

using open-source implementation

Problems: how to replicate models by reading the papers.

• Look for open-source implementation.

From GitHub

For instance: <https://github.com/KaimingHe/deep-residual-networks> implementation of Resnet, by the author

uses the Caffe framework.

Advantages: can use others' pre-trained model. and do transfer learning

## Transfer Learning

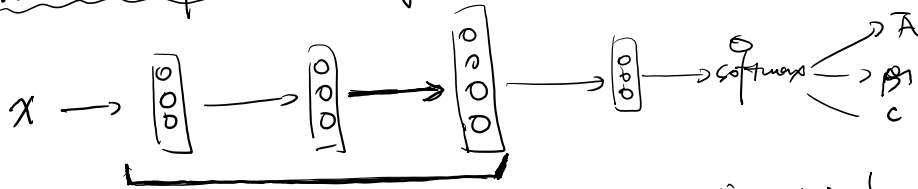
use pre-trained model  $\Rightarrow$  use it on your own task

Training is resource consuming.

can download open-sourced weights  $\Rightarrow$  use on your own problem

Ex cat detection

Download pre-trained weights



freeze trainable parameters - if pre-trained weights,  $\>$  is  $\frac{1}{2}$  train  $\frac{1}{2}$  test.

save to disk the first part of the model

- If your dataset is larger you can freeze fewer layer (more trainable parameters).
- To an extreme, ~~to~~ use the pretrained weights to initialize then train/update ALL of them

In practice, you will do MUCH BETTER using weights pretrained by others on large dataset

## Data augmentation:

Having more data will always help. for the majority of CV tasks

Mirror / cropping / rotation / shearing / local

Color shifting - (play with RGB three channels)  $+50, 0, -10$  之类

$\Rightarrow$  learning algorithm more robust to change of color

Advanced different ways to sample RGB colors. | Implemented in AlexNet

PCA color augmentation: if your image is mainly purple.

$\Rightarrow$  add & subtract a lot of blue & red. But leave G alone.

Implementing distortion during training

hard disk

CPU thread.

distortion



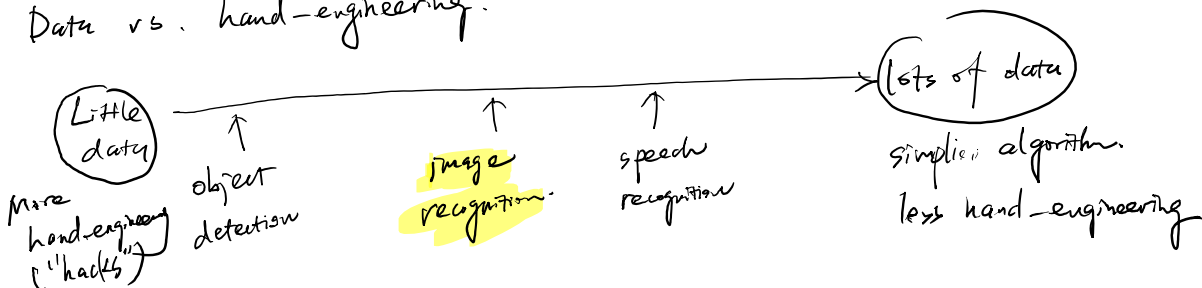
mini-batch

training.

用独立的CPU/GPU来跑distortion (实时)  
即不是都在主线程上跑。

## State of computer vision

Data vs. hand-engineering.



Two sources of data

• Labeled data

• Hand engineered features / network architecture / other component.

Hand engineered features / network architecture

the fact that we don't have much data leads us to more complex network architecture

When you have lots of <sup>(labeled)</sup> data

→ Better off spending time on building good sys. instead of hand-engineering

When you have little data — one thing that helps is TRANSFER LEARNING.

Tips for Doing well on Benchmark / winning competition

### Ensembling

3-15 networks

- Train several NN independently, & average outputs.  $\uparrow$   
[not averaging weights]

— time consuming! so not a good idea for production. (unless you have a generous computational budget)



Multi-crop for test time. [crop images]

- Run classifier on multiple versions of test images and average results

to test image to crop



Use open source code

- use architectures of networks published in the lit
- use open source implementation.
- use pre-trained model and fine-tune with your datasets