

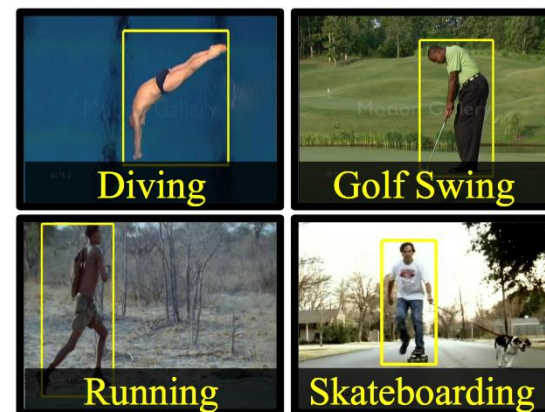
Learning from Synthetic Data

Thomas Brox
University of Freiburg, Germany

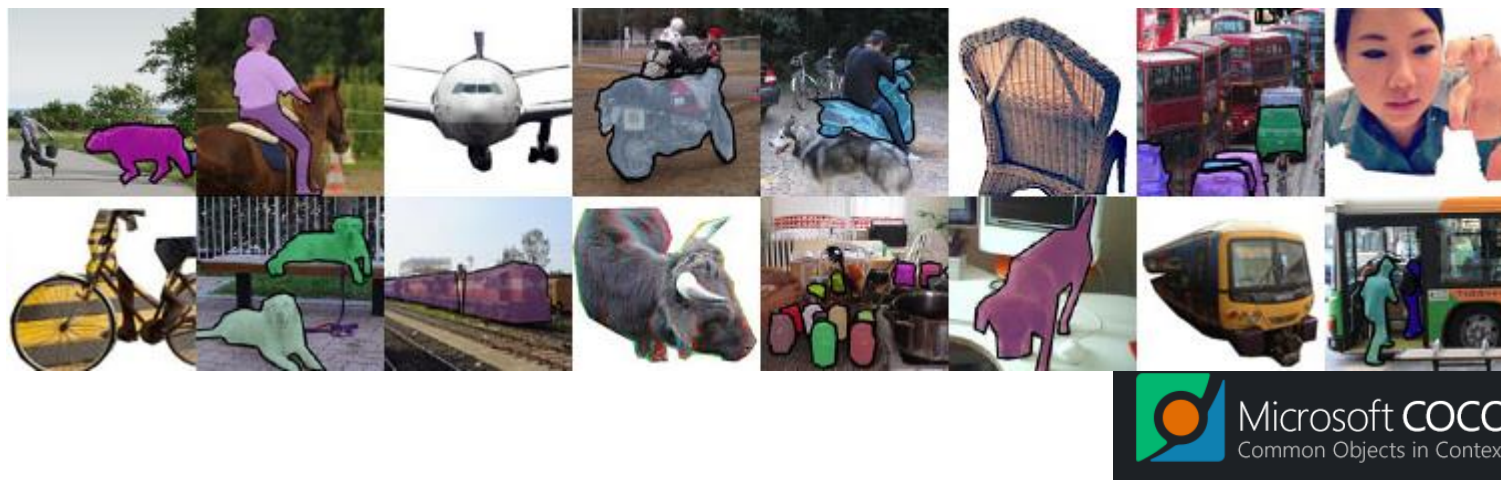
Research funded by the ERC Starting Grant VideoLearn, the German Research Foundation,
and the Deutsche Telekom Stiftung



Deep learning wants data... input and output

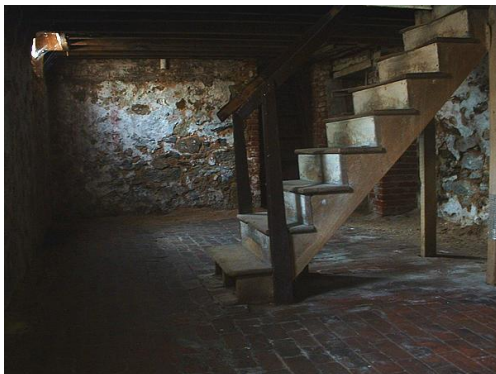


UCF 101



Typical strategy to set up a data set

1. Collect lots of images/videos
 - usually from the web
2. Manually annotate the desired output



Lock students in a
cellar



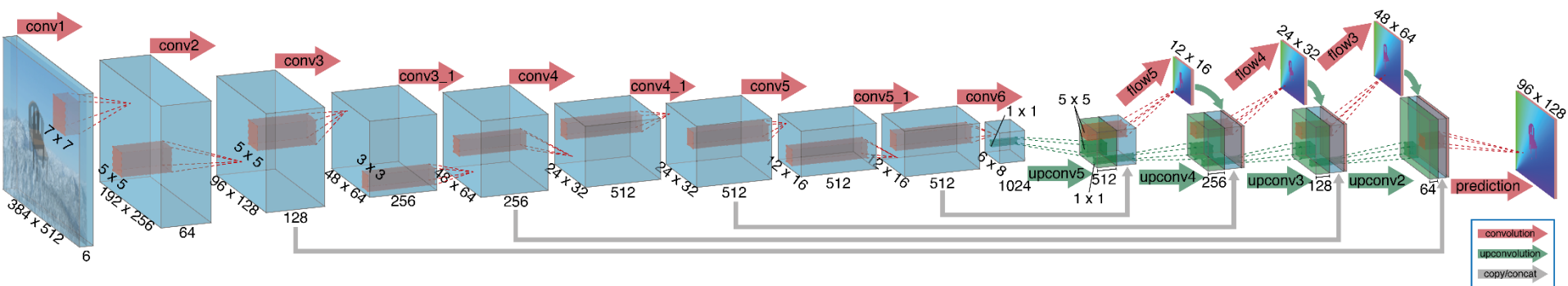
Crowd sourcing



Workers at companies

Works well for recognition

Learning tasks besides recognition: FlowNet



- Can networks learn to find correspondences?
- New learning task!
(very different from classification, etc.)

