

21st Vienna



Deep Learning
Meetup

15th October 2018 @ Marx Palast

#VDLM



Vienna Deep Learning Meetup

The Organizers:



Alex Schindler
AIT & TU Wien



René Donner
Contextflow



Thomas Lidy
Musimap



Jan Schlüter
OFAI & UTLN



Vienna Deep Learning Meetup



Agenda:

- Welcome
- Introduction by Anexia (Alexander Hendler, Project Manager)
- Deep Learning for Music & Audio Analysis (Thomas Lidy, Musimap, Alexander Schindler, AIT)
- Drum Transcription via Joint Beat and Drum Modeling using Convolutional Recurrent Neural Networks (Richard Vogl, TU Wien & CP, JKU Linz)
 - *30 minutes break*
- Latest News & Hot Topics (Thomas Lidy, Jan Schlüter, Alexander Schindler, René Donner)
- Networking and Discussions

VDLM T-Shirts

anexia

THE DENT



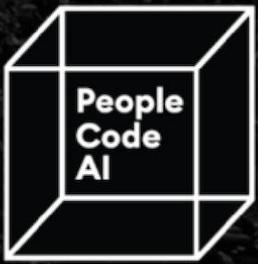
THE RIPPLE



Vienna
Deep
Learning
Meetup

The logo consists of the word "Vienna" in a bold, blue, sans-serif font. To the right of "Vienna" is a stylized graphic element: a blue circle connected by a line to a blue dot, which is then connected to another blue dot, forming a partial loop or a chain-like structure.

Announcements



WeAreDevelopers AI Congress Vienna



4-5 December, 2018



Hofburg Vienna, Austria

WeAreDevelopers

AI Congress Vienna



About the Congress

The Congress will focus on human-machine interactions and will bring together two sides: The academy and the industry. We will try to answer questions such as: Can we trust computer decisions? How to deal with decisions bias? How can we improve the user experience for machine learning software? And many, many more.

The Congress revolves around the interaction between wo/man and machines. From trusting and rationalizing the black box, to improving the interface through which we communicate with the models.

4,000m²
of AI fun

Main Topics

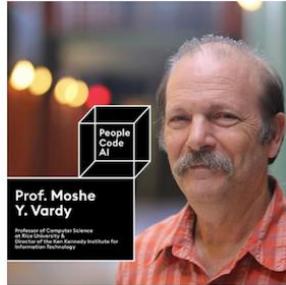
- ◎ Machine vs Human – Trusting Computers' Decisions
- ◎ Blockchain: Decentralized Artificial Intelligence
- ◎ Deep Learning – Demystifying the Black Box
- ◎ GDPR and Privacy In the Era of Big Data
- ◎ The Future of AI for Personal Usage
- ◎ Security and Safety in AI
- ◎ Ethical Challenges of AI



Confirmed Speakers for AI Congress 2018



"BIAS IN NLP"



"AN ETHICAL CRISIS IN COMPUTING?"



TOPIC
TBA



TOPIC
TBA



"HOW WE DEMOCRATIZED ARTIFICIAL INTELLIGENCE WITH ACUMOS AI"



TOPIC
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DATA NATIVES

November 22-23, Berlin

DATANATIVES.IO

Getting to Know Data Natives

Data Natives is Europe's leading events platform for the data-driven generation.

Our aim is to educate and connect our community of data enthusiasts through interactive events. We aspire to spark innovation across industries and throughout our vast network of entrepreneurs, researchers, and students. We also share cutting-edge research and thought-provoking content through our media platform, Dataconomy,

In addition to our conference, Data Natives' Big Data Meetup is well established in 47 cities across the world.



100+ events in 51 cities worldwide



35,000+ attendees



60,000 members across our communities

Speaker Highlights

06



Annina Neumann
VP DATA TECHNOLOGY
PROSIEBENSAT 1



Bart de Witte
CHAIR FACULTY OF DIGITAL
HEALTH
FUTUR/IO



Daniel Molnar
DATA ENGINEER
SHOPIFY



Cassie Korzykov
CHIEF DECISION SCIENTIST
GOOGLE



Johannes Starlinger
HEALTH DATA RESEARCHER
CHARITÉ – UNIVERSITÄTSMEDIZIN
BERLIN



Noa Tamir
DATA SCIENCE TEAM LEAD
BABBEL



Stewart Rogers
DIRECTOR OF MARKETING TECHNOLOGY
VENTUREBEAT

VDLM on Github

- history
- slides
- videos
- Wiki

| Meetups | | | | | | |
|---------|------------|------------------------------------|-------|----------------------|-------|----------------------|
| # | Date | Place | Topic | Link | Video | Meetup.com |
| 1 | 2016-04-07 | Sector 5 | intro | more | | link |
| 2 | 2016-05-09 | Sector 5 | | more | | link |
| 3 | 2016-06-06 | Sector 5 | | more | | link |
| 4 | 2016-07-07 | TU Wien | | more | | link |
| 5 | 2016-09-22 | Automic Software GmbH | | more | | link |
| 6 | 2016-10-12 | Sector 5 | | more | | link |
| 7 | 2016-12-01 | Agentur Virtual Identity | | more | | link |
| 8 | 2017-01-17 | TU Wien Informatik | | more | | link |
| 9 | 2017-02-21 | bwin.party services (Austria) GmbH | | more | | link |

| Talks | | | | |
|------------|-----|----------------------|---|---------------------|
| Date | MU# | Speaker | Topic | Slides |
| 2016-04-07 | 1 | Thomas Lidy | An overview presentation of Deep Learning | pdf |
| 2016-04-07 | 1 | Jan Schlüter | History, Approaches, Applications | pdf |
| 2016-05-09 | 2 | Alex Champandard | Neural Networks for Image Synthesis | |
| 2016-05-09 | 2 | Gregor Mitscha-Baude | Recurrent Neural Networks | pdf |
| 2016-06-06 | 3 | Jan Schlüter | Open-source Deep Learning with Theano and Lasagne | pdf |
| 2016-09-22 | 5 | Josef Puchinger | Deep Learning & The Future of Automation | |
| 2016-09-22 | 5 | Christoph Körner | Going Deeper with GoogleNet and CaffeJS | pdf |

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dychief update photos Latest commit [2 days ago](#)

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README.md fixes 21 days ago

README.md



Overview

Deep Learning is currently a big & growing trend in data analysis and prediction - and the main fuel of a new era of AI. Google, Facebook and others have shown tremendous success in pushing image, object & speech recognition to the next level.

But Deep Learning can also be used for so many other things! The list of application domains is literally endless.

Although rooted in Neural Network research already in the 1950's, the current trend in Deep Learning is unstoppable, and new approaches and improvements are presented almost every month.

<https://github.com/vdlm/meetups>

VLDL Wiki

Getting Started in Deep Learning

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These are the recommended links for getting into Deep Learning:

Video for a really basic overview: Neural Networks demystified: <https://www.youtube.com/watch?v=bxe2T-V8XR>

A quick video intro to Deep Learning: <https://www.youtube.com/watch?v=N4gDikie8E&list=TLGGOKMRn558jYwNzAyMjAxNw>

A Brief Overview of Deep Learning <http://yyue.blogspot.co.at/2015/01/a-brief-overview-of-deep-learning.html>

Google's Machine Learning & Deep Learning 101: https://docs.google.com/presentation/d/1kSuQyW5DtnkvZtjGjYCKfOvxzGFe2WBBy4e9Uedd9k/preview?imm_mid=0fb7e&cmp=em-data-na-na-newsitr_20171213&slide=id.g24521cd995_0_459

A Primer on Deep Learning <https://www.datarobot.com/blog/a-primer-on-deep-learning/>

Keras Tutorial: <https://keras.io/getting-started-30-seconds-to-keras> (Note: Keras is a framework on top of Tensorflow to make your life easier)

Activation Functions

Activation Functions used in (Deep) Neural Networks and their properties:

Note that there is no solid theory of how activation functions should be chosen. There are infinitely many possible activation functions. Which is the best for any given problem? It is unknown.

However, Glorot and Bengio (2010) investigated the effect of various activation functions with respect to the Vanishing Gradient problem and found these conclusions:

- The more classical neural networks with **sigmoid** or **hyperbolic tangent units** and standard initialization fare rather **poorly**, converging more slowly and apparently towards ultimately poorer local minima.
- Sigmoid activations (not symmetric around 0) should be avoided when initializing from small random weights, because they yield poor learning dynamics, with initial saturation of the top hidden layer.
- The **softmax** networks seem to be more robust to the initialization procedure than the tanh networks, presumably because of their gentler non-linearity.
- For tanh networks, the proposed normalized initialization can be quite helpful, presumably because the layer-to-layer transformations maintain magnitudes of activations (flowing upward) and gradients (flowing backward).

