

Deep Optical Flow Networks and Applications

CS 6384 Computer Vision
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a visual scene caused by the relative motion between an observer and a scene

Lucas-Kanade Method for Optical Flow Estimation

- Brightness Constancy: the intensity or brightness of a pixel remains constant while moving from one frame to another
- Small Motion: the motion between consecutive frames is small
- Spatial Coherence: neighboring pixels have similar motion

$$I_x u + I_y v + I_t = 0$$





$$\begin{bmatrix} I_x(\mathbf{p_1}) & I_y(\mathbf{p_1}) \\ I_x(\mathbf{p_2}) & I_y(\mathbf{p_2}) \\ \vdots & \vdots \\ I_x(\mathbf{p_{25}}) & I_y(\mathbf{p_{25}}) \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = - \begin{bmatrix} I_t(\mathbf{p_1}) \\ I_t(\mathbf{p_2}) \\ \vdots \\ I_t(\mathbf{p_{25}}) \end{bmatrix}$$

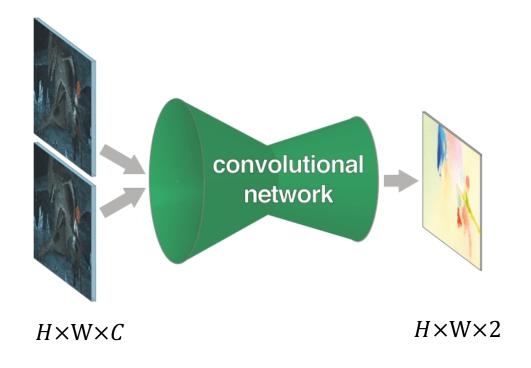
$$A \qquad d \qquad b$$

$$25 \times 2 \qquad 2 \times 1 \qquad 25 \times 1$$

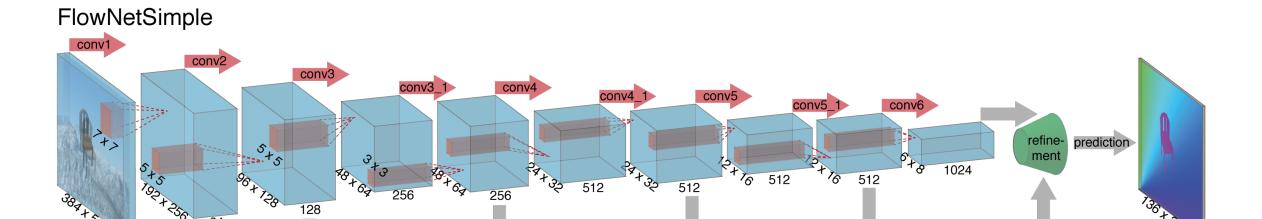
Can we use deep nets to estimate optical flow?

Estimating Optical Flow using Deep Networks

• Given two consecutive image frames: I_t and I_{t+1} , we aim to estimate the motion field (u,v) between them for each pixel



Regression Loss: L2, L1, ...

