Project QuickSilver

Personalized Sales Email Generation



Outline

The Team

Problem Statement

Solution Proposal

Step 1 - LinkedIn scraping

Step 2 - Email generation system

Step 3 - Template based approach

Next Steps

The Team



Maverick.ai



Tejas Baindur MEng IEOR'18



Samuel Lin MEng IEOR'19



Aman Tripathi MEng IEOR'18



Soumya Gupta MEng IEOR'18



Ting Wang MEng IEOR'18

Problem Statement

What is the biggest challenge in Sales?



Personalized mails are -

- Manual heavy
- Expensive

Explore -

- Automate current email generation process
- Create an AI engine to generate emails
- Create personas for clients to add personal feel
- Transfer to human salesperson when time is right

Key aspects of personalized sales email

Exciting Subject Line

Attractive Offer in Email Content

Personal Feel

Subject Line: Oski Bear, open this email or you can't beat Stanford!

Dear Mr. Oski Bear,

Don't you hate it when your favorite Pink Floyd song is ruined due to bad earphones?

I really enjoy music and want to fully experience what the musician was trying to convey. A bad earphone just doesn't work for me and for music affectionados!

To allow this at an affordable price, we at Musica, have developed a new earphone which provides great listening experience at a very low price. It would be great to meet you to discuss this product and hopefully enjoy some Pink Floyd songs the way they are intended to!

Yours sincerely,

XXXXXXXX

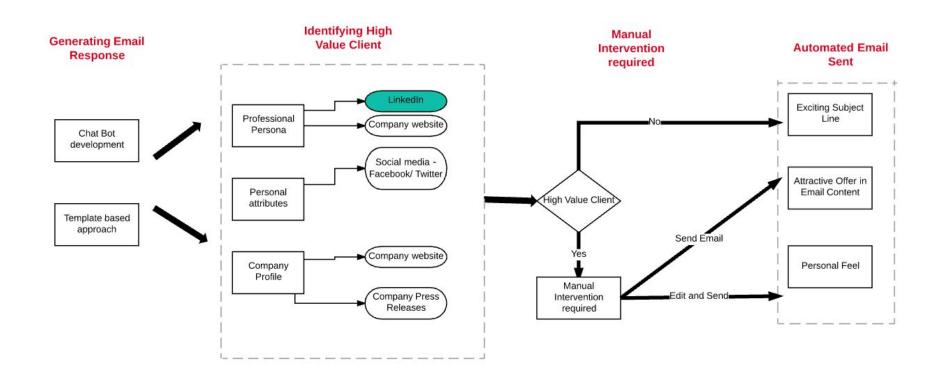


Oski Bear

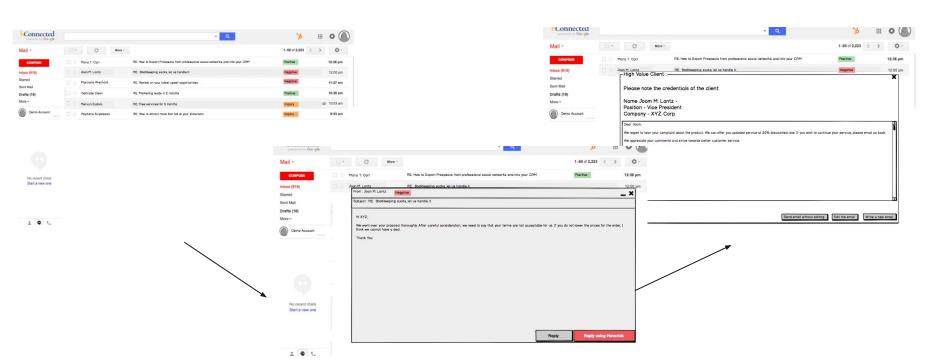


Solution Proposal

Architecture of the project



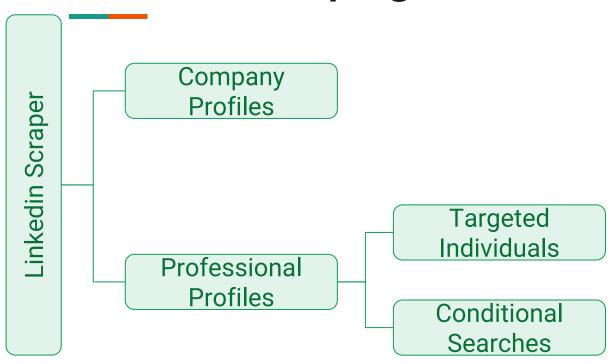
User Interface/Wireframing



Link to Interactive UI

LinkedIn Scraping & Customer Scoring

LinkedIn Scraping & Data Cleansing



- Python Package:Selenium Webdriver
- Scrapes 1,000+
 Professional Profiles
 a day
- Scrapes everything available on Linkedin