Conferences about Neural Networks

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Since some guys asked about conferences about neural networks I want to bring up some:

[b]NIPS[/b] Neural Information Processing Systems <https://nips.cc/>

[b]ICLR[/b] (pronounced: "i clear") International Conference on Learning Representations <http://www.iclr.cc>

[b]ICML[/b] International Conference on Machine Learning <http://www.icml.cc>

[b]ESANN[/b] European Symposium on Artificial Neural Networks, Computational Intelligence and Machine Learning <https://www.elen.ul.ac.be/esann/>

[b]ICAAI[/b] International Conference on Agents and Artificial Intelligence <http://www.icaai.org/>

[b]AAAI[/b] Association for the Advancement of Artificial Intelligence <http://www.aaai.org/>



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Books on Deep Learning

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Deep Learning Book (<http://www-labs.iro.umontreal.ca/~bengioy/dlbook/>) by Yoshua Bengio (work in progress)

Deep learning Book by Microsoft (<http://research.microsoft.com/pubs/209355/DeepLearning-NowPublishing-Vol7-SIG-039.pdf>)

Neural Networks and Deep Learning (<http://neuralnetworksanddeeplearning.com>) a free online book (very good one)

Deep Learning - A Practitioner's Approach - O'Reilly (<http://shop.oreilly.com/product/0636920035343.do>) Starting from almost 0, nevertheless covering pretty much everything to get started

Maschinelles Lernen (<https://www.amazon.de/Maschinelles-Lernen-Ethem-Alpaydin/dp/3486581147>) really good book in german not only about deep learning but machine learning

<https://github.com/vdlm/meetups/wiki>

VDLM Youtube Channel



Vienna Deep Learning Meetup 198 Abonnenten VON 198 ABOANIERT

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Ethics and Bias in Artificial Intelligence - 18th Vienna
964 Aufrufe • vor 4 Monaten gestreamt



Ethics and Bias in Artificial Intelligence - 18th Vienna
Keine Aufrufe • vor 4 Monaten



17th Vienna Deep Learning Meetup (part 2):
195 Aufrufe • vor 4 Monaten gestreamt

BELIEBTE KANÄLE



Kurzgesagt – In a Nuts...

ABONNIEREN



7-SEKUNDEN-RÄTSEL

ABONNIEREN



Dinge Erklärt – Kurzge...

ABONNIEREN

<https://www.youtube.com/ViennaDeepLearningMeetup>

Hot Topics & Latest News

a short block at every meetup
to briefly present recent papers and news

Send us contributions (tom.lidy@gmail.com)
or come with slides to do a short block yourself!

ISMIR 2018

ismir
2018

19th International Society for Music Information Retrieval Conference
September 23–27, Paris, France

(see separate slides for ISMIR 2018 Hot Topics)

Efficient Semantic Segmentation in Videos

Efficient Semantic Segmentation in Videos

- **What is Semantic Segmentation?**
- **How to we make it fast?**

What is Semantic Segmentation?

Learning Deep Representations for Semantic Image Parsing:
a Comprehensive Overview

Lili Huang, Jiefeng Peng, Ruimao Zhang, Guanbin Li, Liang Lin

School of Data and Computer Science, Sun Yat-Sen University

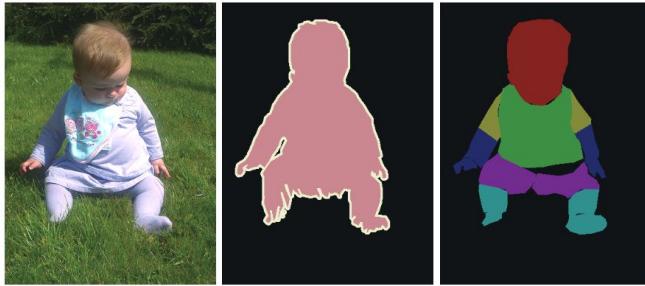


Figure 1: Illustration of the category-level semantic segmentation. The left is the original image. The middle is the basic semantic segmentation result, and the right is the semantic part segmentation result.

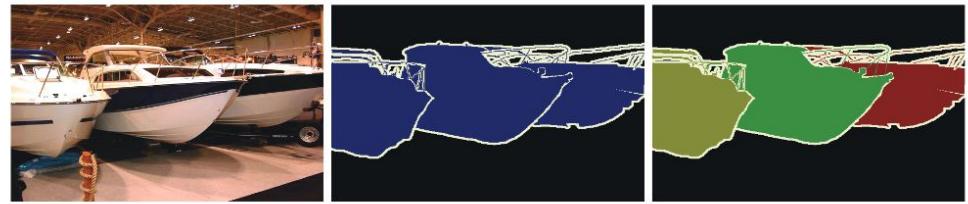
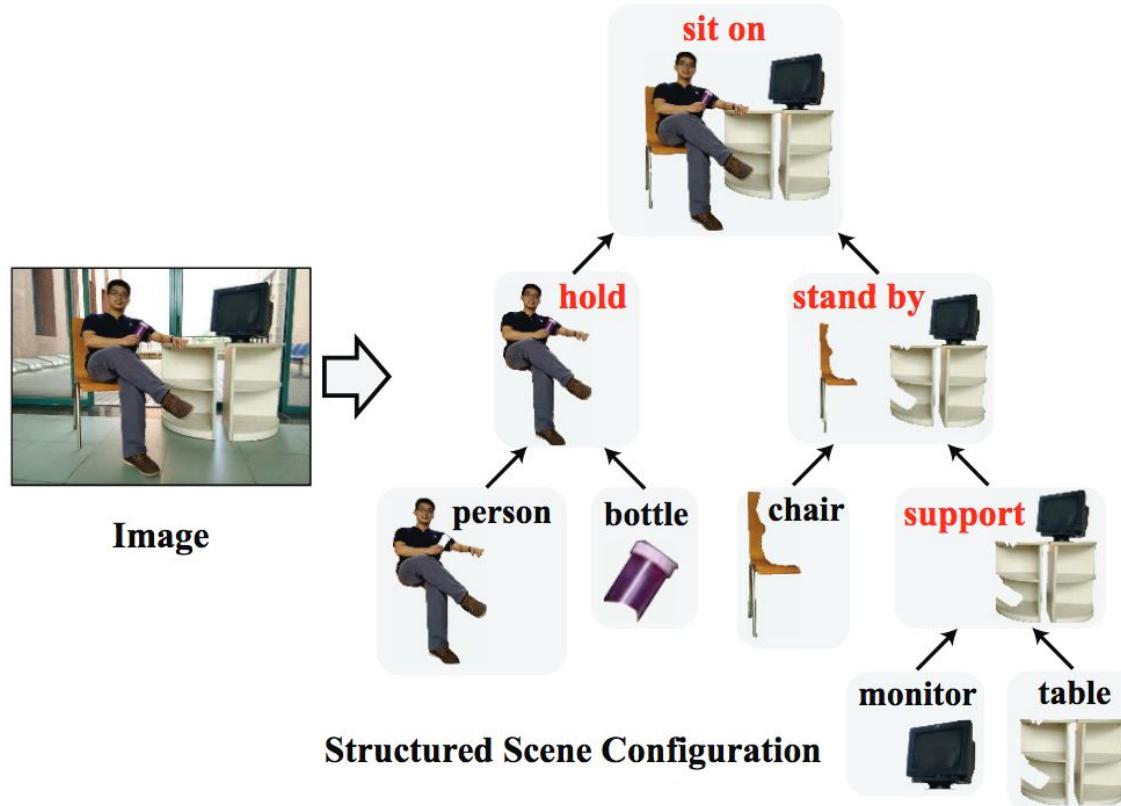
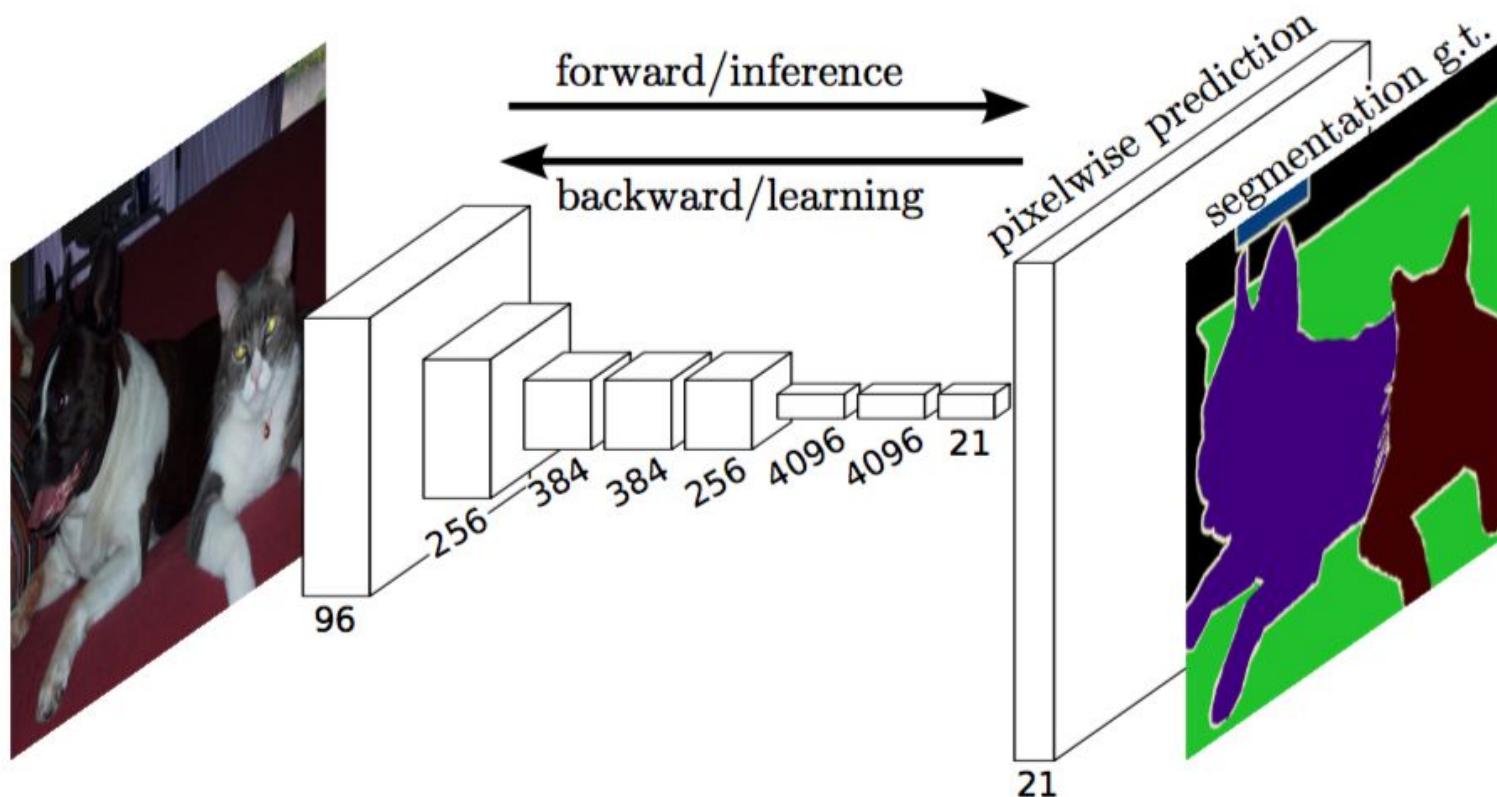


