PROFILE

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2022.06.20 ~ 현재

<u>언디파인드</u> 개발팀

챗봇 & 추천 알고리즘 개발

2019.12.26 ~ 2020.02.29

<u>카카오</u> 추천팀

추천 알고리즘 개발

2018.11.05 ~ 2019.04.22

<u>한빛소프트</u> AR팀 인공지능파트

텍스트/음성 챗봇 개발

PROJECTS

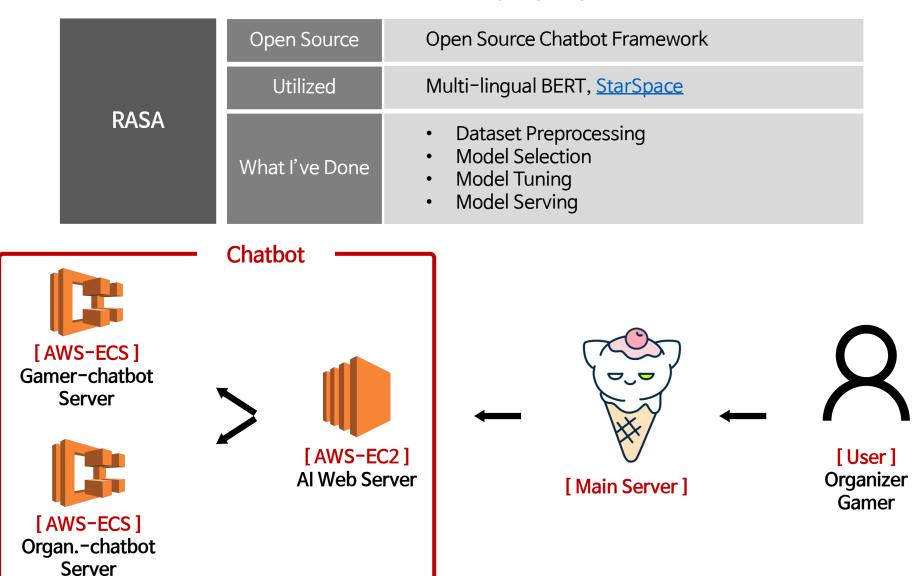
언디파인드	[1] FAQ Chatbot	2 개월
년 의 의 년—	[2] Competetion Rule Recommendation (Dev)	2 주
	[1] DNN Model Quantization	1년 8개월
한양대학교 대학원	[2] Network Embedding Generation	1년 8개월
Machine Learning System Lab.	[3] Artificial Intelligence Assistant	2 개월
	[4] DNN Model Quantization	3 개월
카카오	[1] Automobile Video Recommendation	2 개월
7 7	[2] Comics Recommendation	2주
<u>하빛소프트</u>	[1] (EN) Text Chatbot	2 개월
~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	[2] (KR) Multi-speaker Speech Synthesis Model	4 개월

