



RSNA 2017 Pediatric Bone Age Challenge

Alexandr Kalinin, University of Michigan
December 27, 2017



About myself and our ODS team

@alxndrkalinin, Alexandr Kalinin, University of Michigan

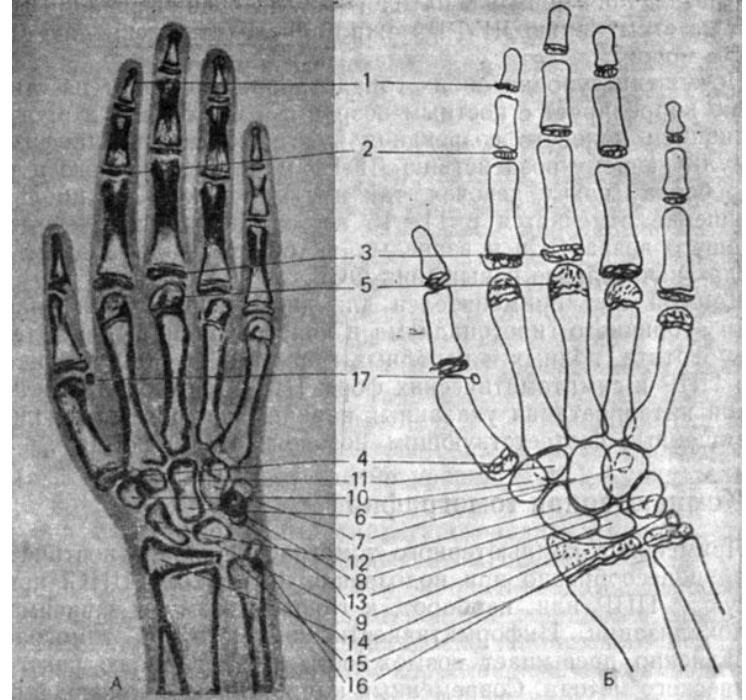
- BSc, MSc in Applied Math and Informatics from NSTU, 2004-2010
- Fulbright Scholar in Statistics at University of California, Los Angeles, 2013
- 5th year PhD Candidate in Bioinformatics at University of Michigan

- @ternaus, Vladimir Iglovikov, Lyft Inc.
- @rakhlin, Alexander Rakhlin, National Research University of Electronic Technology
- @shvetsiya, Alexey Shvets, MIT

Goals of the project

Pediatric bone age assessment – is a common clinical practice to diagnose endocrine and metabolic disorders in child development.

Goal: Develop an algorithm which can most accurately determine skeletal age on a validation set of pediatric hand radiographs.



Challenge organization and logistics

Radiological Society of North America (RSNA) – international society of radiologists, medical physicists and other medical professionals with more than 54,000 members from 136 countries. Custom competition platform.

- Training phase start: Aug. 5, 2017, midnight
- Leaderboard phase start: Sept. 1, 2017, midnight
- Test phase start: Oct. 7, 2017, midnight
- Competition end: ~~Oct. 15~~ Oct. 17, 2018, 3:55 a.m.

The Challenge winners invited to join the Challenge panel at the RSNA Annual Meeting in Chicago, Nov 26. Data made publicly available.

Data description

Data structure:

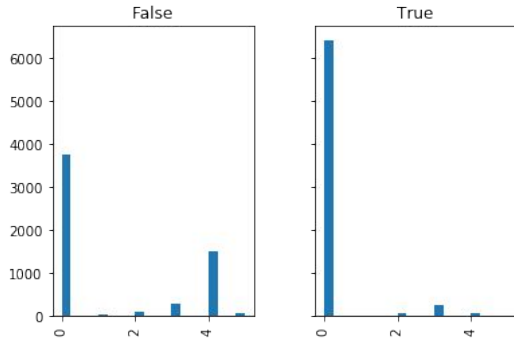
- X-ray images of a left hand
- patient sex
- [outcome] bone age in month

Number of patients:

- Training: 12.6k
- Validation set: 1425
- Test: 200

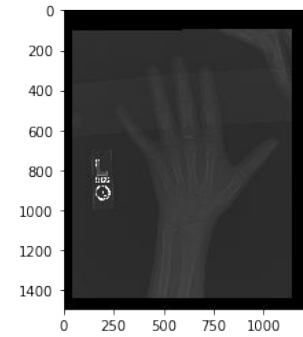
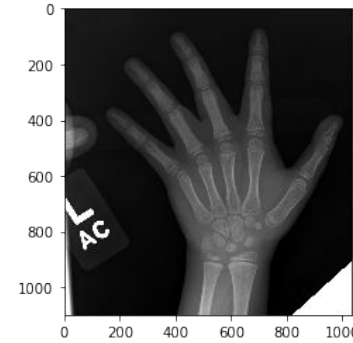
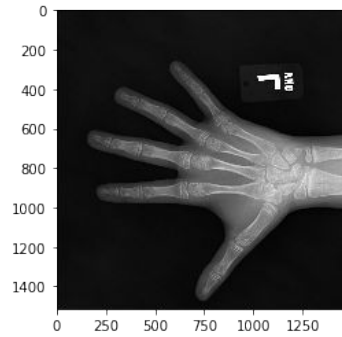
EDA: train set patient sex distribution

- Approx 50/50 sex ratio
- up to 20 years old
- most ages divisible by 6

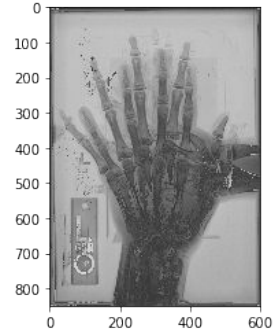
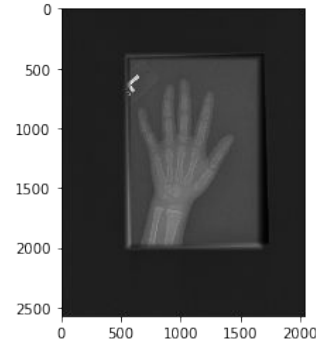
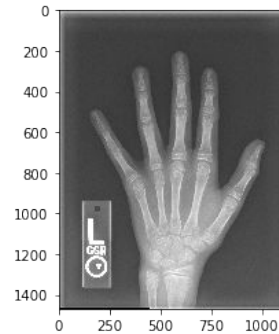
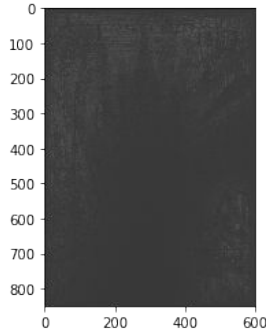


EDA: varying quality of radiographs

- annotations
- very light/dark areas
- rotated images
- contrast issues
- etc



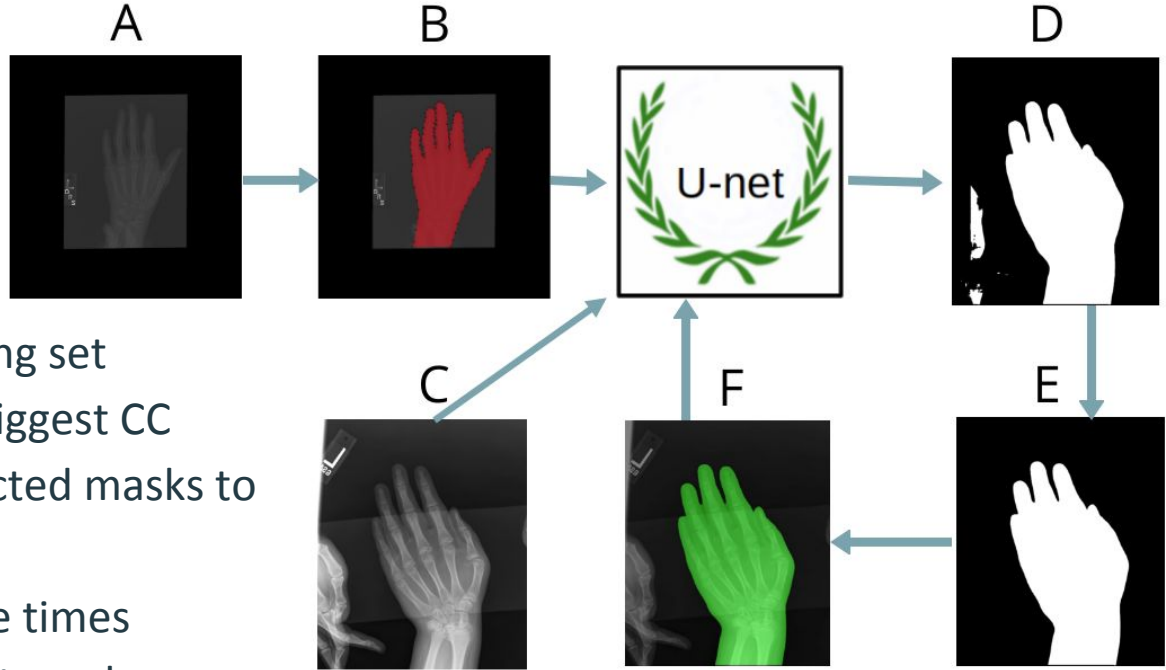
average:



Preprocessing: segmentation by @ternaus

U-Net positive mining:

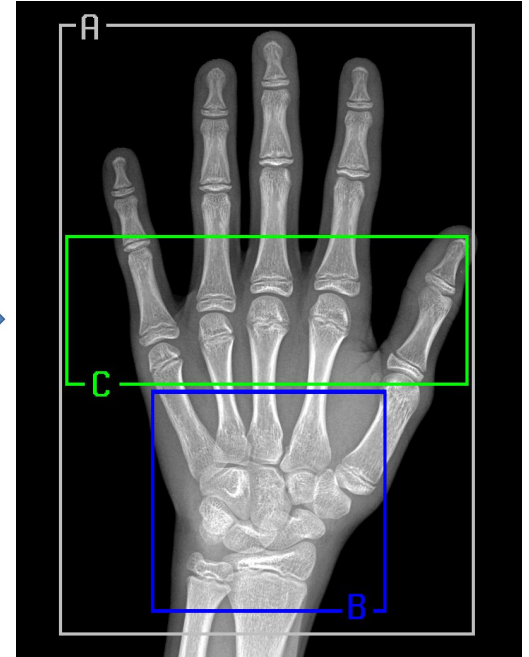
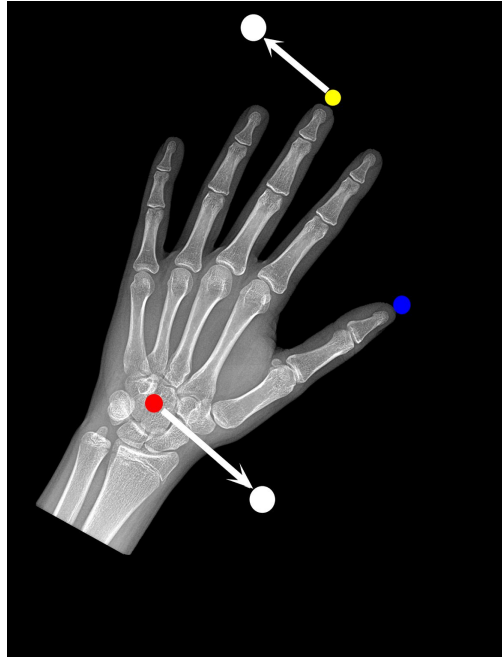
1. manually label 100 masks with Supervisely
2. train U-Net
3. predict the rest of training set
4. in every image leave 1 biggest CC
5. visually inspect all predicted masks to keep those of good quality
6. repeat steps 2-5 x5 more times
7. manually label 100 worst masks



Preprocessing: registration by @rakhlin

Needed to evaluate different areas of the hand (A,B,C)

1. manually label 800 radiographs with keypoints
2. train VGG-like net on downsides images
3. predict coordinates on the rest
4. register with affine transform



VGG-like architectures

- ELU
- Adam

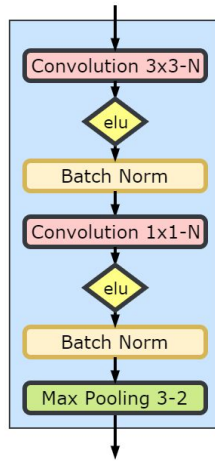
Regression:

- MAE loss
- age in $[-1, 1]$

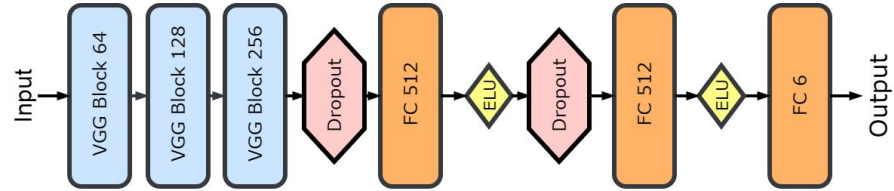
Classification:

- BCE loss
- 240 age bins

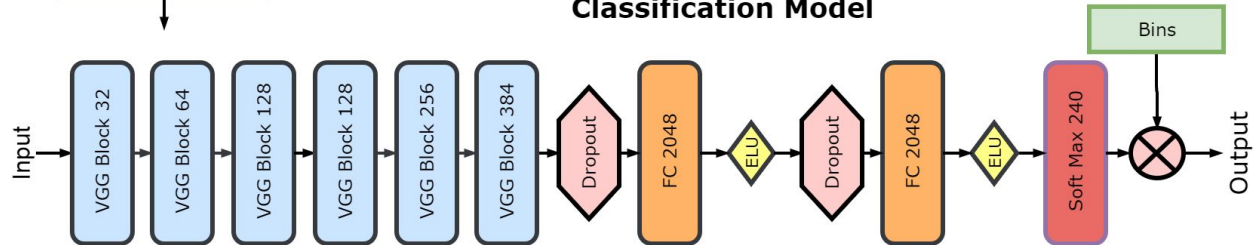
VGG Block



Key Points Model



Classification Model



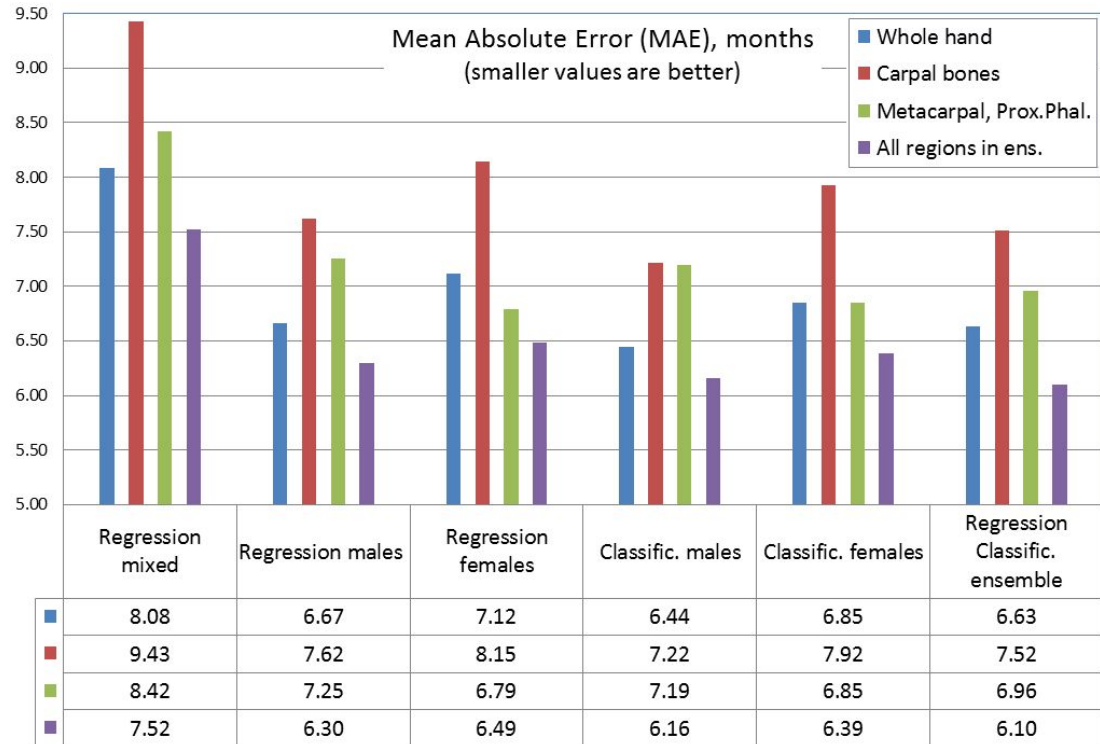
Results: different models, bones, and sex

Sex is important

MAE(B) > MAE(C) > MAE(A)
> MAE (ensemble)