Figure 2: Comparison of category-level and instance-level semantic segmentation. The left is the original image. The middle is the category-level semantic segmentation result, and the right is the instance-level semantic segmentation result.

What is Semantic Segmentation?



How does it work?



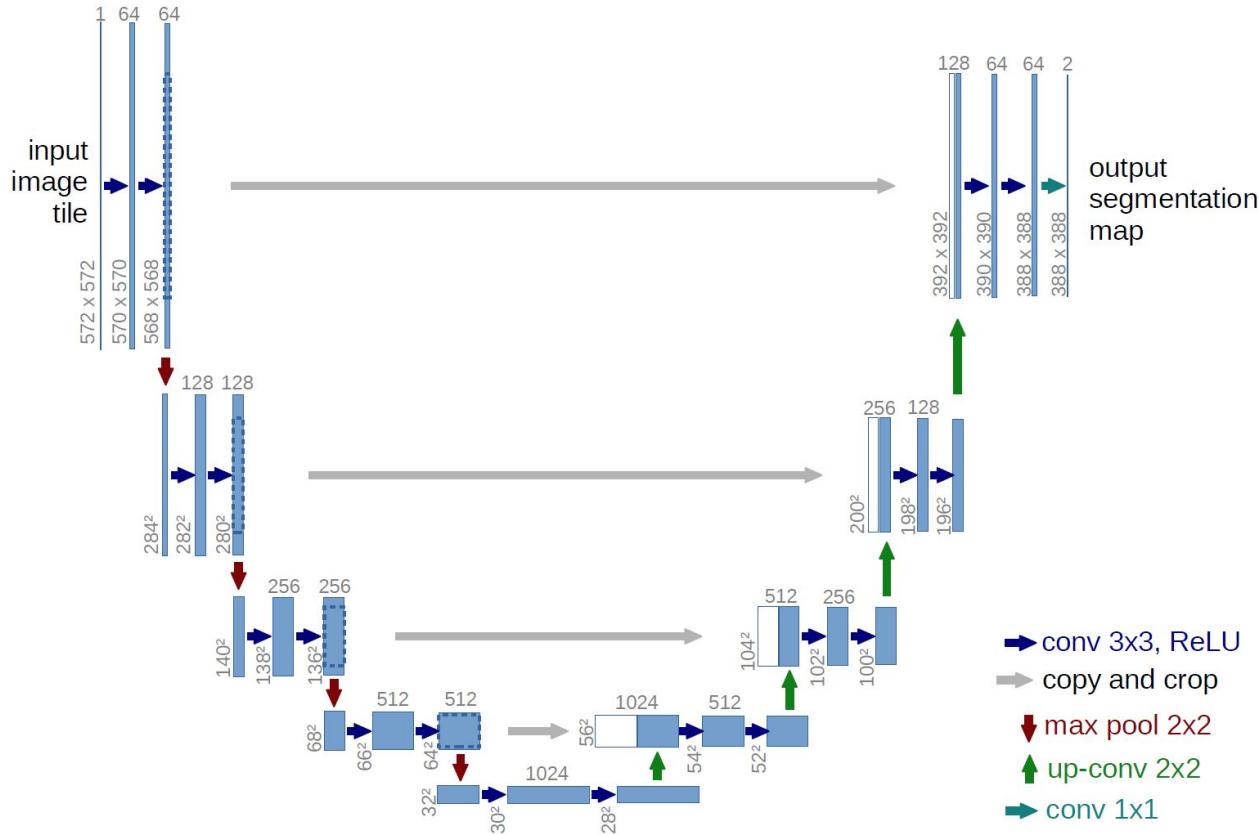
U-Nets



U-Net: Convolutional Networks for Biomedical Image Segmentation doi>

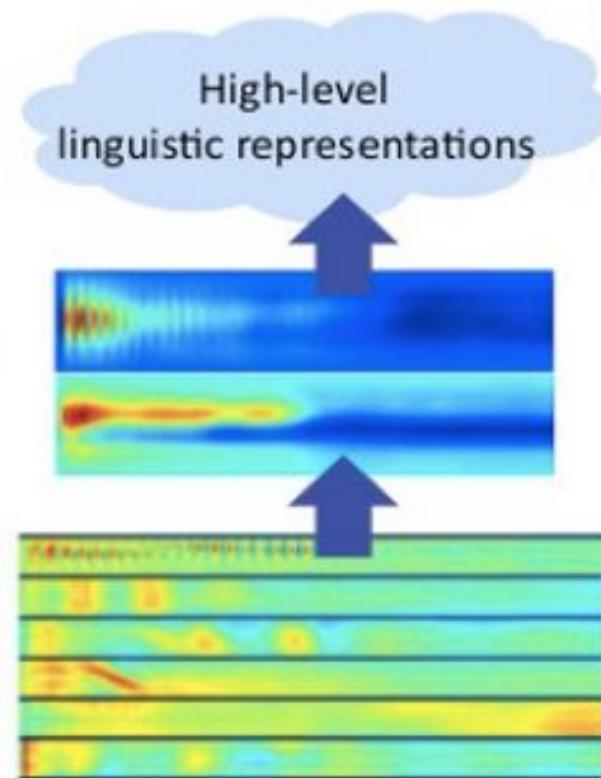
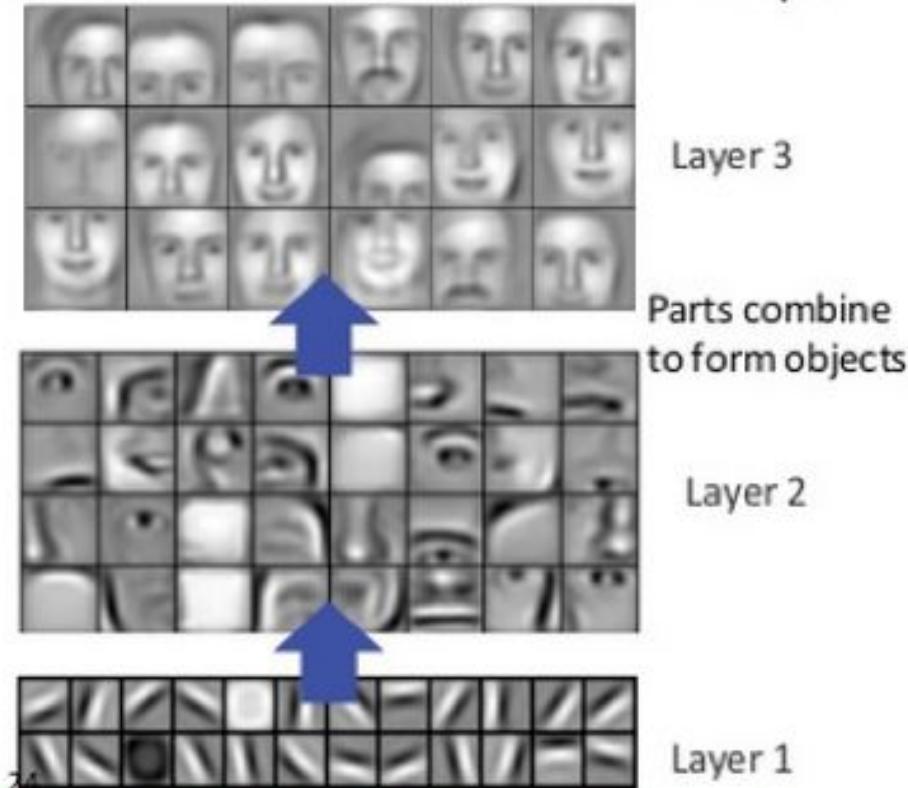
Olaf Ronneberger, Philipp Fischer, Thomas Brox

Medical Image Computing and Computer-Assisted Intervention (MICCAI), Springer, LNCS, Vol.9351: 234--241, 2015,
available at arXiv:1505.04597 [cs.CV]