Dosovitskiy et al.
ICCV 2015

Enough data to train such a network?

- Getting ground truth optical flow for real videos is hard
- Existing datasets are small:

	Frames with ground truth
Middlebury	8
KITTI	194
Sintel	1041
Needed	>10000

Solution: large scale synthetic data

Realism is overrated: the “flying chairs” dataset

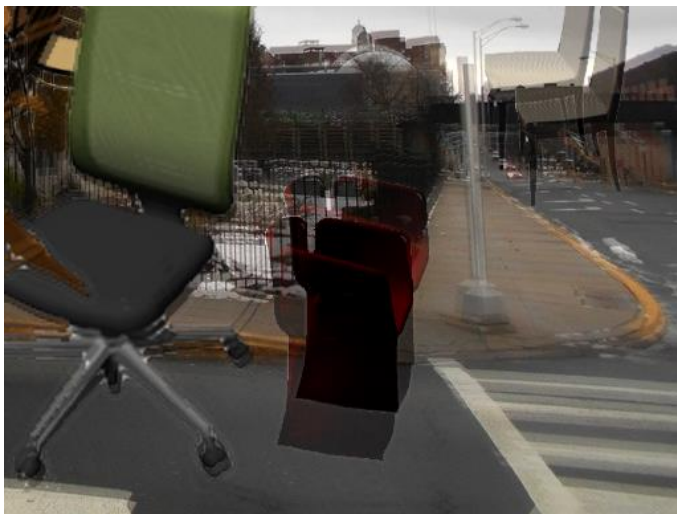


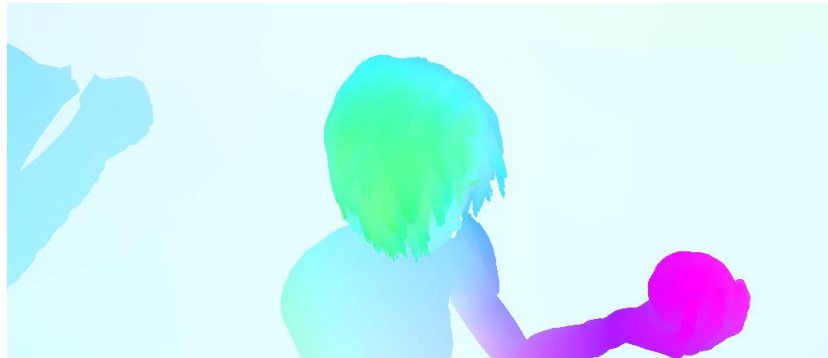
Image pair

Optical flow

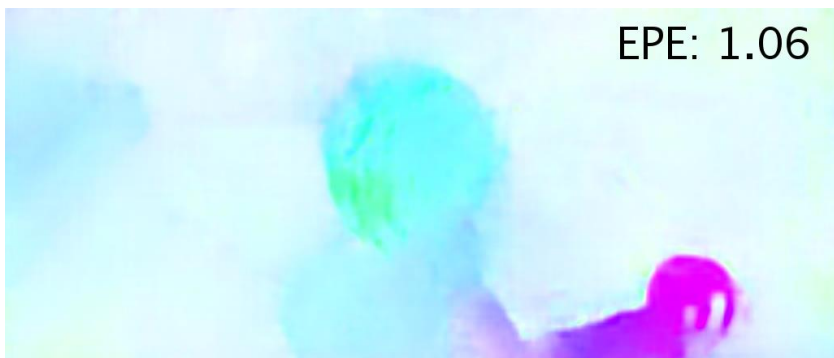
What about generalization?



Input images

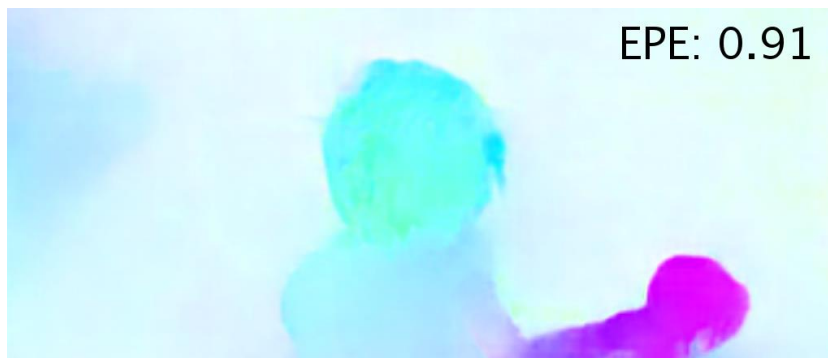


Ground truth



EPE: 1.06

FlowNetSimple



EPE: 0.91

FlowNetCorr

Although the network has only seen flying chairs for training, it predicts good optical flow on Sintel