# 언디파인드

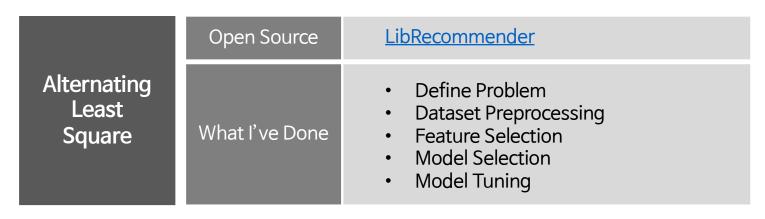
[1] FAQ Chatbot

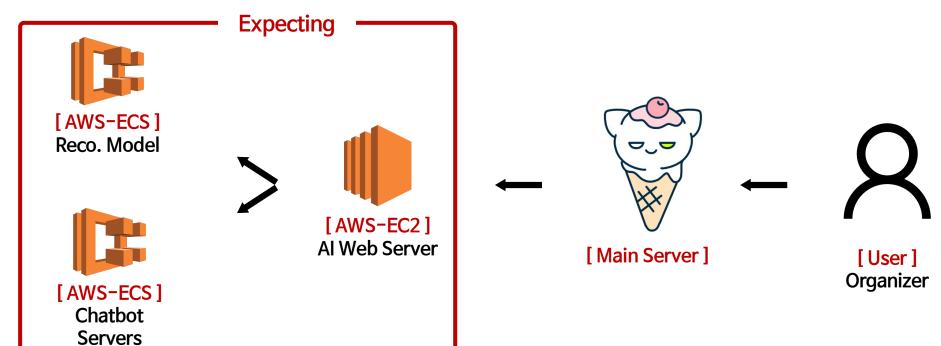
[2] Competetion Rule Recommendation

# FAQ Chatbot (EN/KR)



# Competetion Rule Recommendation (Dev)





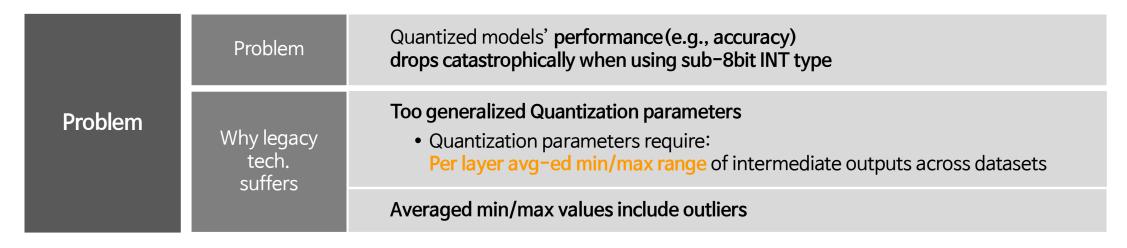
# 한양대학교 대학원 · Machine Learning System Lab.

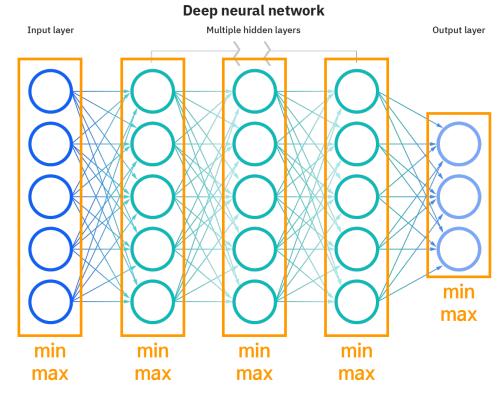
- [1] DNN Model Quantization
- [2] Network Embedding Generation
- [3] Artificial Intelligence Assistant
- [4] DNN Model Quantization



# **DNN Model Quantization**

Definition  What is	What is	General DNN models use Float32 type variables
	VVIIat 15	Quantized models use low-bit INT types at inference
	What for	<ul> <li>Model storage</li> <li>In memory load</li> <li>Matrix multiplication</li> <li>with Float32 type cause bottleneck/unusability in low performance H/W</li> </ul>







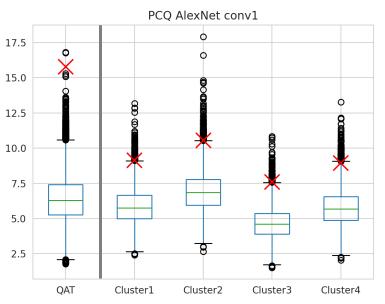
#### Granular Exponential Moving Average (Granular EMA)

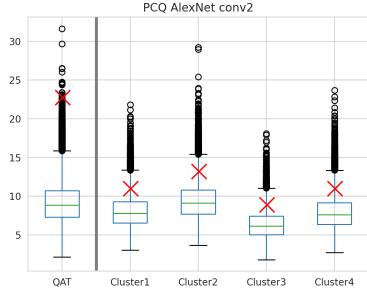
Train Quantization Parameters while excluding outliers

#### **Neural Network Aware Clustering (NNAC)**

Train Quantization Parameters separately across clusters of input images

- Some data might need **shorter min/max range**
- Shorter range means less information loss





