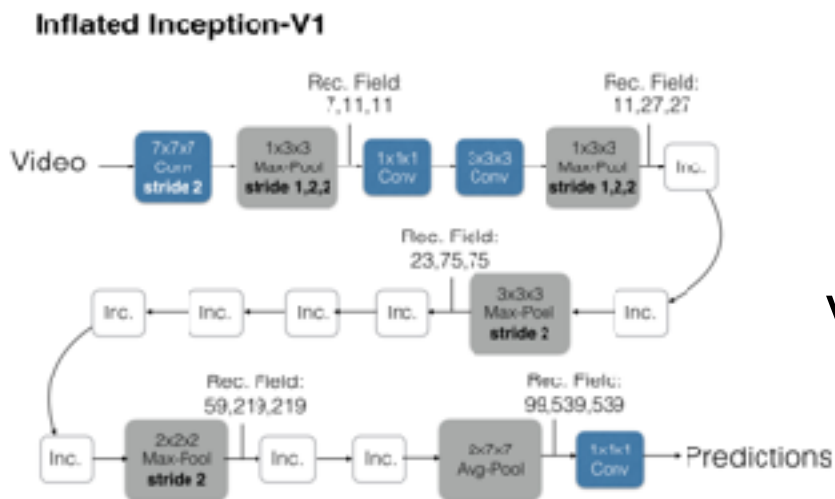
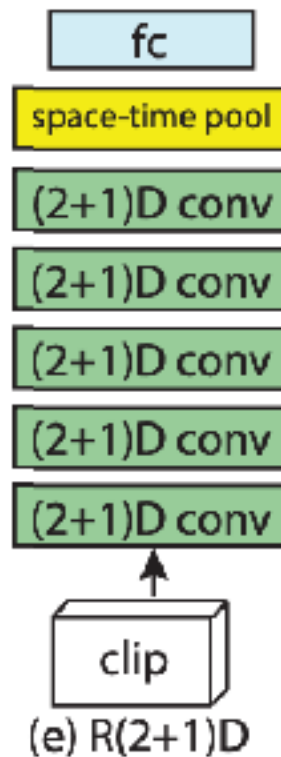


Paper Battle #1



I3D [CVPR'17]

vs.



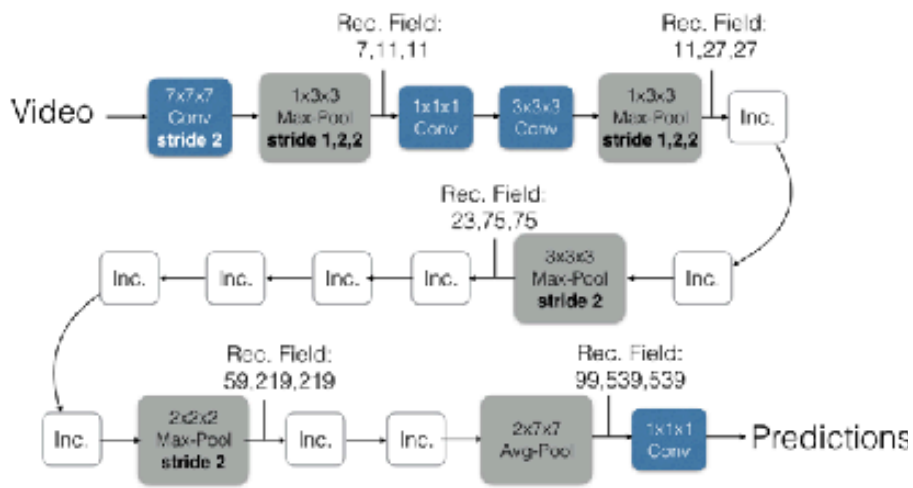
R(2+1)D [CVPR'18]

Arguments for I3D

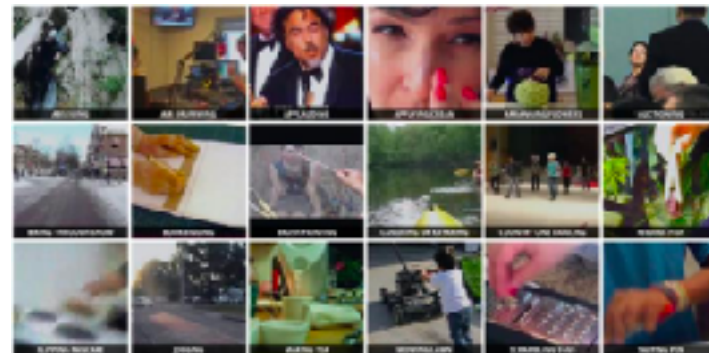
Dataset and Model Contribution

- In addition to proposing a new video model, the paper also introduces a new large-scale dataset.