Generalization to real data

Img1

Img2



FlowNetS

FlowNetC

FlowNet

P. Fischer,
A. Dosovitskiy,
E. Ilg,
P. Häusser,
C. Hazırbaş,
V. Golkov,
P. v.d. Smagt,
D. Cremers,
T. Brox

Can handle large displacements



Input images



Ground truth



FlowNet

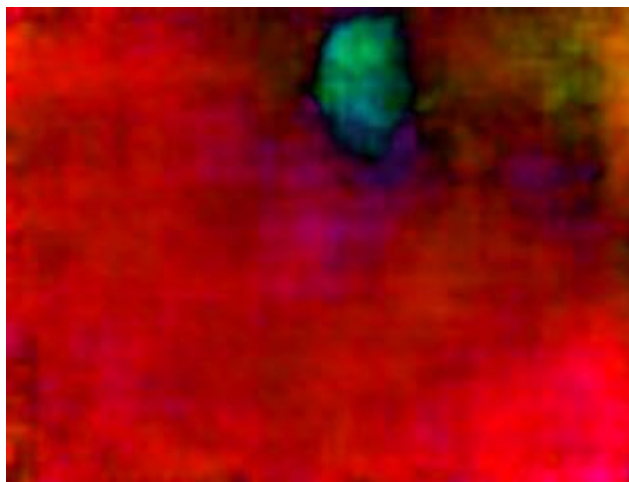


EpicFlow (Revaud et al. CVPR 2015)

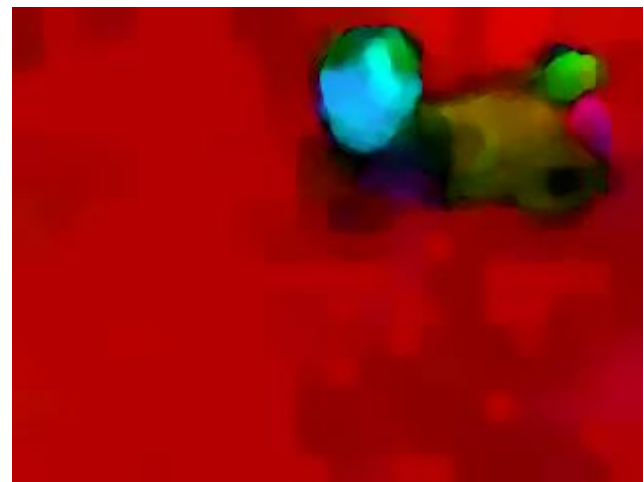
Weak spot (old FlowNet): small displacements



Example from UCF 101



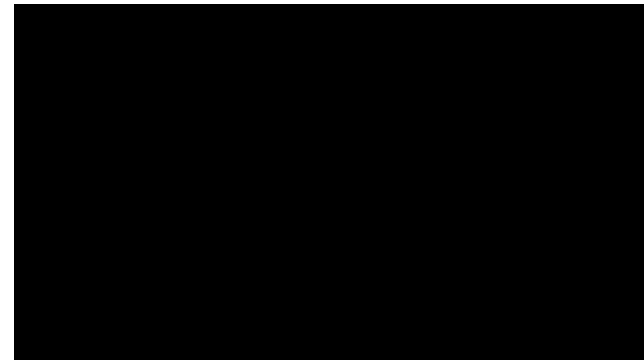
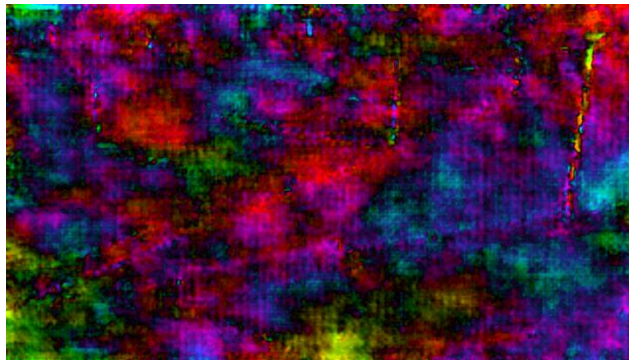
FlowNet