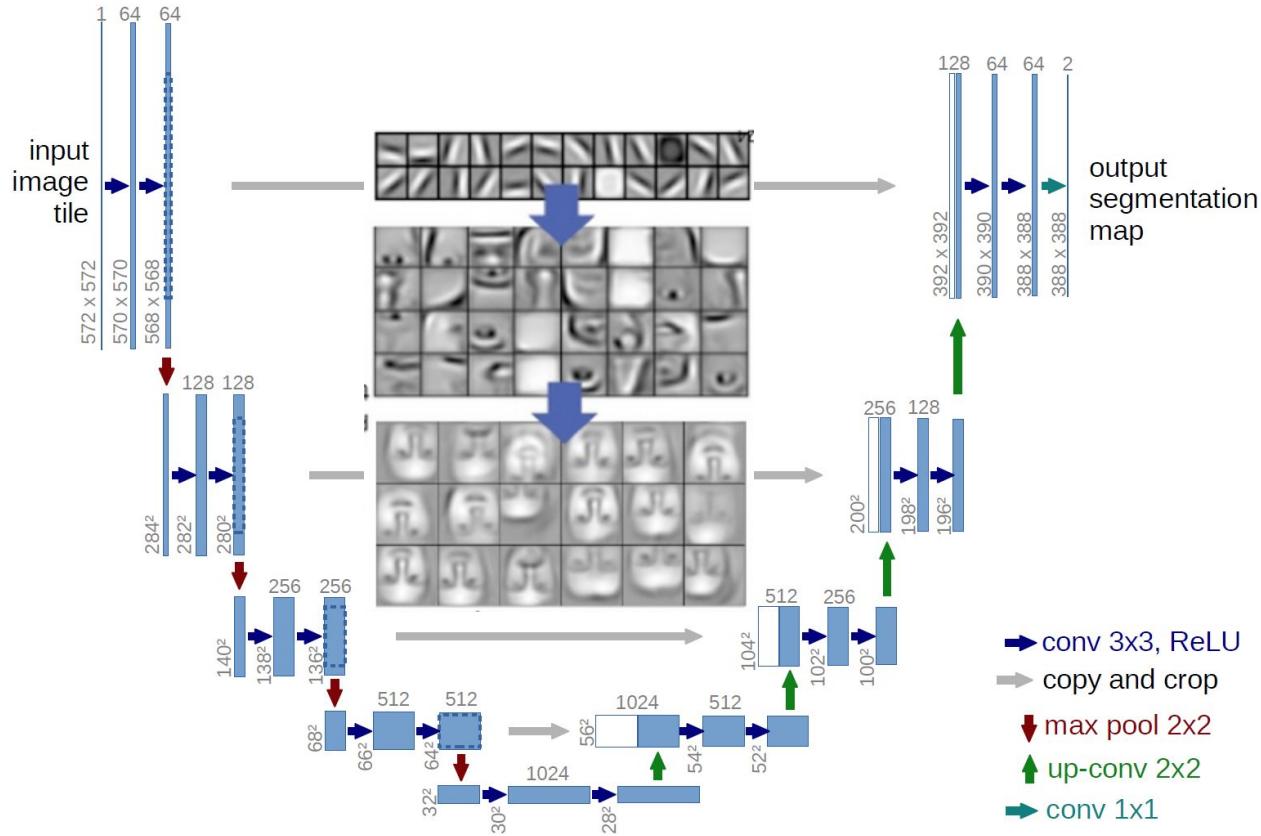
Feature hierarchy

Successive model layers learn deeper intermediate representations



Prior: underlying factors & concepts compactly expressed w/ multiple levels of abstraction

Features in a U-Net



Semantic Segmentation



Learning Deep Representations for Semantic Image Parsing: a Comprehensive Overview

Lili Huang, Jiefeng Peng, Ruimao Zhang, Guanbin Li, Liang Lin

School of Data and Computer Science, Sun Yat-Sen University

- **Approaches based on CNNs**
- **Approaches based on RNNs**
- **Instance segmentation**
- **Evaluation metrics and benchmarks**
- **Public datasets**

Efficient Semantic Segmentation in Videos

- **What is Semantic Segmentation?**
- **How to we make it fast?**
 - Reuse information from neighbouring frames!

Efficient Semantic Segmentation in Videos

■ How to we make it fast?

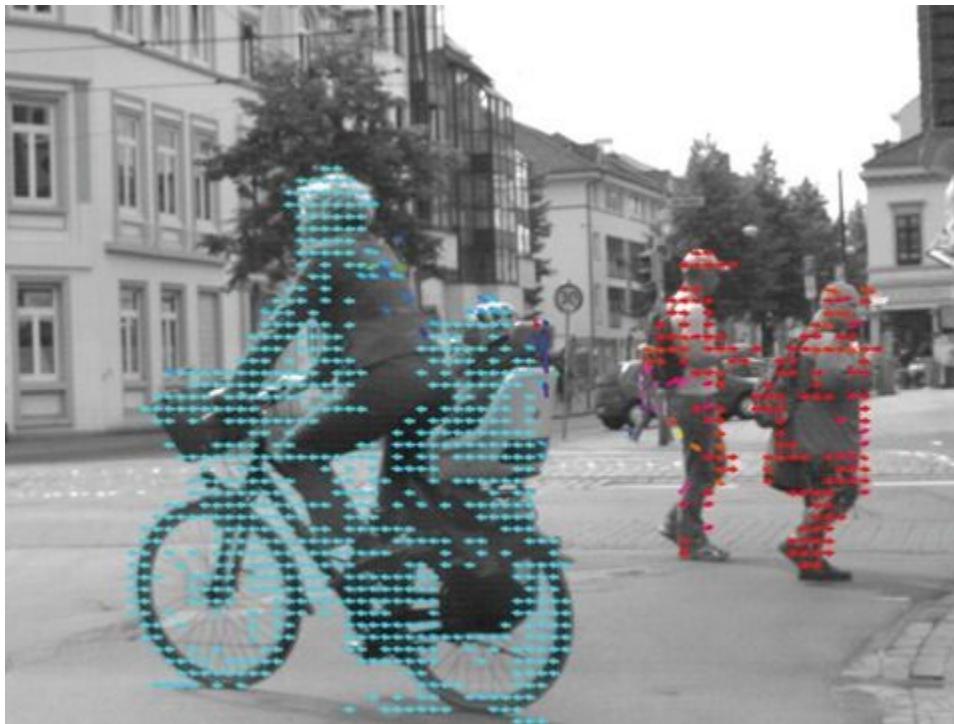
- Reuse information from neighbouring frames!
 - Using optical flow
 - [predict optical flow]
 - Reuse information from video encoding internals

Deep Feature Flow for Video Recognition

Deep Feature Flow for Video Recognition

Xizhou Zhu^{1,2*} Yuwen Xiong^{2*} Jifeng Dai² Lu Yuan² Yichen Wei²
¹University of Science and Technology of China ²Microsoft Research
ezra0408@mail.ustc.edu.cn {v-yuxio,jifdai,luyuan,yichenw}@microsoft.com

