# Hyeongchan Kim

# https://github.com/kozistr, http://kozistr.tech/about

#### **EDUCATION**

# **Korea University of Technology and Education**

Mar 2016 -

# CHALLENGES & AWARDS

# **Kaggle Challenges :: Competition Expert**

top 1% Google Brain – Ventilator Pressure Prediction (20 / 2605), 2021.

top 4% SIIM-FISABIO-RSNA-COVID-19 Detection (47 / 1305), 2021.

top 7% **Shopee – Price Match Guarantee** (166 / 2426), 2021.

top 2% Cornell Birdcall Identification (24 / 1395), 2020.

top 9% ALAKSA2 Image Steganalysis (93 / 1095), 2020.

top 4% Tweet Sentiment Extraction (84 / 2227), 2020.

top 4% Flower Classification with TPUs (27 / 848), 2020.

top 4% Bengali.Al Handwritten Grapheme Classification (67 / 2059), 2020.

top 3%, Kannada MNIST Challenge (28 / 1214), 2019.

## **Domestic Challenges**

6<sup>th</sup> place, **NAVER NLP Challenge**, SRL Task, 2018.

4<sup>th</sup> / 13<sup>th</sup> place, **NAVER A.I Hackathon**, 2018.

Final Round (Digital Forensic), A.I R&D Challenge, 2018.

9<sup>th</sup> place (3<sup>rd</sup> price, A book as an award), **TF-KR MNIST Challenge**, 2017.

#### **PUBLICATIONS**

[1] Kim et al, CNN ARCHITECTURE PREDICTING MOVIE RATING FROM AUDIENCE'S

REVIEWS WRITTEN IN KOREAN. Jan. 2020.

# INDUSTRY EXPERIENCE

Watcha, Seoul, South Korea

Jun 2020 - Dec 2021

Machine Learning Researcher

- Developed a new sequential recommendation architecture. (named Trans4Rec)
  - Newly proposed transformer architecture to improve the performance in a general manner.
  - Apply proper post-processing logic into the model.
  - In A/B (online) test, *FutureFLAT vs Trans4Rec* (statistically significant p-value < 0.01)

#### Click Ratio: improved 1.01%+

- Developed a music recommendation system (prototype)
- Developed a training recipe to train sequential recommendation architecture robustly. (In service), (named *FutureFLAT*)
  - Build a new module to understand better at the time of inference.
  - Apply augmentations to the various features, leads to performance gain and robustness.
  - In A/B (online) test, FLAT vs FutureFLAT (statistically significant p-value < 0.05)
    - Compared to the previous model, there's been no (statistically significant) change.
    - However, it still seems to be better on the **offline metrics** & **training stability**. So, we chose to use it.
  - In A/B (online) test, <u>Div2Vec</u> vs FutureFLAT (statistically significant p-value < 0.05)</li>

\*Viewing Days (mean): improved 1.012%+

\*Viewing Minutes (median): improved 1.015%+

- Developed the model to predict users' view-time of the contents.
  - Predict how many people going to watch, how much time people going to watch the content before the content is supplied.
  - Find out which features impact users' watch.
- Developed the pipeline to recognize main actors from the poster & still-cut images.
  - Utilize SOTA face detector & recognizer.
  - Optimize pre/post-processing routines for low latency.
- Developed a novel sequential recommendation architecture to recommend what content to watch next. (In service), (named FLAT)
  - Achieve SOTA performance compared to previous SOTA architectures (e.g. BERT4Rec).
  - In A/B (online) test, *previous algorithm* vs *FLAT* (statistically significant p-value < 0.05)

Paid Conversion: improved 1.39%p+

\*Viewing Days (mean): improved 0.25%p+

\*Viewing Minutes (median): improved 4.10%p+

Click Ratio: improved 4.30%p+ Play Ratio: improved 2.32%p+

- Developed Image Super-Resolution model to upscale movie & tv posters, still-cuts.
  - Optimize the codes for fast inference time & memory-efficiency on CPU.
  - In internal evaluation (qualitative evaluation by the designers), it catches details better & handles higher resolution & takes a little time.

- Working as a full-time.
- % \*Viewing Days: how many days users active on an app each month.
- % \*Viewing Minutes: how many minutes user watched the contents.

#### Rainist, Seoul, South Korea

Nov 2019 - Jun 2020

Machine Learning Engineer

- Developed the category classification model of card transactions, designed lightweight purpose for low latency. (In service)
  - In A/B (online) test (statistically significant p-value < 0.05)
    - \*Accuracy: improved about 25 ~ 30%p
- Developed the RESTful API server to serve (general purpose) machine learning models.
  - About 1M MAU service, 500K ~ 1M transactions / day. (1 transaction = about 100 samples)
  - Utilized inference-aware framework (onnx) to reduce the latency. median 100 ~ 200ms / transaction
  - **zero failure rate** (zero 40x, 50x error)
  - Deployed & managed with Kubernetes.
- Developed the classification model, forecasting the possibility of loan overdue.
- Worked as a full-time.

