

# 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.AI 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

**Toss core**, Seoul, South Korea

Dec 2021 – Present

Data Scientist

- Developed the text classification model to categorize users' reviews.
  - Visualize the summarized trend of the keywords, which show the point of the opinion.
  - Boost the analyze the users who give feedback with rich information (may help to boost NPS score).
- Developed the robust captcha model to predict numeric captchas.

- Light-weight CNN model for real-time inference (about *1000 TPS* for batch transaction, *50 TPS* for a sample on CPU)
- Build augmentations to build a robust model.
- Save **\$10,000 ~ \$30,000 / year**
- In A/B (online) test, google vision OCR *vs New Captcha Model* (statistically significant p-value < 0.05)  
**Accuracy** (top-1): improved **50%p** (49% to 95%)  
**Latency** (p95): reduced by **x80** (about 1000ms to 12ms)
- Developed the model to forecast the transactions' category to purchase next month & few weeks.
  - Transformer-based architecture (customed with newly proposed methods).
  - Calibration-aware training.
  - In A/B (online) test, *previous ML model vs AdsClassifier* (statistically significant p-value < 0.05)  
 Soon!
- Developed the loan overdue prediction model.
  - EDA to find the useful features correlated with the overdue users.
  - Build the robust CV & ensemble strategy in an aspect of the on/offline performance.
- Developed the card transaction category classification model.
  - Transformer-based architecture, about 900 TPS on a single GPU.
  - Handle noisy-text (transaction) & label, class-imbalanced problem.
- Contributed to the team-culture (e.g. collaboration tools, style-guides, etc)
- Working as full-time.

**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, [\*Div2Vec\*](#) vs *FutureFLAT* (statistically significant p-value < 0.05)
  - \*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 and how much time people are going to watch the content before the content supplied.
  - Find out which features impact users' watches.
- 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 low latency & memory-efficiency on CPU.
  - An internal evaluation (qualitative evaluation by the designers), it catches details better & handles higher resolution & takes a little time.
- Working as 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 for forecasting the possibility of loan overdue.
- Worked as 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 products.
- 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

<b>RESEARCH EXPERIENCE</b>	<b>Heterogeneous Parallel Computing Lab</b> , Cheonan, Korea      Sep 2018 - Dec 2018 Undergraduate Research <ul style="list-style-type: none"> <li>Wrote a paper about the CNN architecture, which utilizes a channel-attention method to TextCNN model, brings performance gain over the task while keeping its latency, generally.</li> <li>Handling un-normalized text with various convolution kernel sizes and spatial dropout.</li> </ul>
<b>TALKS</b>	<b>NAVER NLP Workshop 2018</b> , Pangyo, Korea      Dec 2018 <ul style="list-style-type: none"> <li>SRL Task, challenging without any domain knowledge. Presented about trials &amp; errors during the competition.</li> </ul>
<b>PROJECTS</b>	
<b>Generative</b>	<b>Awesome Generative Adversarial Networks (Stars 700+)</b> 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.  <b>gan-metrics (Stars 5)</b> Mar 2020 – Implement lots of metrics for evaluating GAN in PyTorch.
<b>I2I Translation</b>	<b>Improved Content Disentanglement (Stars 3+)</b> Sep 2019 Re-implement / tune 'Content Disentanglement' paper in PyTorch.
<b>Image Inpainting</b>	<b>Improved Edge-Connect (Stars 9)</b> Oct 2019 Re-implement / tune 'Edge-Connect' paper in PyTorch.
<b>Style Transfer</b>	<b>Neural Image Style Transfer</b> Mar 2018 Implement a neural image style transfer.
<b>Segmentation</b>	<b>Awesome Segmentation (Stars 65+)</b> Aug 2018 Implement lots of image semantic segmentation and ordered the papers.
<b>Optimizer</b>	<b>pytorch-optimizer (Stars 30+)</b> 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

[#61](#)