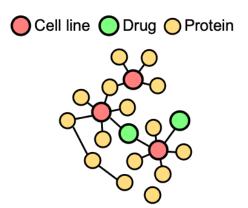
#### Figures' Description

- Shows that our method
  - how efficiently exclude outliers
  - how to work with clusters
- QAT : Baseline (Google)
- Cluster*: Ours
- X: Trained maximum value
- Box-plots: Actual max values per image

# **Network Embedding Generation**

* Currently submitted to BIB (Briefings in Bioinformatics)

[Human Cell lines - Cancer Drugs] Response Prediction Network(graph) dataset consist of **Project**  Cell line nodes description • Drug nodes Protein nodes (connected to Cell lines) My Task: Train embedding vectors of Cell lines and Drugs Extremely unbalanced dataset • About 20,000 Protein nodes About 900 Cell line nodes. About 300 Drug nodes **Problem** Fails to reflect the relationships between Cell lines & Drugs As a result, we got poor response prediction performance

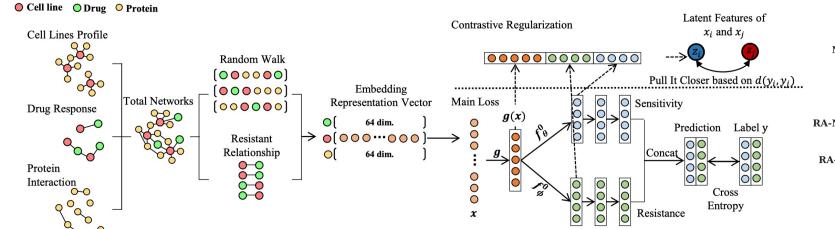


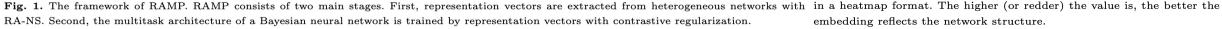
Solution

Make training process to focus on relationships between Cell lines & Drugs

Response-aware Negative Sampling (RA-NS)

- Cell line & Drug nodes use resistant Drug & Cell line nodes as their negative samples
  - * Tested Models: Node2Vec, Graph Convolutional Network, Graph Transformer Network





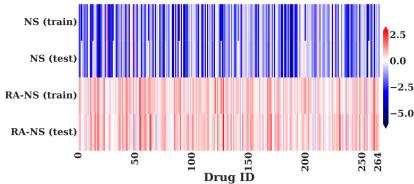
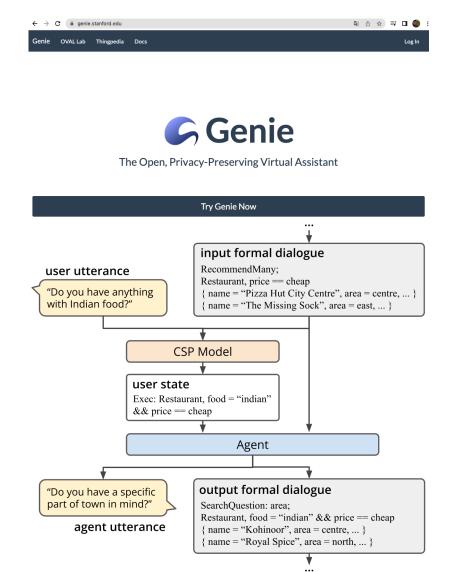


Fig. 2. Embedding similarities among drug and cellines. We subtract the similarity of a drug and its resistant cell lines from the similarity of the drug and its responsive cell lines. The results are normalized and plotted in a heatmap format. The higher (or redder) the value is, the better the embedding reflects the network structure.