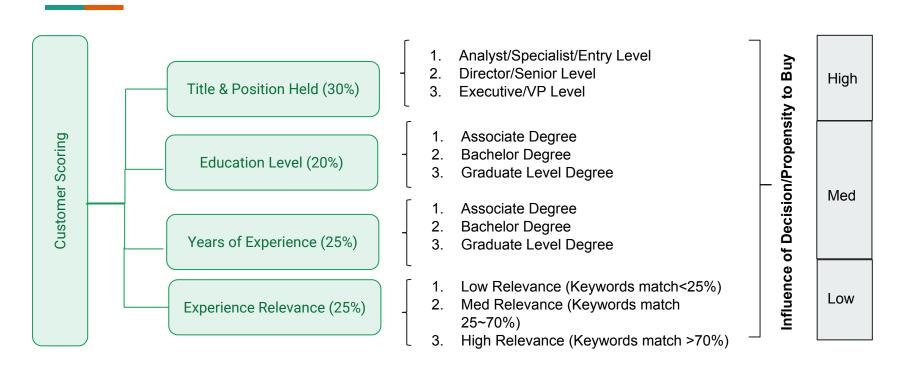
video:

https://www.youtube.com/watch?v=Ile k6cfgHv0

Linkedin Data

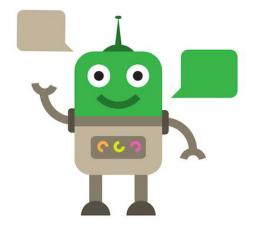
Name	PageURL	HeaderPosition	Location	Indust	ry		Не	aderEducation	1		
Chinmay Abhyankar	https://www.lir	Director, Business Operations at eBay	San Francisco Bay Area	Financ	cial Services	5	Ur	niversity of Cali		October 2015 F @eBay Director, Busin	@eBay Inc
Cherian Abraham	https://www.lir	MBA Student @ UC Berkeley - Haas School	l Plano, Texas	Retail			Ur	niversity of Cali		February 2015 @Fashion For Advisor	
Nikhil Achwal	https://www.lir	Systems Program Manager at Google	San Francisco Bay Area	Intern	et		На	as School of B		March 2015 Pro @Google Learning & De	@Google
Romi Agarwalla	https://www.lir	Economist at Shell Martinez Refinery	San Francisco Bay Area	Oil & E	Energy		U	C Berkeley Haas		August 2012 Pr @Shell Martin Economist	
Anurag Aggarwal	https://www.lir	Product Manager, Cisco	San Jose, California	Inform	nation Techi	nology and Se	rvices U(C Berkeley Haas		2014 Present (: @Cisco Product Mana	@Cisco Sy
Arthur Amador	https://www.lir	Vice President, Financial Consultant	San Francisco Bay Area	Financ	cial Services	5	Ur	niversity of Cali		2008 Present (@Fidelity Inve Vice President	s @The Van
	1		Name	- Linkedin Name	_	PageURL =	Industry.	= Specialties =	Tuno	− Size	_
			Google	Google	-	https://www.lin	•			= 312e mpan 10,001+ emp	ployees
/	/		Facebook	Facebook		https://www.lir	Internet	not found	Public Co	mpan, 10,001+ em	ployees
/			Apple	Apple				-		mpan 10,001+ em	
/			AirBnB	Airbnb		https://www.lir				Held 1001-5000 e	
/			TESLA Uber	Tesla Motors Uber		https://www.lin				mpan 10,001+ em Held 1001-5000 e	
/			McKinsey & Company	McKinsey & Com	npany			t C Management	- '		
Professional		Company Profiles	Amazon	Amazon	1	https://www.lin				mpan 10,001+ em	
		Company i folies	LinkedIn	LinkedIn		https://www.lin	Internet	Online Profess	ic Public Co	mpan 5001-10,000	employees
Profiles			Boston Consulting Group	The Boston Cons	ulting Group	https://www.lin	Managemer	t C Consumer insi	g Partnersh	ip 5001-10,000	employees