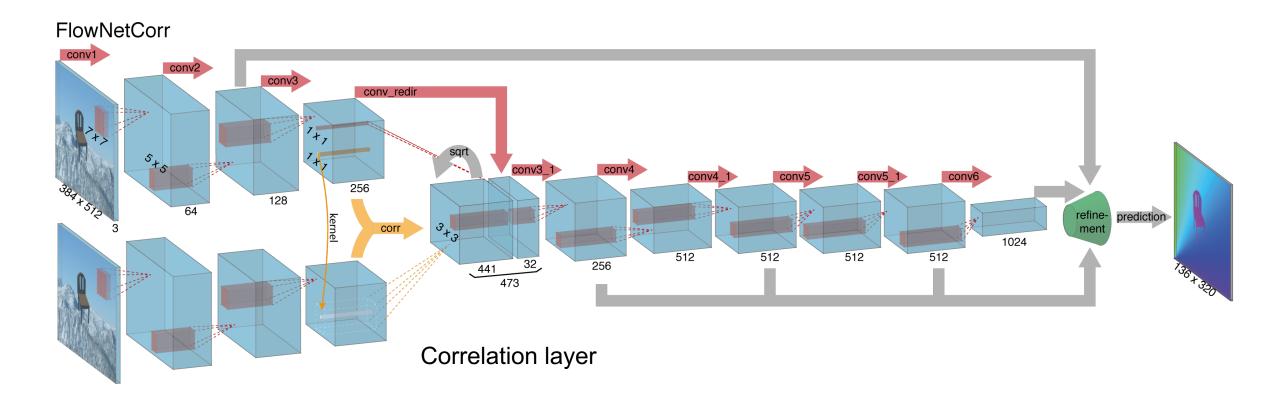


Stack two images

x-y flow fields

The architecture is similar to FCN for semantic segmentation

$$\frac{dx}{dt}$$
, $\frac{dy}{dt} = (u, v)$



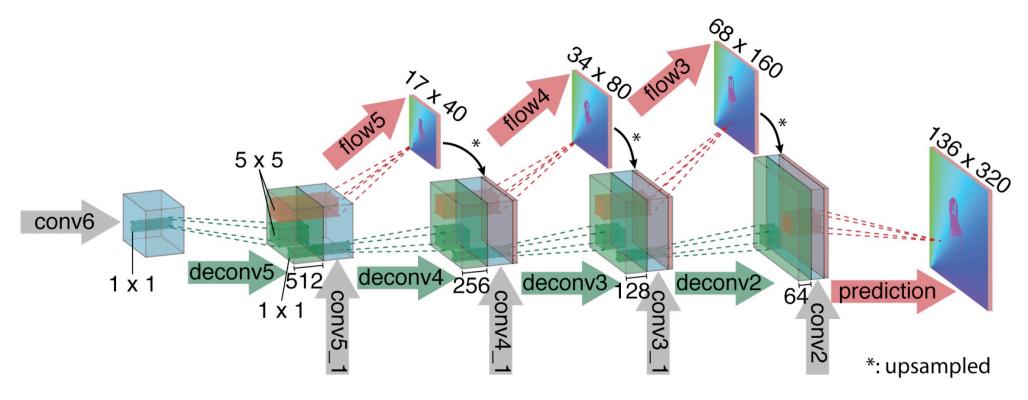
Correlation layer: multiplicative patch comparison between two feature maps

$$c(\mathbf{x}_1, \mathbf{x}_2) = \sum_{\mathbf{o} \in [-k, k] \times [-k, k]} \langle \mathbf{f}_1(\mathbf{x}_1 + \mathbf{o}), \mathbf{f}_2(\mathbf{x}_2 + \mathbf{o}) \rangle$$

- Two patches centered at x1 and x2, with size K = 2k + 1
- Convolve data with another data
- Limit the patches for comparison with maximum displacement d
- Only compare patches in a neighborhood with size D = 2d + 1
- Output size $(w \times h \times D^2)$