Inflated Inception-V1



a) I3D Model



b) Kinetics Dataset

Research Impact

- Arguably, the I3D paper had a larger impact on the video recognition community.

Quo vadis, action recognition? a new model and the kinetics dataset

J Carreira, A Zisserman
proceedings of the IEEE Conference on Computer Vision and Pattern ...

7233

2017

A closer look at spatiotemporal convolutions for action recognition

D Tran, H Wang, L Torresani, J Ray, Y LeCun, M Paluri
Proceedings of the IEEE conference on Computer Vision and Pattern ...

2683

2018

[deepmind/kinetics-i3d](#)

Convolutional neural network model for video classification trained on the Kinetics dataset.

● Python ·  1.7k · Updated on Sep 12, 2019

[facebookresearch/VMZ](#)

VMZ: Model Zoo for Video Modeling

● Python ·  1k · Updated on Aug 31, 2021

Better Results

- Even though I3D was one year older than R(2+1)D, it still achieved better results at the time of R(2+1)D publication.

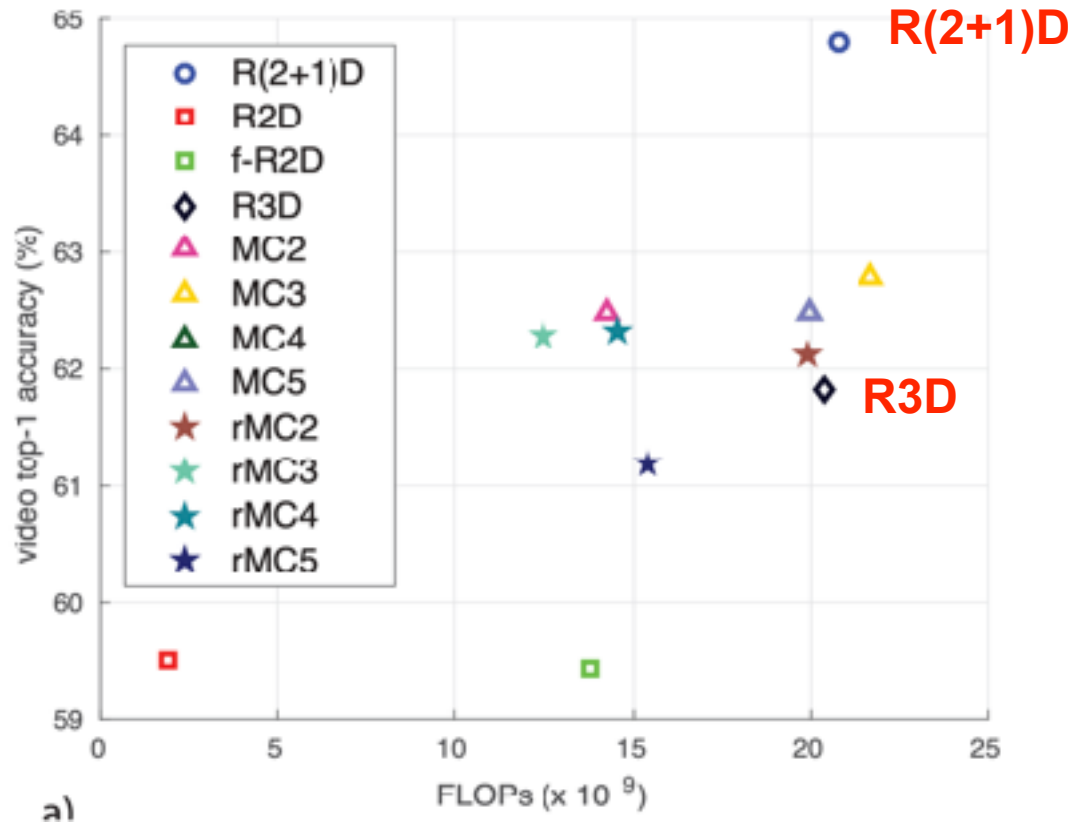
| method | pretraining dataset | top1 | top5 |
|--------------------|---------------------|-------------|-------------|
| I3D-RGB [4] | none | 67.5 | 87.2 |
| I3D-RGB [4] | ImageNet | 72.1 | 90.3 |
| I3D-Flow [4] | ImageNet | 65.3 | 86.2 |
| I3D-Two-Stream [4] | ImageNet | 75.7 | 92.0 |
| R(2+1)D-RGB | none | 72.0 | 90.0 |
| R(2+1)D-Flow | none | 67.5 | 87.2 |
| R(2+1)D-Two-Stream | none | 73.9 | 90.9 |
| R(2+1)D-RGB | Sports-1M | 74.3 | 91.4 |
| R(2+1)D-Flow | Sports-1M | 68.5 | 88.1 |
| R(2+1)D-Two-Stream | Sports-1M | 75.4 | 91.9 |