EpicFlow



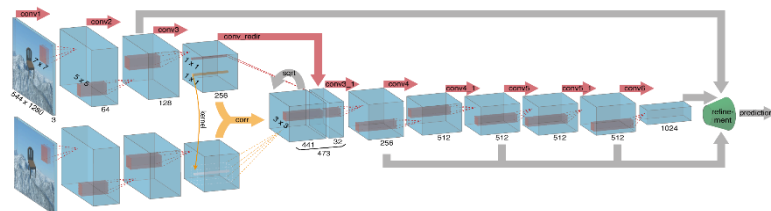
Identical images



DispNet: disparity estimation in 60ms



Mayer et al.
CVPR 2016



estimate

ground truth

DispNet: disparity estimation in 60ms



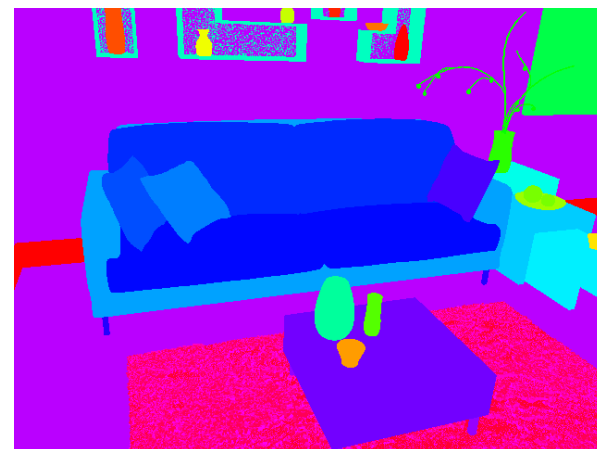
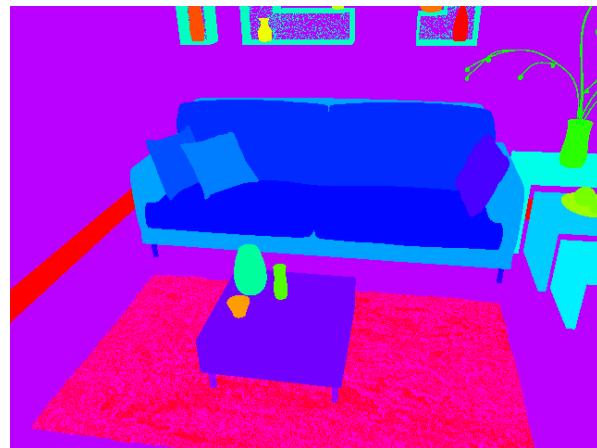
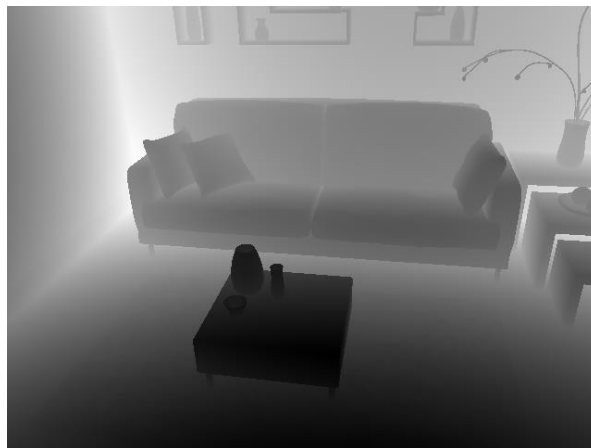
DispNet result on KITTI 2015