Optical Flow



Deep Feature Flow for Video Recognition

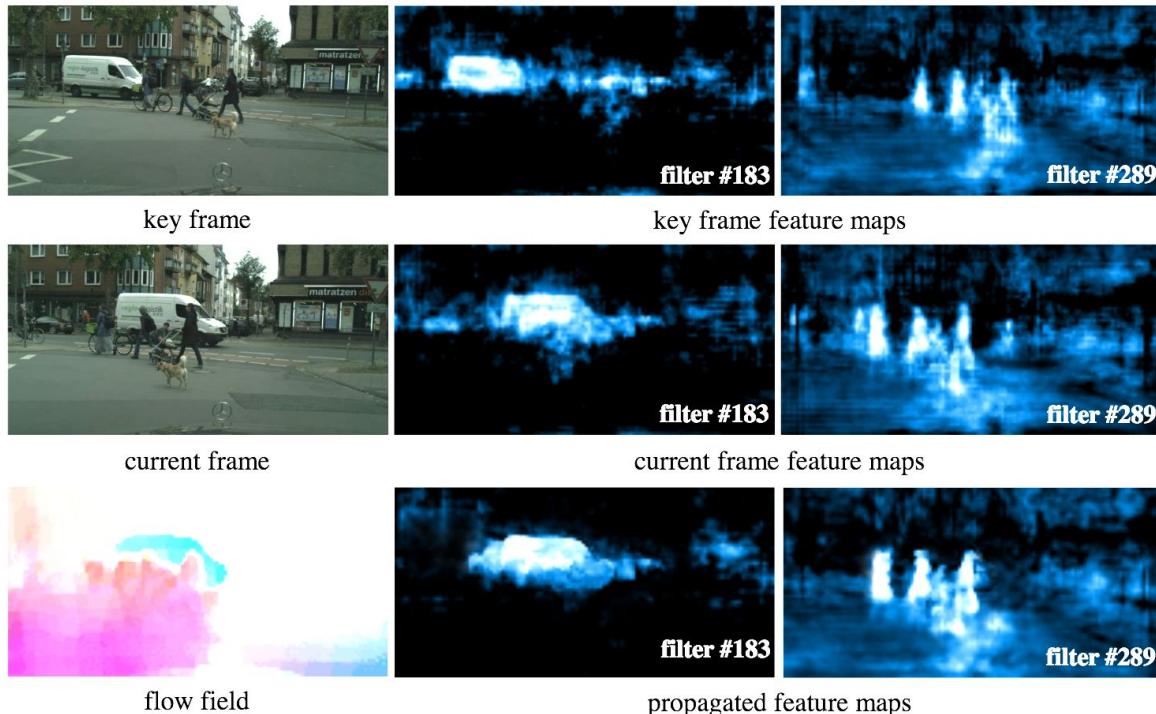
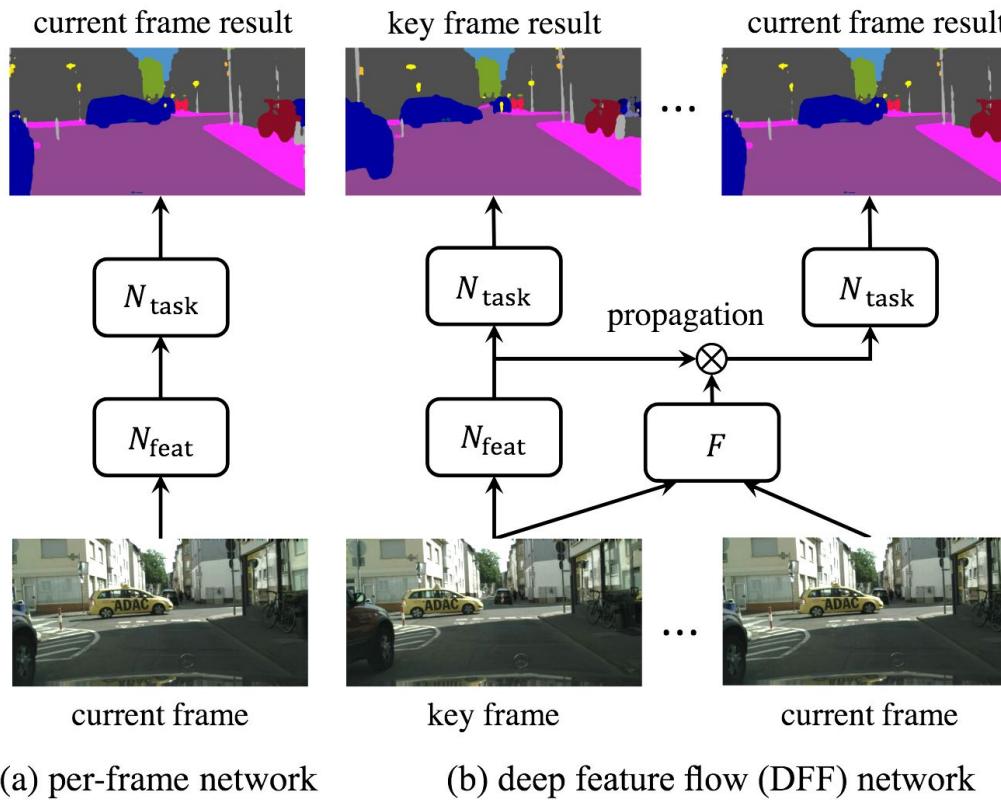
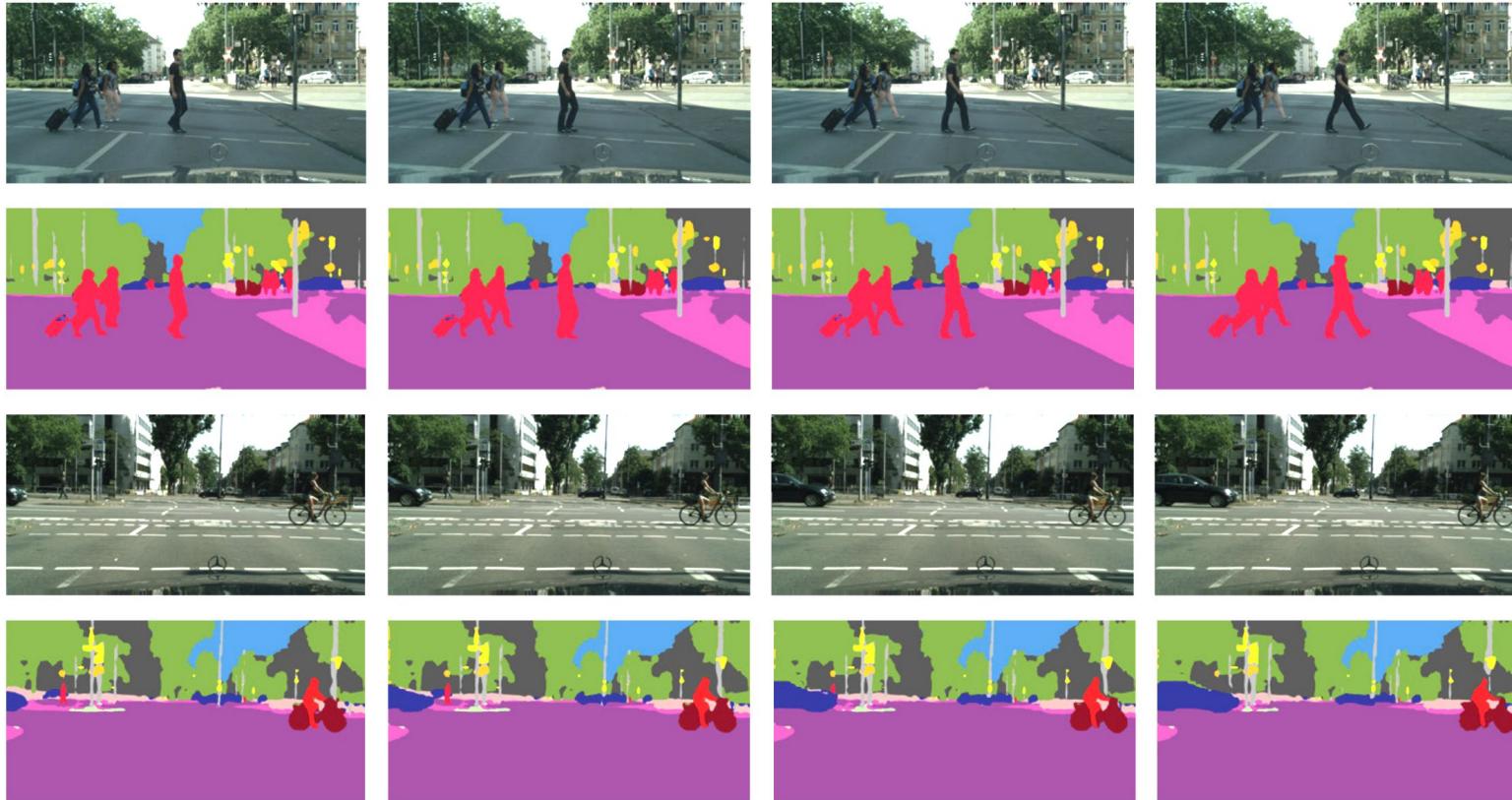


Figure 1. Motivation of proposed *deep feature flow* approach. Here we visualize the two filters' feature maps on the last convolutional layer of our ResNet-101 model (see Sec. 4 for details). The convolutional feature maps are similar on two nearby frames. They can be cheaply propagated from the key frame to current frame via a flow field.

Deep Feature Flow for Video Recognition



Deep Feature Flow for Video Recognition



Deep Feature Flow for Video Recognition



Efficient Semantic Segmentation in Videos

■ How to we make it fast?

- Reuse information from neighbouring frames!
 - Using optical flow
 - [predict optical flow]
 - Reuse information from video encoding internals

FlowNet: Learning Optical Flow with Convolutional Networks

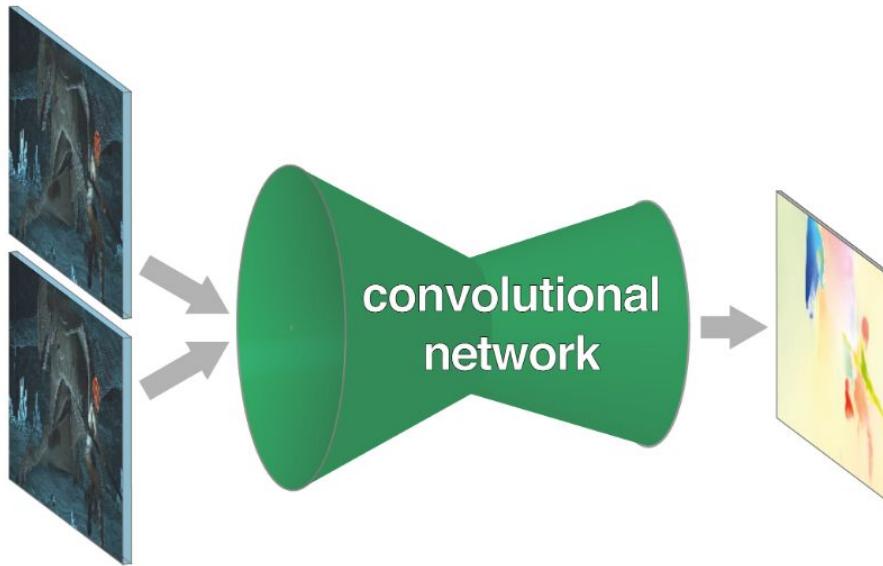
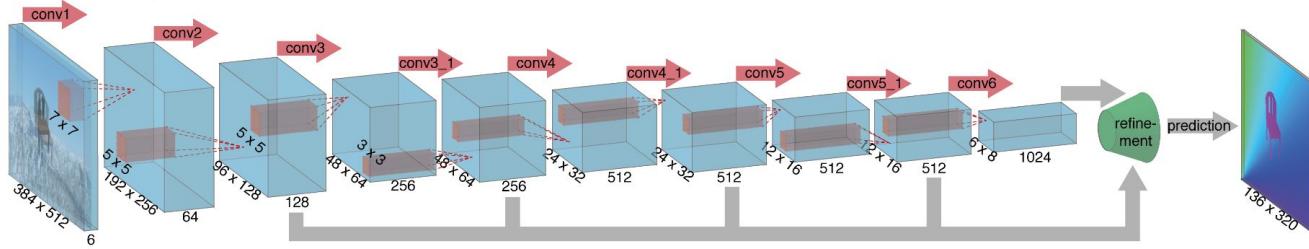


