč

 Zahrnuté zálohy
 0,00 Kč

 Zaokrouhlení
 0,05 Kč

K úhradě 475,00 Kč







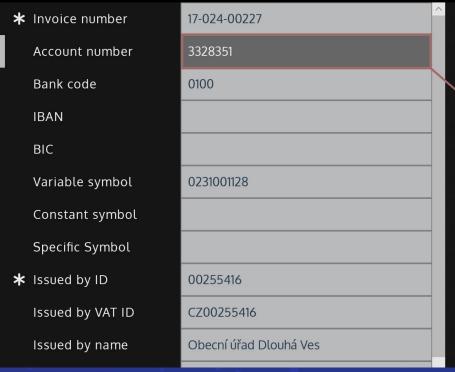


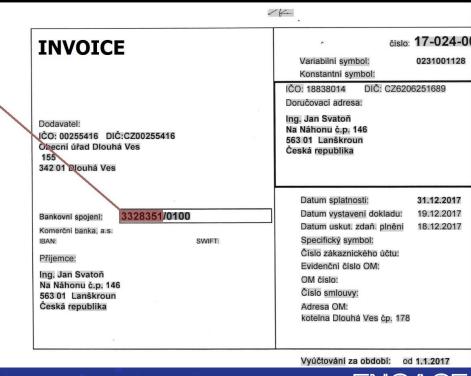




DlouhaVesDoklady2017Vyrobni06.pdf

Inserted 22, 5, 2019 13:29





- A lot of training data is required to reach good accuracy
 - 1,000s to 10,000s of training documents required to reach >95% accuracy
 - Might be challenging to manually create such volume of training documents

- Looking at a specific document, two users might have two completely different answers which field it is (e.g. PO / Invoice #)
 - Generates lot of noise in the training data
 - We created "rule book" for users to address these ambiguities so they create consistent training data

- Users do not expect AI to have perfect accuracy, but once they give feedback they expect AI won't make the same mistake again
 - Infrastructure challenge as this would mean re-training the AI model on the spot to be ready instantly for the next processed document

Questions?

Come visit us at the Al booth