Driving, Monkaa, FlyingThings3D datasets publicly available

Presentation at ECCV VARVAI Workshop 16.10.2016

Powerful concept: render input and any output



Rendered input

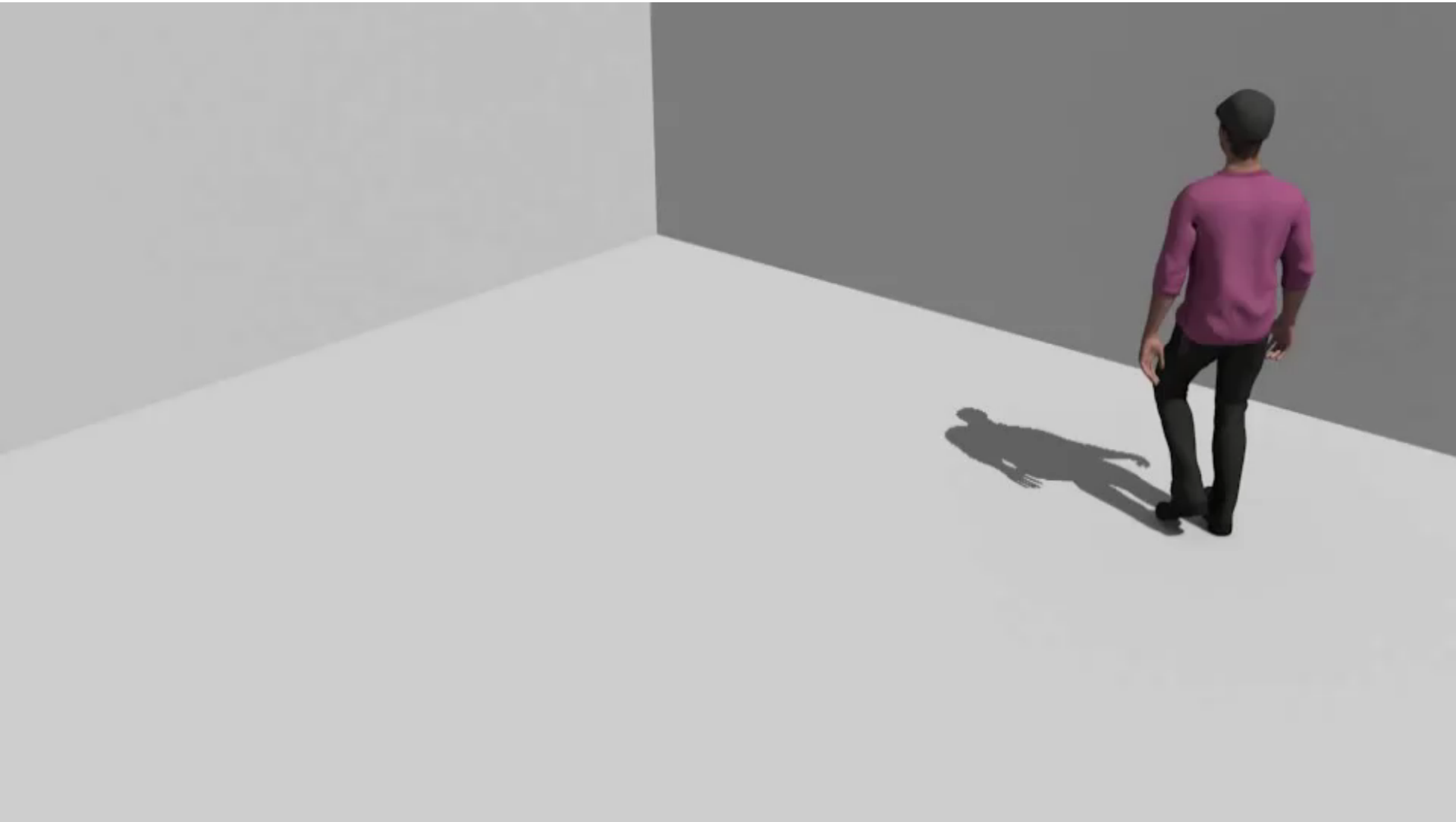
Rendered depth image

Rendered segmentation

Indoor and outdoor scenes possible



Adding animated persons



Modeling plants swinging in the wind



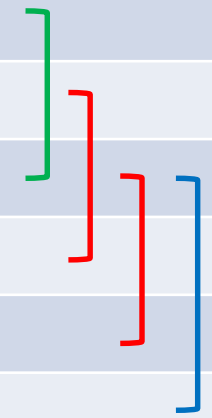
Existing scenes vs. new scenes

- Open source movies (Sintel, Monkaa), 3D Games
 - Effort: hacking the rendering engine to produce ground truth
 - Limited to the scenes that exist
- Manually created scenes
 - Full control about the scene
 - Enormous work
- Procedurally created scenes
 - Scales well, allows for diversity
 - Harder to create realistic scenes



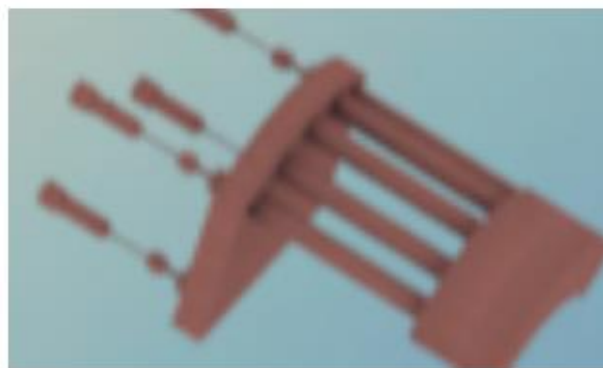
Some dataset studies on FlowNet

	Sintel Clean Validation
FlowNet, 600k on FlyingChairs	4.45
FlowNet, 1.2M on FlyingChairs	4.24
FlowNet, 1.7M on FlyingChairs	4.21
FlowNet, 1.2M on FlyingThings3D	5.07
FlowNet, 1.7M on FlyingThings3D	4.51
FlowNet, 1.2M on FlyingChairs, 500k on FlyingThings3D	3.79



- Lessons:
 - Longer training helps a little
 - FlyingChairs works better than FlyingThings3D
 - Fine-tuning on FlyingThings3D improves results

How important are realistic scenes?