Classification perform
slightly better than
regression

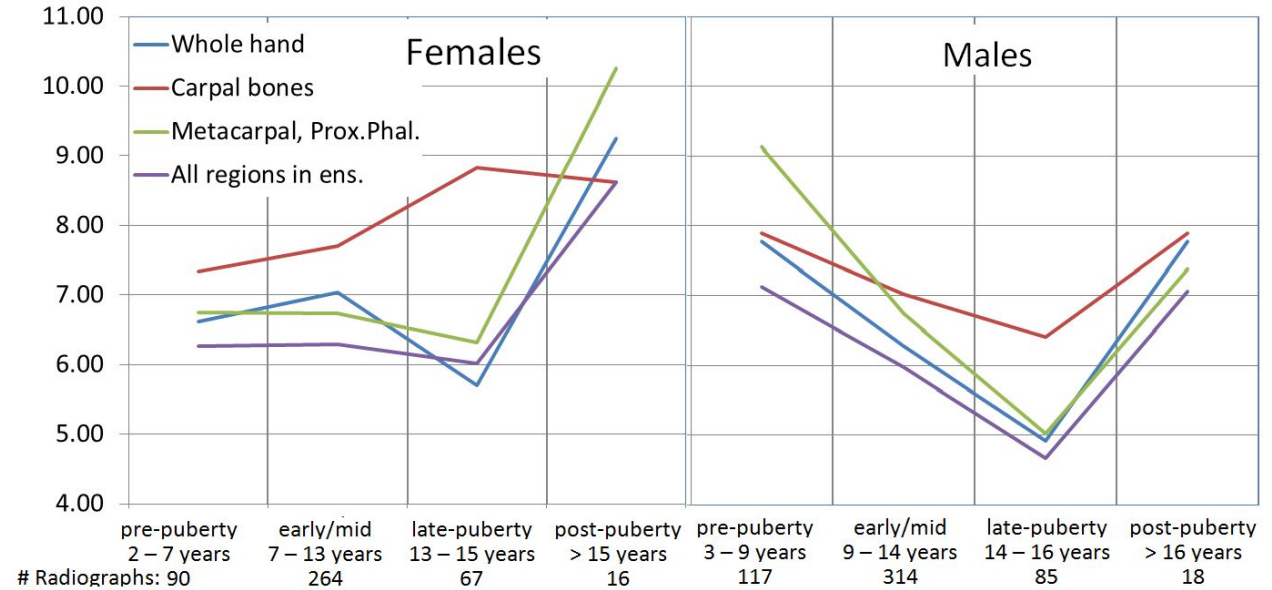
Ensembling improves
results



Results: skeletal development stages

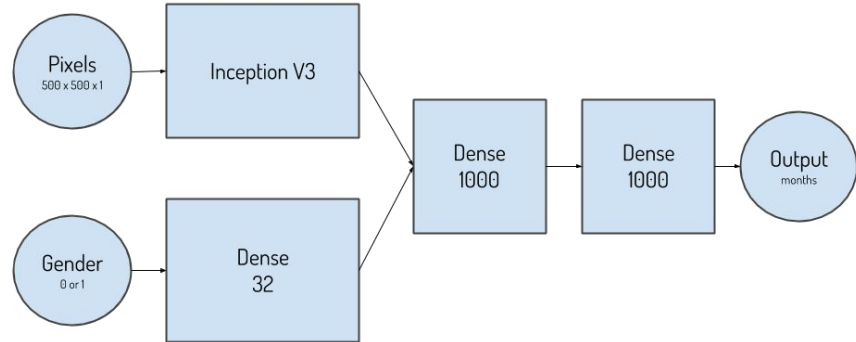
Unlike previous studies, we found carpal bones not the most important compared to others

More data per age – higher accuracy



Results: winning solution

- based on Inception v3
- images resized to 500x500
- no preprocessing
- 5 networks in final ensemble
- TTA-10



Conclusions

- we joined late & didn't win the competition

BUT

- we learned a lot
- we produced potentially useful solution
- we published a preprint that was downloaded >300 times in 12 days

Pediatric Bone Age Assessment Using Deep Convolutional Neural Networks

 Vladimir Iglovikov,  Alexander Rakhlin,  Alexandr Kalinin, Alexey Shvets

doi: <https://doi.org/10.1101/234120>

This article is a preprint and has not been peer-reviewed [what does this mean?].

Abstract

Info/History

Metrics

 Preview PDF

ARTICLE USAGE

Article lifetime

Last 6 months

This month

Article usage: December 2017 to December 2017

Show by month	Abstract	Pdf
Total	1,173	315

Framework

- dynamic, imperative, fast
- aims to be a drop-in replacement for NumPy with GPU support and autodiff
- provides convenient DataLoader with out-of-the-box multiprocessing
- provides many DL abstractions
- supports multi-GPU out-of-the-box
- has decent tutorials, forum, and not-too-steep learning curve

PYTORCH

Deep Learning with PyTorch