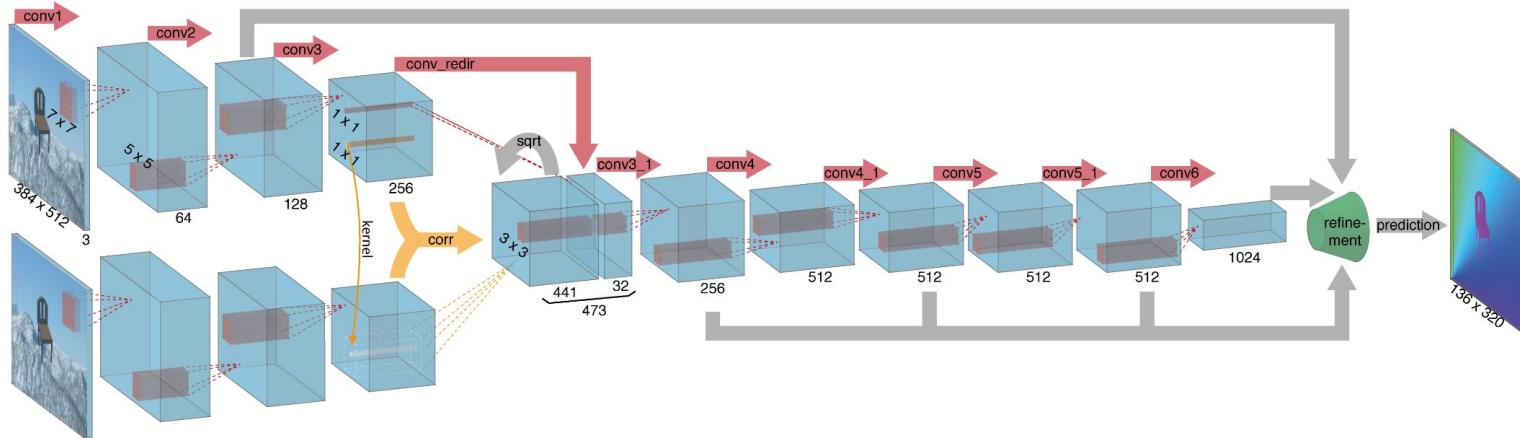
Figure 1. We present neural networks which learn to estimate optical flow, being trained end-to-end. The information is first spatially compressed in a contractive part of the network and then refined in an expanding part.

FlowNet: Learning Optical Flow with Convolutional Networks

FlowNetSimple



FlowNetCorr



FlowNet: Learning Optical Flow with Convolutional Networks

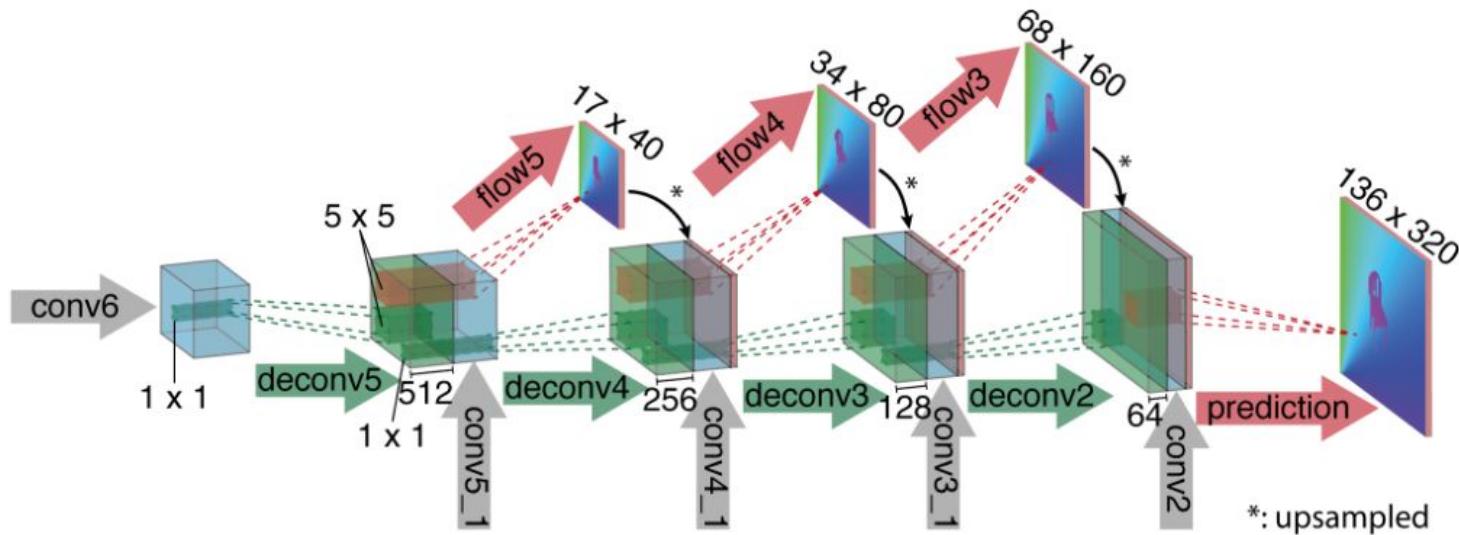
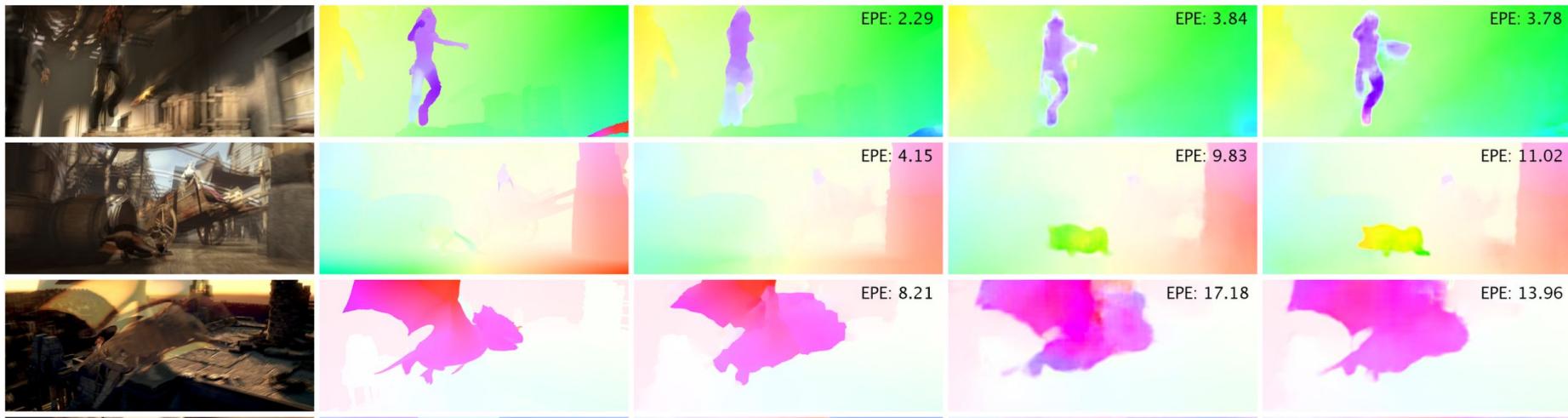


Figure 3. Refinement of the coarse feature maps to the high resolution prediction.

FlowNet: Learning Optical Flow with Convolutional Networks



Efficient Semantic Segmentation in Videos

■ How to we make it fast?