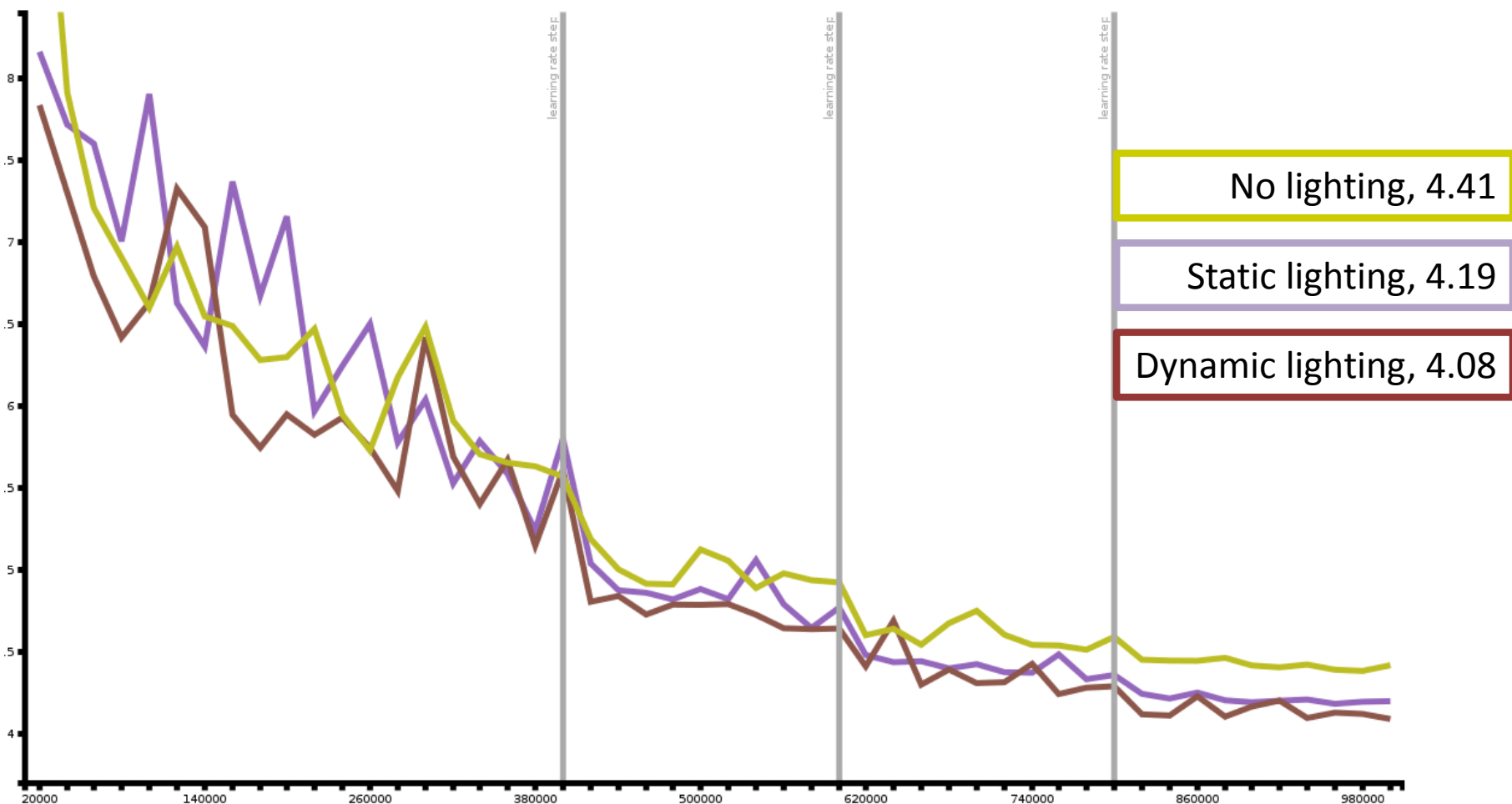


Dynamic lighting

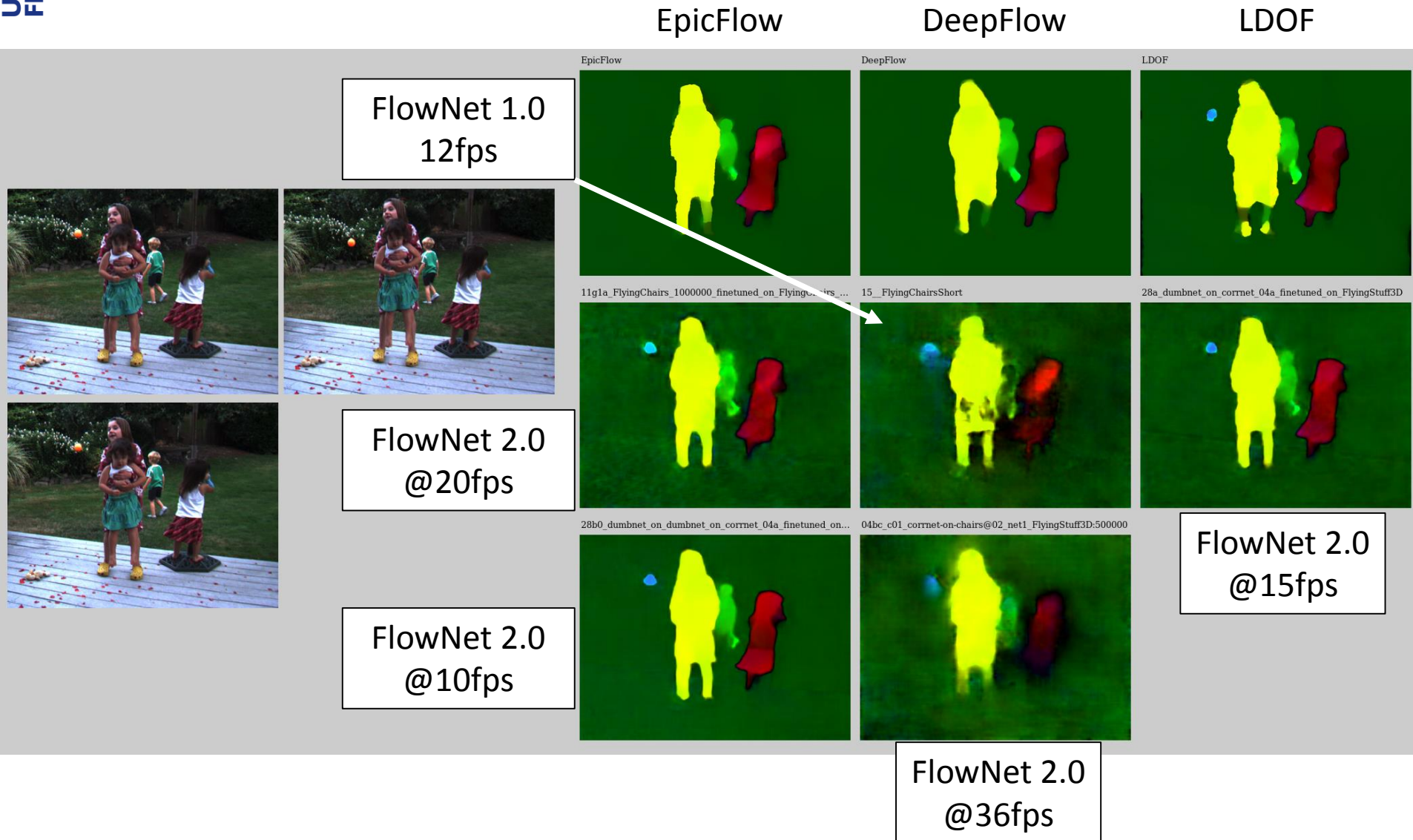
Static lighting

No lighting

How important is realistic lighting?