% \*Accuracy: how many people don't update/change their transactions' category.

## VoyagerX, Seoul, South Korea

Jan 2019 – Sep 2019

Machine Learning Engineer

- Developed speaker verification & diarization models to recognize the arbitrary speakers recorded from the noisy environments.
- Developed a semantic image segmentation model to identify a region of hair.
- Developed an image in-paint model to remove hair naturally from the face.
- Worked as an intern.

#### **ELCID, Pangyo, Korea**

Jun 2016 - Aug 2016

Penetration Tester

- Penetrated some products related to network firewall and anti-virus.
- Worked as a part-time job.

#### **OUTSOURCING**

Korea University Course Information Web Parsing, ITL July 2017 – Mar 2018

AWS CloudTrail logger analyzer / formator, ELCID Sep 2019 – Oct 2019

# RESEARCH EXPERIENCE

**Heterogeneous Parallel Computing Lab**, Cheonan, Korea Sep 2018 - Dec 2018 Undergraduate Research

- Wrote a paper about the CNN architecture, which utilizes a channelattention method to TextCNN model, brings performance gain over the task while keeping its latency, generally.
- Handling un-normalized text with various convolution kernel size and spatial dropout.

# TALKS NAVER NLP Workshop 2018, Pangyo, Korea

Dec 2018

• SRL Task, challenging without any domain knowledge. Presented about trials & errors during the competition.

#### **PROJECTS**

#### Generative

# Awesome Generative Adversarial Networks (Stars 670+)

July 2017 -

Implement lots of Generative Adversarial Networks in TF 1.x. & 2.x. Novelty of this project is implementing lots of GANs in TF 1.x & 2.x based on the papers with some tweaks.

#### gan-metrics (Stars 5)

Mar 2020 -

Implement lots of metrics for evaluating GAN in pytorch.

#### **121 Translation**

# Improved Content Disentanglement (Stars 3+)

Sep 2019

Re-implement / tune 'Content Disentanglement' paper in pytorch.

# Image Inpainting Improved Edge-Connect (Stars 9)

Oct 2019

Re-implement / tune 'Edge-Connect' paper in pytorch.

#### Style Transfer Neural Image Style Transfer

Mar 2018

Implement a neural image style transfer.

#### Segmentation Awesome Segmentation (Stars 65+)

Aug 2018

Implement lots of image semantic segmentation and ordered the papers.

**Optimizer** 

## pytorch-optimizer (Stars 20+)

Sep 2021-

Bunch of optimizer implementations in PyTorch with clean-code, strict types. Also, including useful optimization ideas. Most of the implementations are based on the original paper, but I added some tweaks.

## AdaBound Optimizer (Stars 40+)

Jan 2019

Implement AdaBound Optimizer (Luo et al. 2019) w/ some tweaks in tensorflow.

## RAdam Optimizer (Stars 4+)

Sep 2019

Implement RAdam Optimizer (Liu et al. 2019) w/ some tweaks in tensorflow.

**Super Resolution** 

## **Deep Residual Channel Attention Network (Stars 40+)**

Sep 2018

Implement a RCAN model in tensorflow.

## **Enhanced Super Resolution GAN (Stars 30+)**

Jun 2019

Implement an ESRGAN model in tensorflow.

## Natural and Realistic SISR w/ Explicit NMD (Stars 5+)

Apr 2020

Implement a NatSR model in pytorch.

#### NLP Improved TextCNN (Stars 4+)

Dec 2018

Implement an improved TextCNN model (Kim et al. 2020)

## **Text Tagging**

Dec 2018

Implement a text category classifier in tensorflow.

R.L

#### Rosetta Stone (Stars 560+)

Sep 2018-

Hearthstone simulator using C++ w/ some R.L.

I contributed to the project by implementing `feature extractor` and `neural network'

in libtorch++.

**Speech Synthesis** 

#### **Tacotron**

Jan 2019

Implement a google tacotron speech synthesis in tensorflow.

**Open Source** 

syzkaller :: New Generation of Linux Kernel Fuzzer

**Contributions** 

#575

**simpletransformers ::** Transformers made simple with training, evaluating, and prediction possible with one line each #290

**pytorch-image-models ::** Pytorch image models, scripts, pretrained weights #1058, #1069

**deit ::** DeiT Data-efficient Image Transformers #140, #147, #148

MADGRAD :: MADGRAD Optimization Method

#11

**tensorflow-image-models**: Tensorflow Image Models (tfimm) is a collection of image models with pretrained weights, obtained by porting architectures from timm to Tensorflow

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