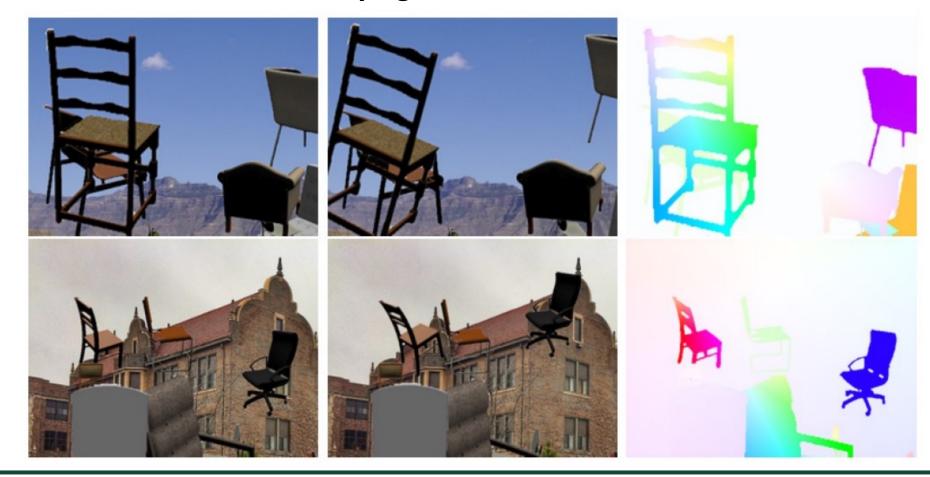


Refinement

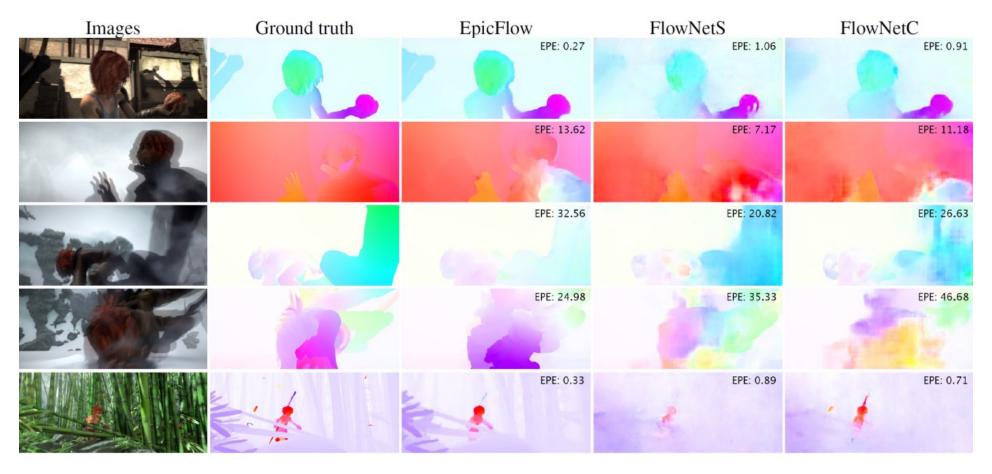


Training Data

Flying Chairs Dataset



Deep Optical Flow Results



Results on Sintel (standard benchmark)

Revisiting the Small Motion Assumption

- Is this motion small enough?
 - Probably not—it's much larger than one pixel (2nd order terms dominate)
 - How to solve this problem?