# **Artificial Intelligence Assistant**

- Al Assistant App, Almond
  - Currently, the service name has been modified to Genie
  - Developed by Stanford OVAL Lab

- Training Korean Seq2SQL Model
  - Dataset preparation
    - □ Web Crawling
    - ☐ Construct templates of sentences (example of sentences)
    - □ Augment sentences based on templates
  - Train & serve model



#### **DNN Model Quantization**

* Published in 2022 ICEIC (International Conference on Electronics, Information, and Communication)

#### **Quantization Aware Training (Google)**

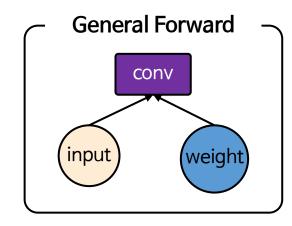
- Fake-quantize all of the weight matrices with a single low-bit type
- Too much quantization errors occur and the trained model gets ruined

#### QuantNoise (Facebook)

- Fake-quantize probabilistically selected subsets of matrices (a subset per matrix)
- Trained models under-prepared for Quantization

#### Fake Single Precision Training (FST)

- Probabilistically select subsets of weight matrices as QuantNoise
- Fake-quantize selected subsets with low-bit type
- Fake-quantize the rests with higher bit type than the selected



# input Dequantized weight weight

#### Solution

**Problem** 

# 카카오

- [1] Automobile Video Recommendation
- [2] Comics Recommendation

Thomson	Purpose	Adjustment of trade-off between exploration & exploitation	
Exp	Exp Sampling 1, 2 h-params tuning	Reason	[Exp-1] High matrix sparsity
1,2			[Exp-2] Considering <b>time bias</b> enhanced by low traffic
F	Ranking	Purpose	Searching the key model among ensembled models
Exp <b>algorithm's</b> 3, 4 h-params tuning	Reason	Other well performing services had been used <b>similar pipelines</b> • Therefore, assumed that the composition of used models are good enough	
	11 21 /	Purpose	Overcome Matrix Factorization model's limitation
Exp 5	Motrix	Reason	Needed to generate reco. results within limited item list  • The limited items rated 30~40th on avg., if we force the limitation off
T deterization	- GCCOTTEGETOTT	3.0112411011	Needed some models which capture information which MF can't

Exp 1, 2 Thomson
Sampling
h-params
tuning

Purpose

Adjustment of trade-off between exploration & exploitation

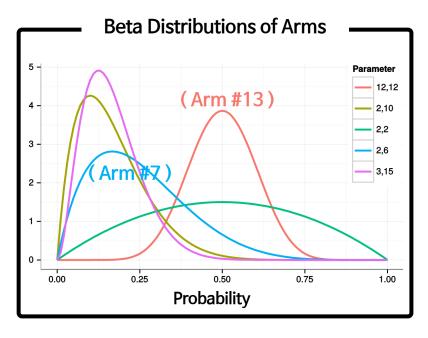
Reason

[Exp-1] High matrix sparsity

[Exp-2] Considering **time bias** enhanced by low traffic







Exp **a** h

Ranking algorithm's h-params tuning Purpose

Searching the key model among ensembled models

Reason

33

0.7070

Other well performing services had been used similar pipelines

• Therefore, assumed that the composition of used models are good enough

Item ID

[ CF ]
Reco. result

[ Text Analysis ]
Reco. result

Item ID **Similarity** 0.8542 17 0.8345 weight=0.3 0.7984 23 0.7784 Example of 33 0.6214 Weighted-sum Item ID **Similarity** 0.8441 0.8385 42 17 0.8001 weight=0.7 51 0.7871

[Ensembled]

Reco. result

Similarity

לוווטו	Similarity
2	0.8303
17	0.8104
33	0.6813
42	0.5870
51	0.5510

#### 〈 MF Model's Reward Matrix 〉



#### (Item2Vec Model's Input Sequence)



# **Comics Recommendation**

Exp 6  Word2Vec input dataset reconstruction	Purpose	Better reflection of Japanese characteristics	
	Doggon	Previously, model used <b>nouns</b> and <b>pronouns</b> only	
	reconstruction	Reason	According to past researches, verbs and adjectives are also important for JP
	Purpose	Strengthen the key model	
	Modified Exp ranking algorithm (RRF to WRF)	<u> </u>	
		Reason	By previous experiment logs, the only MF used reco. pipeline without ensemble method outperformed ensembled pipeline
(KKF LO WKF)	J WKF)	But the ranking algorithm the system was using weakened MF's power	