Customer Scoring/Cart Model Development



Email Generation System Development

Email Generation System

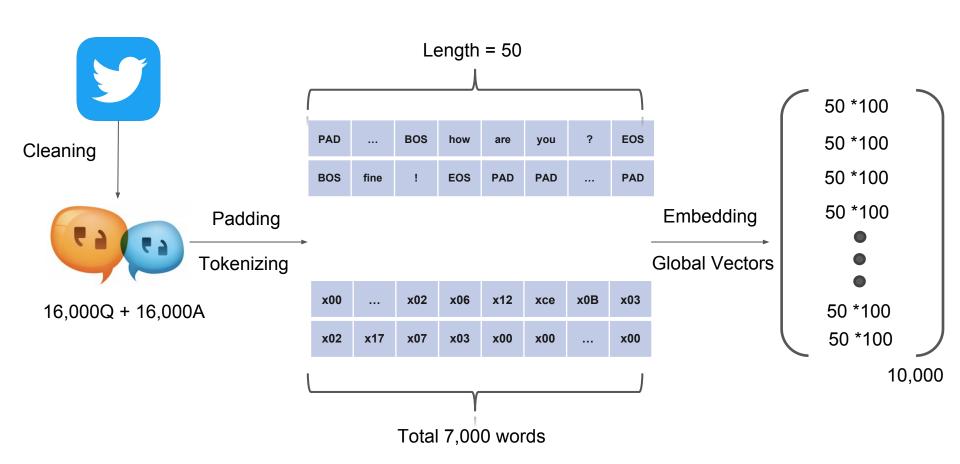


Seq2Seq Model for email generating



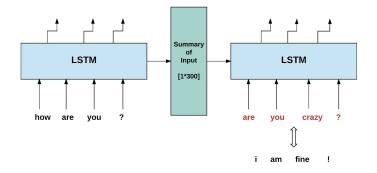
chatbot2017fall@gmail.com

Email Generation - Step 1: Data Preparation for Seq2Seq Training



Email Generation - Step 2: Seq2Seq Model Building and Training

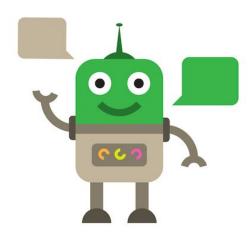
Build 2 LSTM layers : Encoding Layer & Decoding Layer



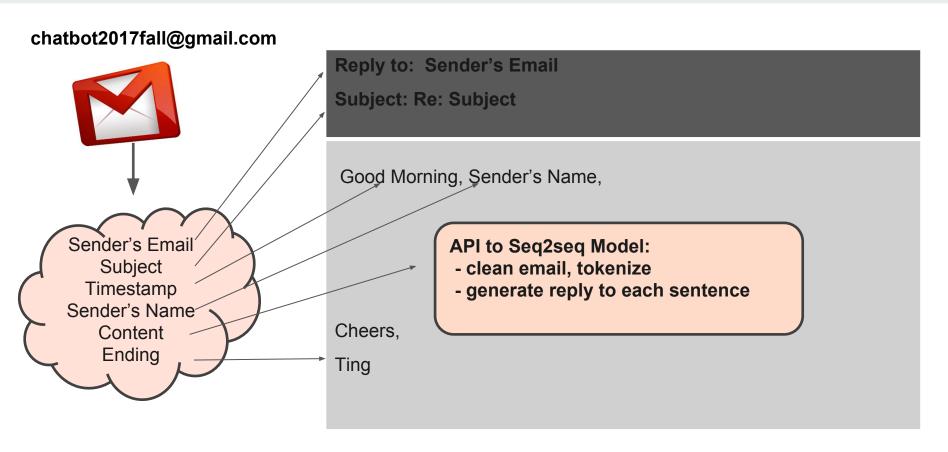
Training With GPU (Titan X) for 160 epochs