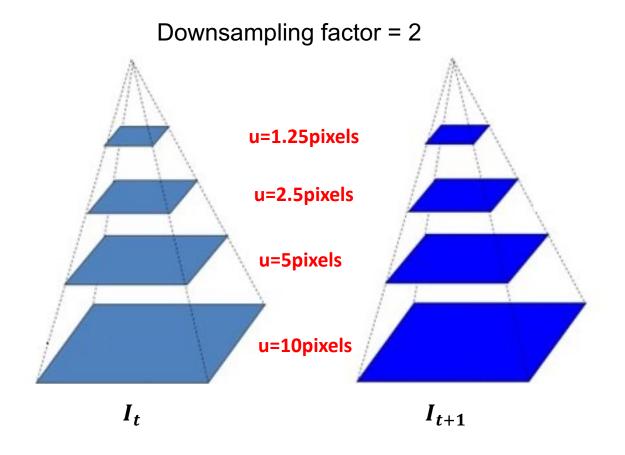


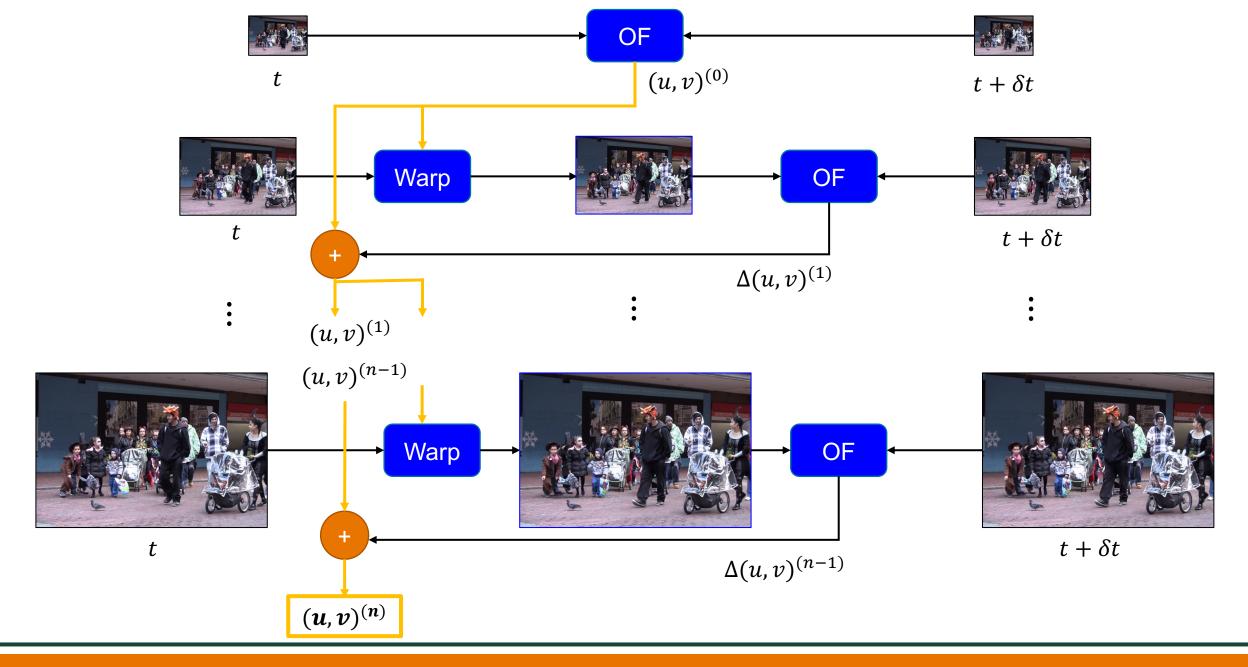


$$I(x+\Delta x,y+\Delta y,t+\Delta t)=I(x,y,t)+rac{\partial I}{\partial x}\Delta x+rac{\partial I}{\partial y}\Delta y+rac{\partial I}{\partial t}\Delta t+ ext{higher-order terms}$$

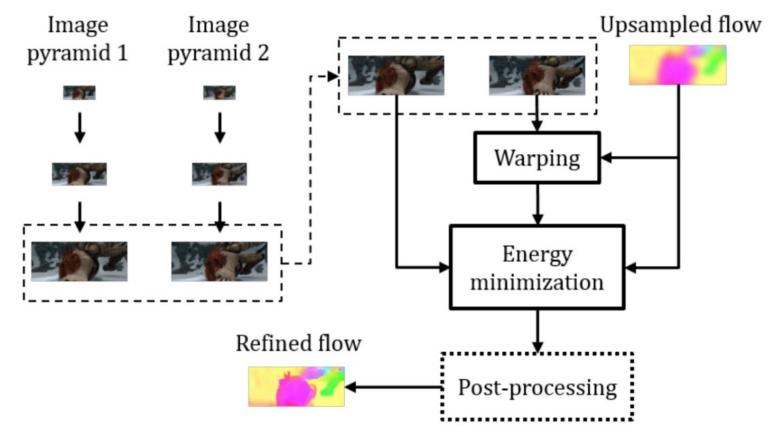
High-order terms will have large values for large motion