Breaking news: FlowNet 2.0



Good on small motion, too

EpicFlow

DeepFlow

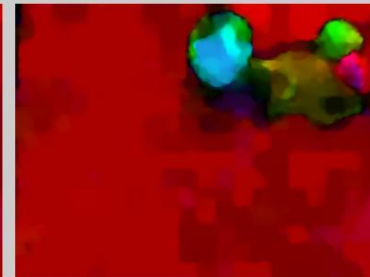
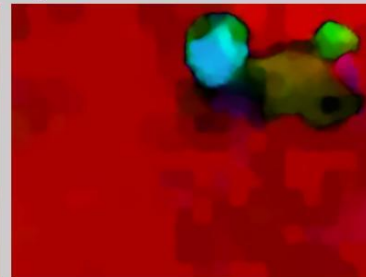
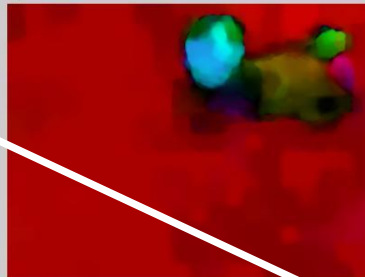
LDOF

FlowNet 1.0
12fps

EpicFlow

DeepFlow

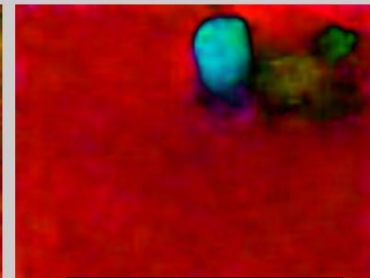
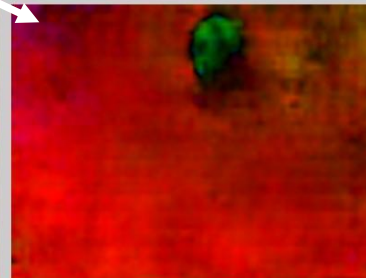
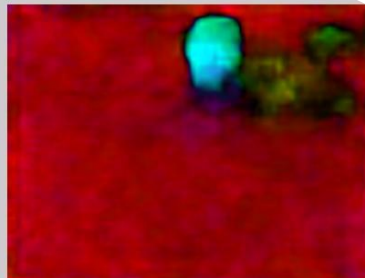
LDOF



11g1a_FlyingChairs_1000000_finetuned_on_FlyingChairsShort

15_FlyingChairsShort

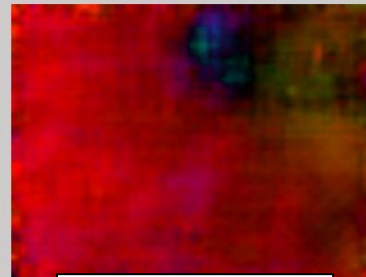
28a_dumbnet_on_corrnet_04a_finetuned_on_FlyingStuff3D



FlowNet 2.0
@20fps

28b0_dumbnet_on_dumbnet_on_corrnet_04a_finetuned_on...

04bc_c01_corrnet-on-chairs@02_net1_FlyingStuff3D:500000



FlowNet 2.0
@15fps

FlowNet 2.0
@10fps

FlowNet 2.0
@36fps

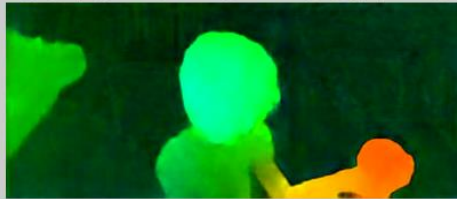
Good on small motion, too



EpicFlow



11g1a_FlyingChairs_1000000_finetuned_on_FlyingChairs_...



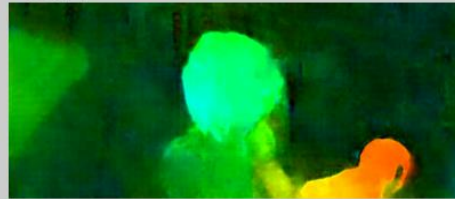
28b0_dumbnet_on_dumbnet_on_cornet_04a_finetuned_on...



DeepFlow



15_FlyingChairsShort



04bc_c01_cornet-on-chairs@02_net1_FlyingStuff3D:500000



LDOF



28a_dumbnet_on_cornet_04a_finetuned_on_FlyingStuff3D

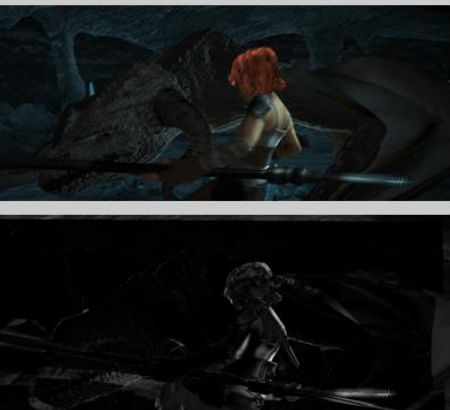


ground truth



EpicFlow	DeepFlow	LDOF
@20fps	Old @12fps	@15fps
@10fps	@36fps	Ground truth

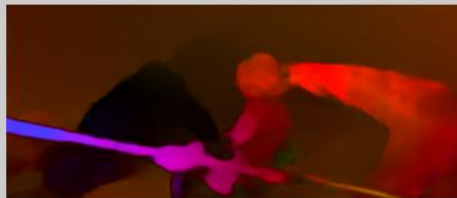
Super precise at small details



EpicFlow



11g1a_FlyingChairs_1000000_finetuned_on_FlyingChairs_...