- 1 hour / epoch

- Loss: 0.03



Email Generation - Step 3: Fetch Email



DEMO

Template-based approach

The Dataset

• Since sales email datasets are confidential/unavailable, we used the Enron email dataset (which consists of over 500,000 emails) for template based follow up approach.

```
df = pd.read_csv('originalemails.csv')
print(df.shape)
df.head()

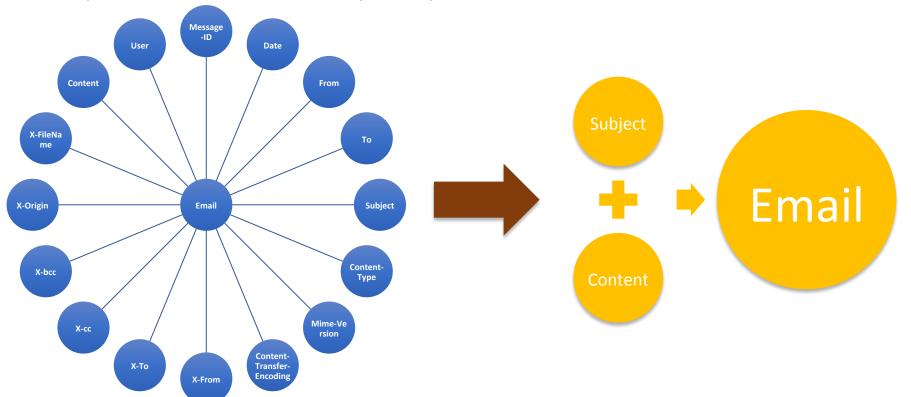
(517401, 2)
```

	file	message
0	allen-p/_sent_mail/1.	Message-ID: <18782981.1075855378110.JavaMail.e
1	allen-p/_sent_mail/10.	Message-ID: <15464986.1075855378456.JavaMail.e
2	allen-p/_sent_mail/100.	Message-ID: <24216240.1075855687451.JavaMail.e
3	allen-p/_sent_mail/1000.	Message-ID: <13505866.1075863688222.JavaMail.e
4	allen-p/_sent_mail/1001.	Message-ID: <30922949.1075863688243.JavaMail.e



Extracting data

• Python libraries OS, Email, sys and pandas were used for data extraction.



Data cleaning and sorting

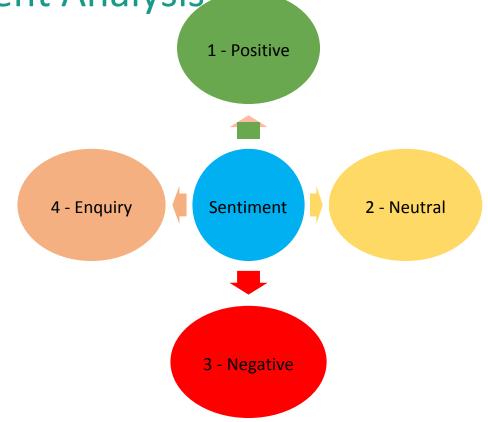
- The emails were sorted chronologically
- Rows with NaN values were removed
- Subject and Content data was kept.
- The special characters, labels, and other irrelevant data removed from the content.



Classification & Sentiment Analysis

The 'content' data was classified into 4 categories as follows:

- Positive: The emails with an overall positive connotation.
- Neutral: Generic emails like information, notices, etc
- 3. Negative: The emails with an overall negative connotation
- Enquiry: The emails which 'ask' for details and information



Template-Based approach

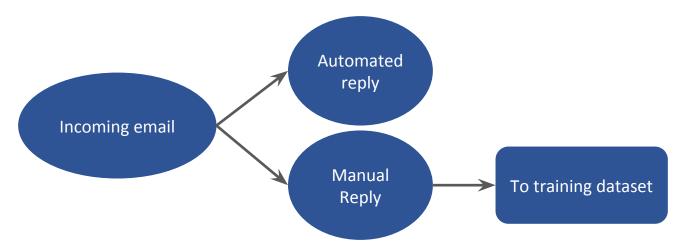
- Based on the sentiment of the incoming email, pre-drafted templates are chosen to be replied by the program
- Optionally a human can intervene for high-value clients



What's Next

What Next?

- Combination of all features into a marketable package
- Use of a genuine sales dataset to increase accuracy of the model
- Introducing feedback mechanism for manually generated/edited replies to enhance agility of automated replies over time



Feedback Mechanism to improve automated replies

References

References

- 1. http://www.seleniumhq.org/docs/
- 2. Global word vector: https://nlp.stanford.edu/projects/glove/
- 3. https://github.com/suriyadeepan/easy-seq2seq
- 4. https://github.com/Currie32/Chatbot-from-Movie-Dialogue
- 5. https://www.kaggle.com/zichen/explore-enron/notebook

Questions?