- Reuse information from neighbouring frames!
 - Using optical flow
 - [predict optical flow]
 - **Reuse information from video encoding internals**

INTER-BMV: INTERPOLATION WITH BLOCK MOTION VECTORS FOR FAST SEMANTIC SEGMENTATION ON VIDEO

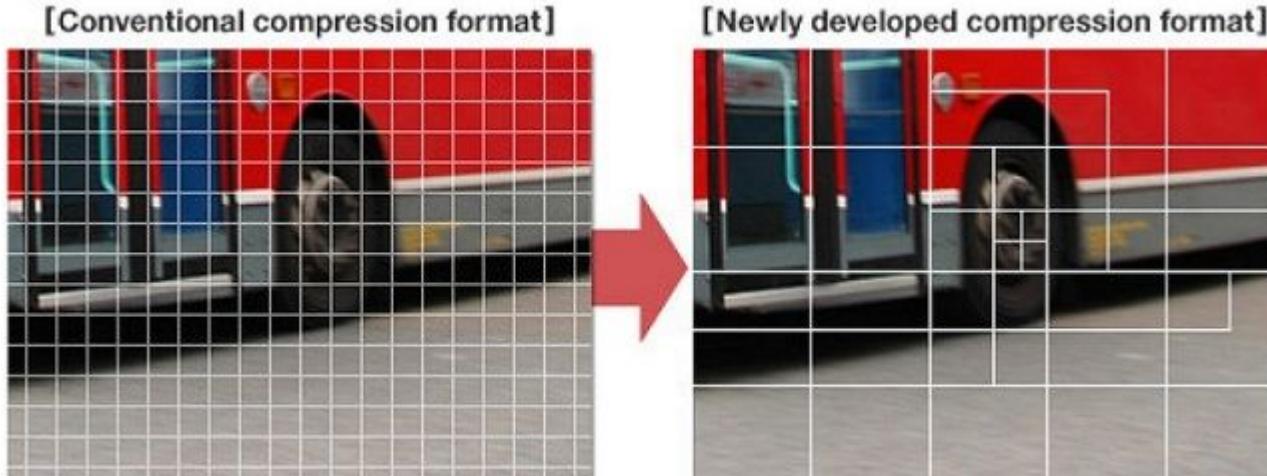
Samvit Jain & Joseph E. Gonzalez

Department of Computer Science

University of California, Berkeley

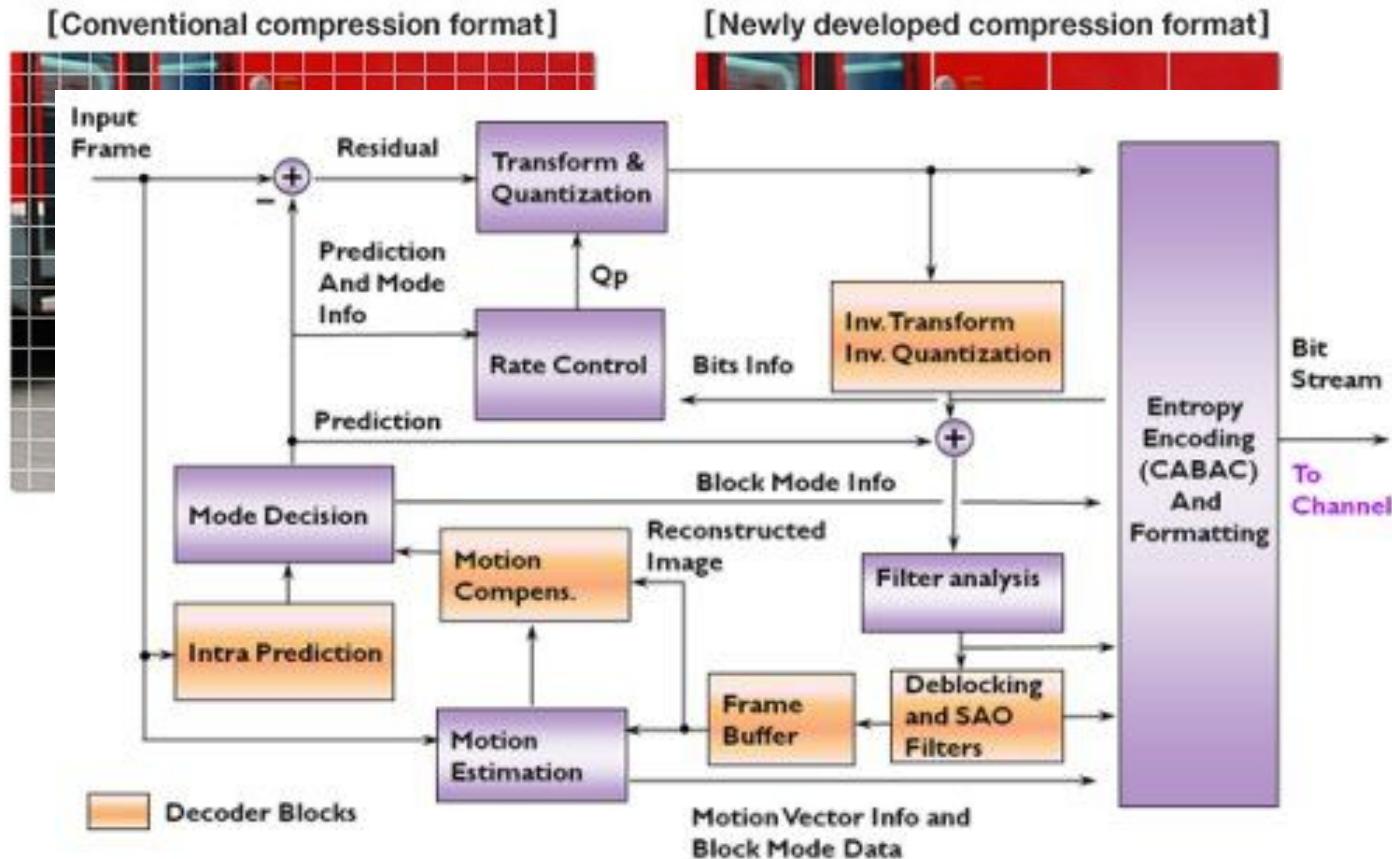
{samvit, jegonzal}@eecs.berkeley.edu

Modern video codes (e.g. h.265)

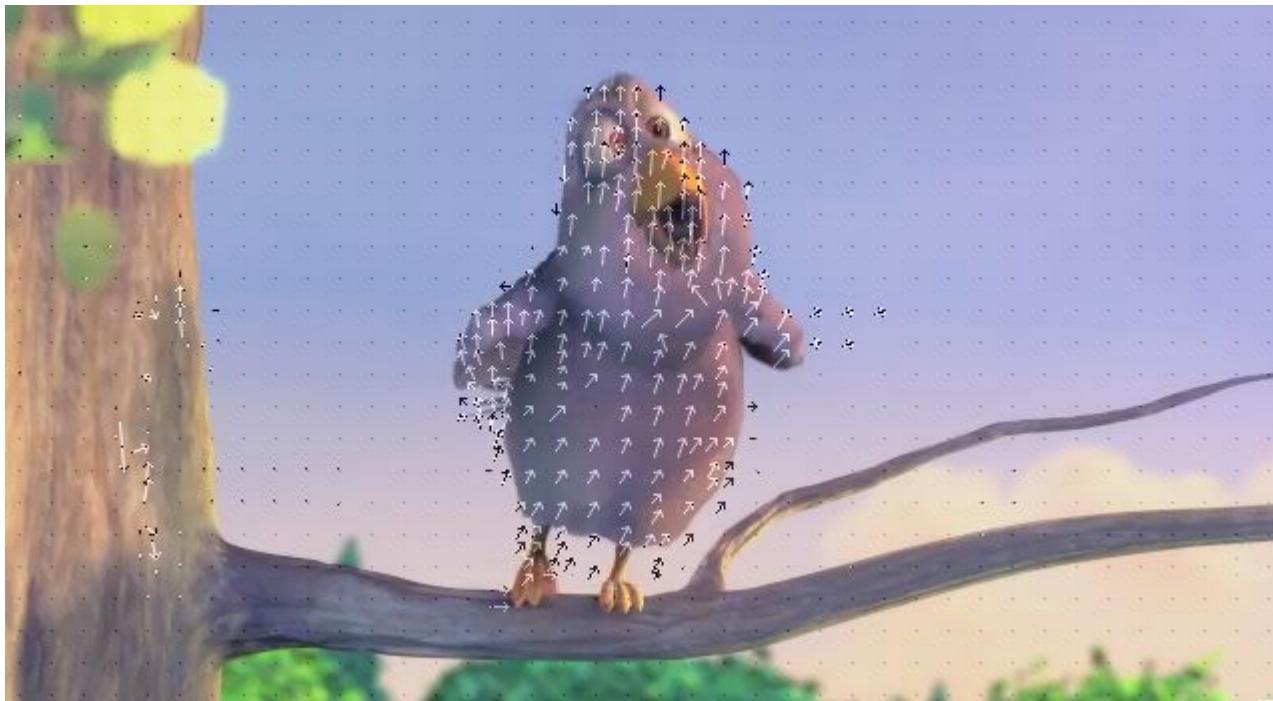


*Videos shown are rendered images.

Interpolation with block-motion vectors



Interpolation with block-motion vectors



Interpolation with block-motion vectors

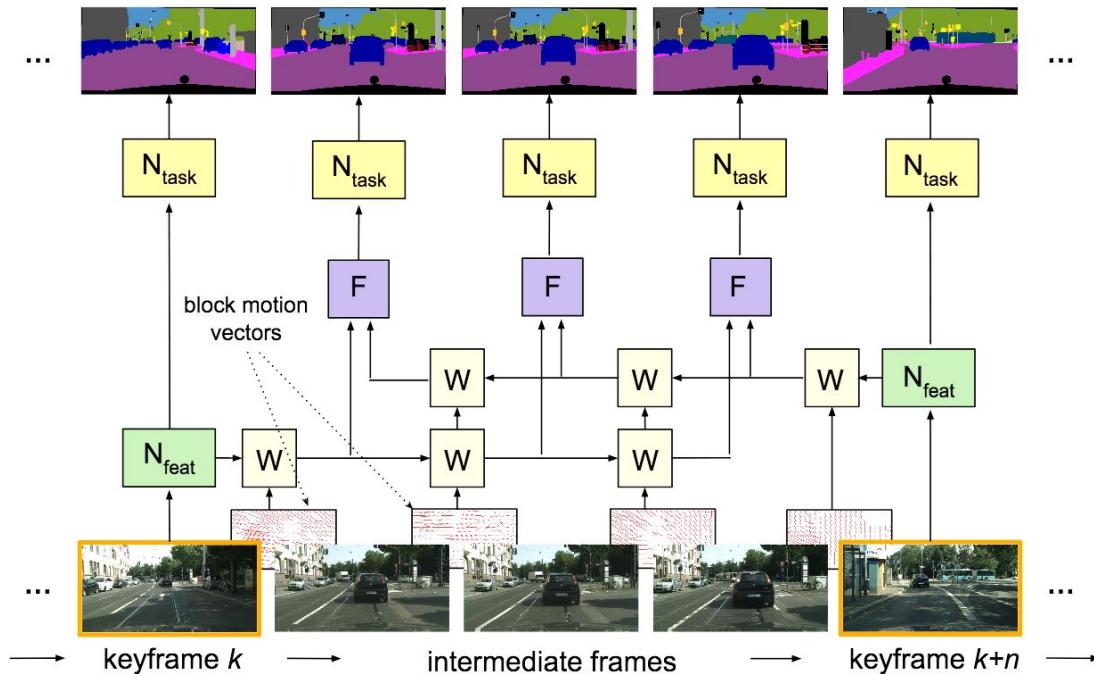


Figure 1: Inter-BMV warps and fuses the features of enclosing keyframes to generate accurate feature estimates for intermediate frames, using block motion vectors present in compressed video.