| method | pretraining dataset | UCF101 | HMDB51 |
|------------------------|---------------------|-------------|-------------|
| Two-Stream [29] | ImageNet | 88.0 | 59.4 |
| Action Transf. [40] | ImageNet | 92.4 | 62.0 |
| Conv Pooling [42] | Sports-1M | 88.6 | - |
| <i>FSTCN</i> [33] | ImageNet | 88.1 | 59.1 |
| Two-Stream Fusion [10] | ImageNet | 92.5 | 65.4 |
| Spatiotemp. ResNet [9] | ImageNet | 93.4 | 66.4 |
| Temp. Segm. Net [39] | ImageNet | 94.2 | 69.4 |
| P3D [25] | ImageNet+Sports 1M | 88.6 | - |
| I3D-RGB [4] | ImageNet+Kinetics | 95.6 | 74.8 |
| I3D-Flow [4] | ImageNet+Kinetics | 96.7 | 77.1 |
| I3D-Two-Stream [4] | ImageNet+Kinetics | 98.0 | 80.7 |
| R(2+1)D-RGB | Sports 1M | 93.6 | 66.6 |
| R(2+1)D-Flow | Sports 1M | 93.3 | 70.1 |
| R(2+1)D-TwoStream | Sports 1M | 95.0 | 72.7 |
| R(2+1)D-RGB | Kinetics | 96.8 | 74.5 |
| R(2+1)D-Flow | Kinetics | 95.5 | 76.4 |
| R(2+1)D-TwoStream | Kinetics | 97.3 | 78.7 |

Arguments for $R(2+1)D$

Accuracy-Efficiency Tradeoff

- R(2+1)D has a lot better accuracy-efficiency tradeoff than 3D CNNs (e.g., I3D).



Industry Impact

- R(2+1)D was pre-trained on 65M Instagram videos and deployed internally at Facebook for various use cases.
- This includes flagging cases of violence, pornography, scams, objectionable content, etc.

Internal large-scale computing platform

To support video research and development, Facebook has built an internal platform called Lumos. Lumos provides a simplified process for developers to train AI models on images and videos. First is the data. Many new tools on Lumos around data annotation can do image clustering. Second is the model. Developers can select off-the-shelf deep neural networks from Lumos and integrate particular features, like image feature and text features, into the model.

Lumos runs on billions of images and has more than 400 visual models for purposes of objectionable-content detection and spam fighting to automatic image captioning.

Scalability

- Due to its efficient design, R(2+1)D is easier to scale to massive datasets (e.g., IG-65M) and larger model sizes.

| Method; pre-training | top-1 | top-5 | Input type |
|------------------------------------|-------------|-------------|--------------------|
| I3D-Two-Stream [11]; ImageNet | 75.7 | 92.0 | RGB + flow |
| R(2+1)D-Two-Stream [14]; Sports-1M | 75.4 | 91.9 | RGB + flow |
| 3-stream SATT [69]; ImageNet | 77.7 | 93.2 | RGB + flow + audio |
| NL I3D [65]; ImageNet | 77.7 | 93.3 | RGB |
| R(2+1)D-34; Sports-1M | 71.7 | 90.5 | RGB |
| Ours R(2+1)D-34; IG-Kinetics | 79.1 | 93.9 | RGB |
| Ours R(2+1)D-34; IG-Kinetics; SE | 79.6 | 94.2 | RGB |
| Ours R(2+1)D-152; IG-Kinetics | 80.5 | 94.6 | RGB |
| Ours R(2+1)D-152; IG-Kinetics; SE | 81.3 | 95.1 | RGB |