28b0_dumbnet_on_dumbnet_on_cornet_04a_finetuned_on...



DeepFlow



15_FlyingChairsShort



04bc_c01_cornet-on-chairs@02_net1_FlyingStuff3D:500000



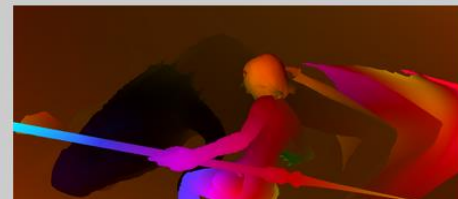
LDOF



28a_dumbnet_on_cornet_04a_finetuned_on_FlyingStuff3D

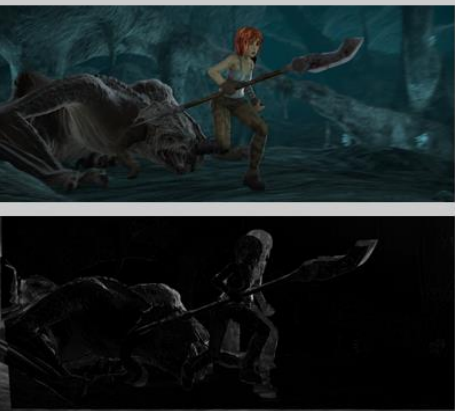


ground truth

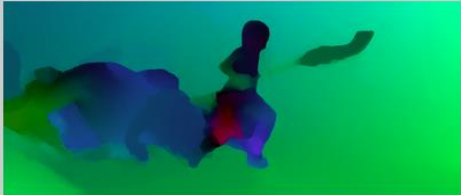


EpicFlow	DeepFlow	LDOF
@20fps	Old @12fps	@15fps
@10fps	@36fps	Ground truth

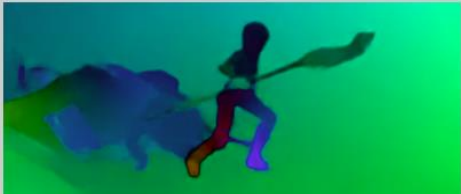
Super precise at small details



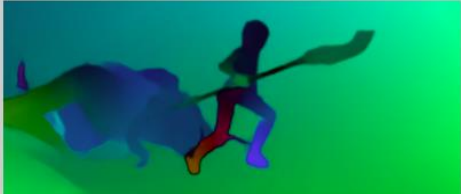
EpicFlow



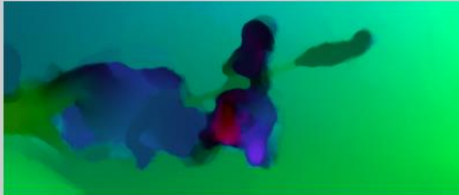
11g1a_FlyingChairs_1000000_finetuned_on_FlyingChairs_...



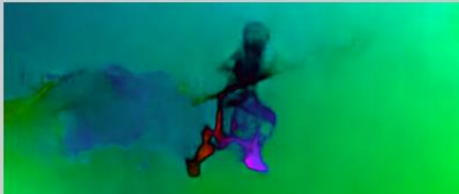
28b0_dumbnet_on_dumbnet_on_corrnet_04a_finetuned_on...



DeepFlow



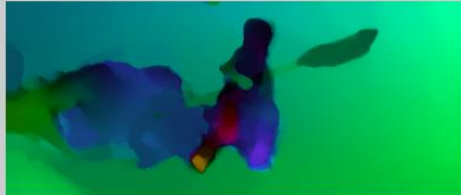
15_FlyingChairsShort



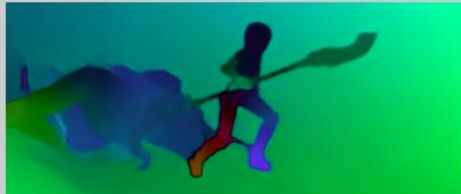
04bc_c01_corrnet-on-chairs@02_net1_FlyingStuff3D:500000



LDOF



28a_dumbnet_on_corrnet_04a_finetuned_on_FlyingStuff3D



ground truth



EpicFlow	DeepFlow	LDOF
@20fps	Old @12fps	@15fps
@10fps	@36fps	Ground truth

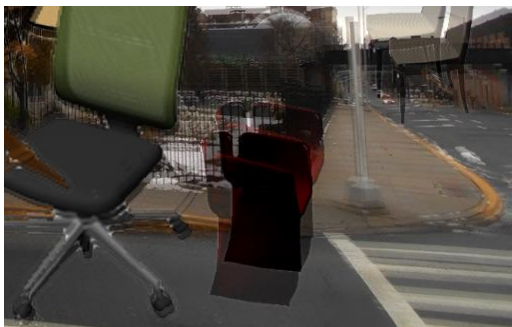
Preliminary numbers on Sintel.train.clean

	EPE	runtime
EpicFlow (Revaud et al. 2015)	2.4	15-20s
FlowNet 1.0 (Dosovitskiy et al. 2015)	4.5	80ms (12 fps)
FlowNet 2.0	2.1	90ms (11 fps)
FlowNet 2.0	2.6	28ms (36 fps)
FlowNet 2.0	4.5	9ms (110 fps)

Sintel was not used for training

(Training on FlyingChairs and FlyingThings3D)

Summary



Realism is “only” important for learning good priors



There are many ways to create synthetic training data



Procedurally generated datasets are important for diversity



FlowNet 2.0 with up to 100fps and very precise flow to come