Interpolation with block-motion vectors

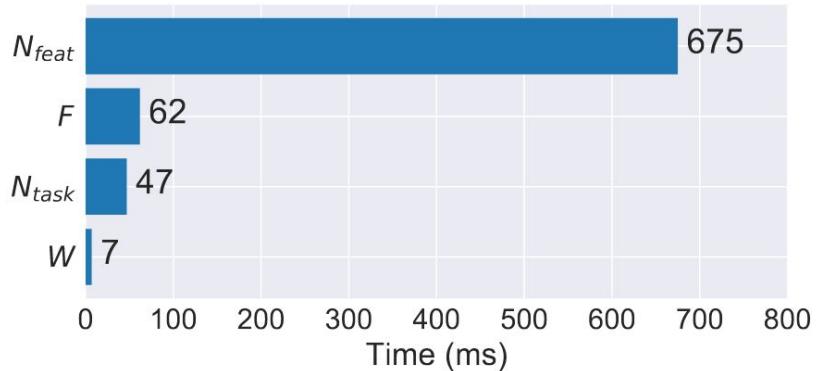
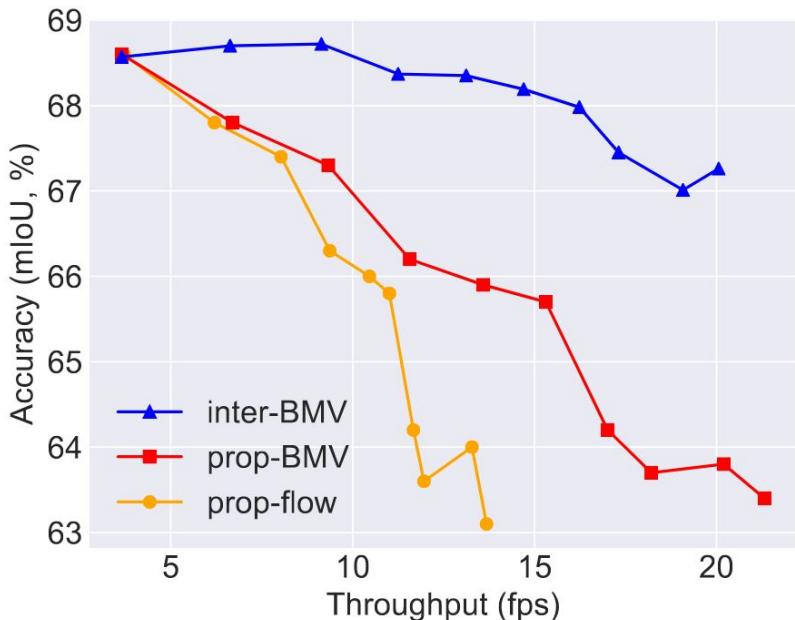
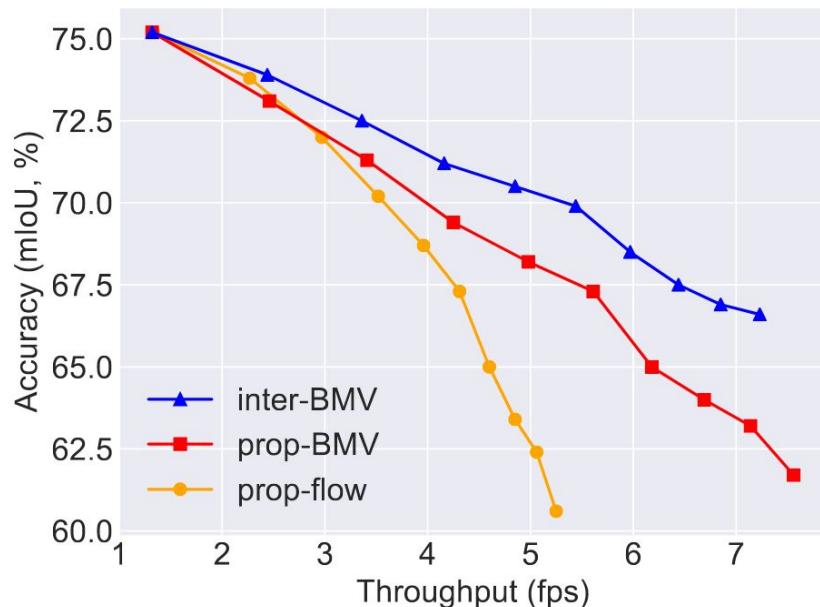


Figure 2: A sample runtime breakdown for a ResNet-101 DeepLab network. F is the optical flow net from Dosovitskiy et al. (2015). W is the warp operator. GPU: Tesla K80. Dataset: Cityscapes.

Interpolation with block-motion vectors



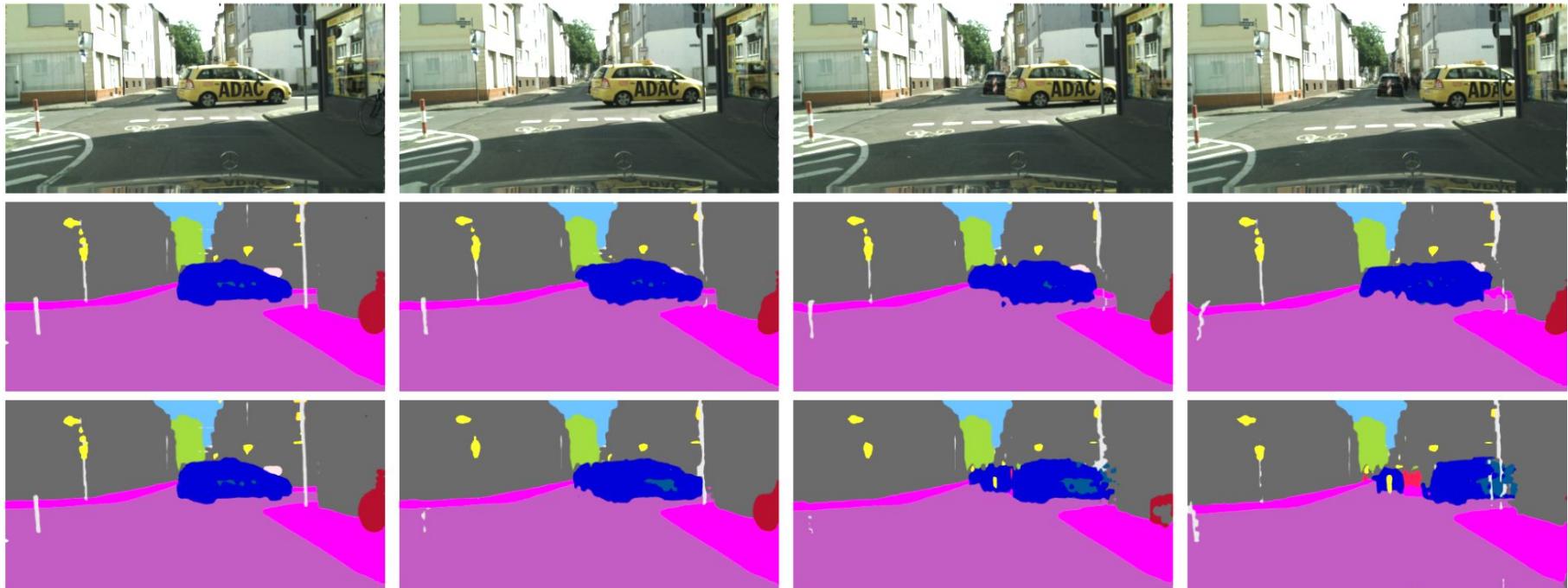
(a) **CamVid**. Data from Table 1.



(b) **Cityscapes**. Data from Table 3 (Appendix).

Figure 3: Accuracy (avg.) vs. throughput for all schemes on CamVid and Cityscapes.

Interpolation with block-motion vectors



(a) k

(b) $k+2$

(c) $k+4$

(d) $k+6$

Efficient Semantic Segmentation in Videos

21st Vienna



Deep Learning

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Next Meetup



Save the date! (sign up tomorrow)

**November 12, 2018
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