#### **Comics Recommendation**

Modified
ranking
algorithm
to
Weighted
Rank Fusion

Purpose

Strengthen the key by giving weight to rank values

Reason

By previous experiment logs, the only MF used reco. pipeline without ensemble method outperformed ensembled pipeline

But the Weighted-sum Ranking Algorithm weakened MF's power

#### [Ensembled]

Item ID	Rank
3	1
2	2
17	3

**Example of**Weighted Rank Fusion



Item ID	Rank	
3	1	
17	2	
2	3	
[CF]		

weight=0.3

Item ID	Rank
2	1
42	2
7	3
FT . A . I . 1	

[ Text Analysis ]

 Item ID
 Rank

 3
 1

 42
 2

 2
 3

weight=0.2

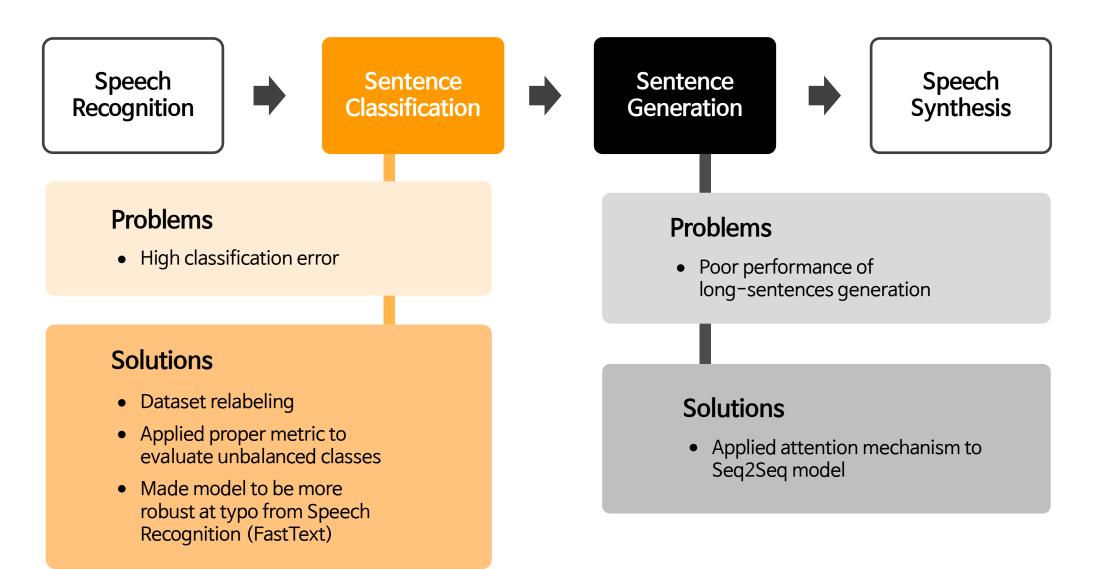
[Image Sim.]

### 한빛소프트

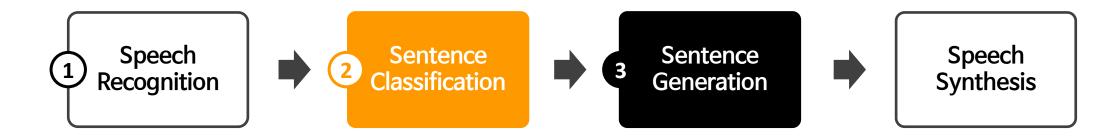
[1] (EN) Text/Audio Chatbot

[2] (KR) Multi-speaker Speech Synthesis Model

# (EN) Text/Audio Chatbot



#### (EN) Text/Audio Chatbot



# (KR) Multi-speaker Speech Synthesis Model

# • Dataset preparation

Web Crawling	Audio files
	Script files
Preprocessing	Cut audio files into files of sentences
	Cut script files into sentences (by comparing STT results)

- H-params optimization
- Demo <a href="https://jarvis08.github.io/pjt_hbs_multi.html">https://jarvis08.github.io/pjt_hbs_multi.html</a>