Coarse-to-fine Optical Flow Estimation



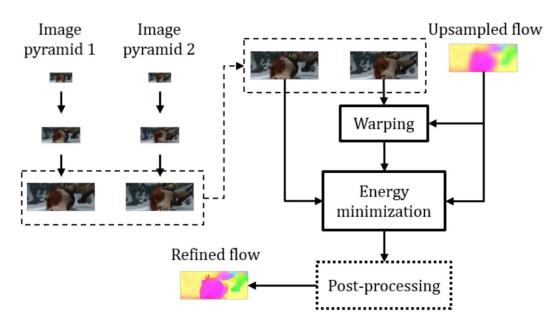


Coarse-to-fine Optical Flow Estimation

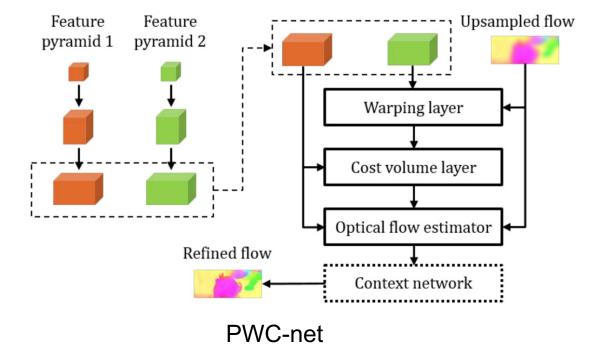


Simplified illustration

Coarse-to-fine Optical Flow Estimation

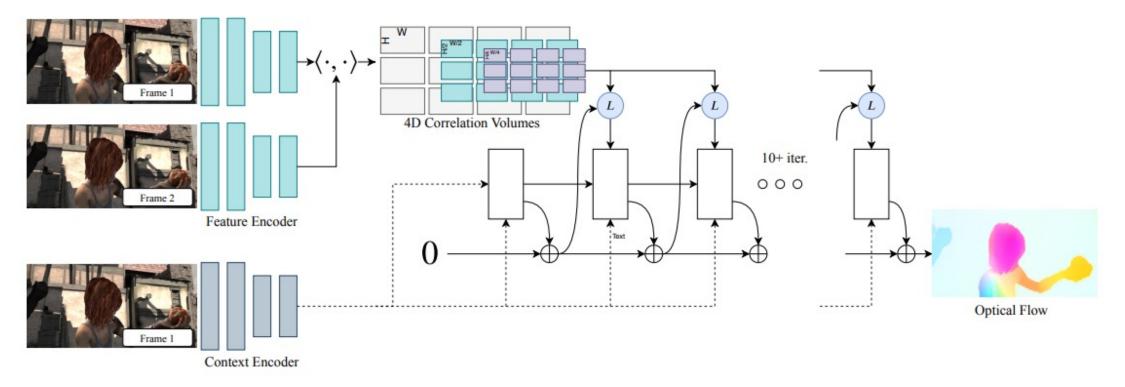


Traditional coarse-to-fine flow



[Sun et al., "PWC-Net", 2018]

RATF



Recurrent All-Pairs Field Transforms (RAFT), a new deep network architecture for optical flow

[Teed and Deng. "RATF", 2020]

Applications

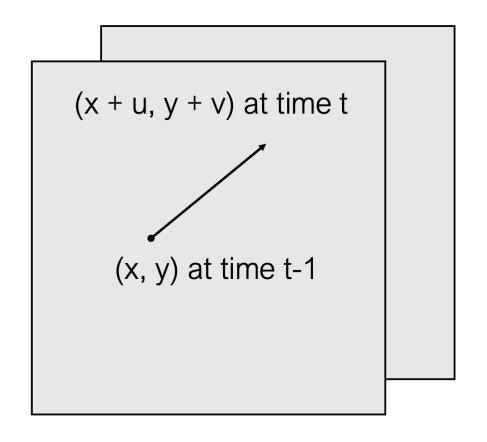
- Video Stabilization
- Video Frame Interpolation
- Action Recognition
- Video Restoration
- Visual Tracking
- •

Video Stabilization – Remove Camera Shake



https://cseweb.ucsd.edu/~ravir/jiyang_cvpr20.pdf [Yu and Ramamoorthi, 2020]

Video Frame Interpolation



 use flow to estimate where pixel will be between two frames

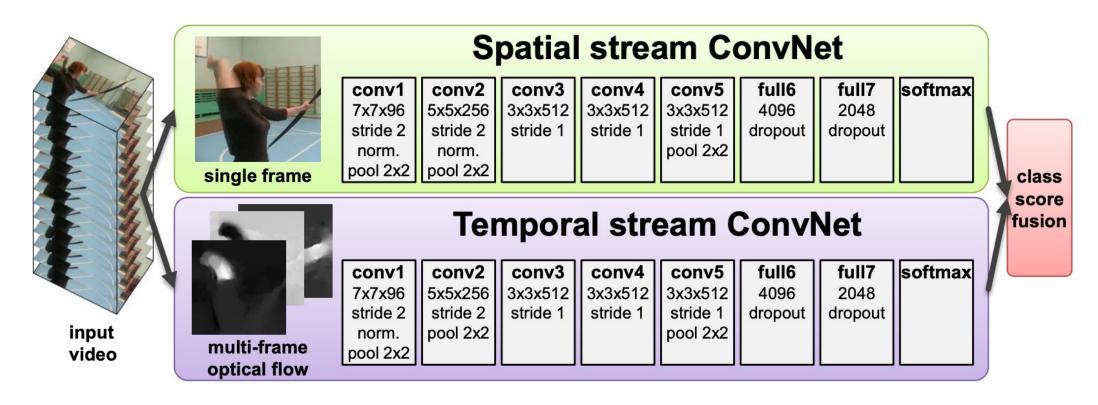
Synthesize intermediate frames to generate slow-motion videos

Credit: Shu Kong



https://www.youtube.com/watch?v=MjViy6kyiqs

Action Recognition

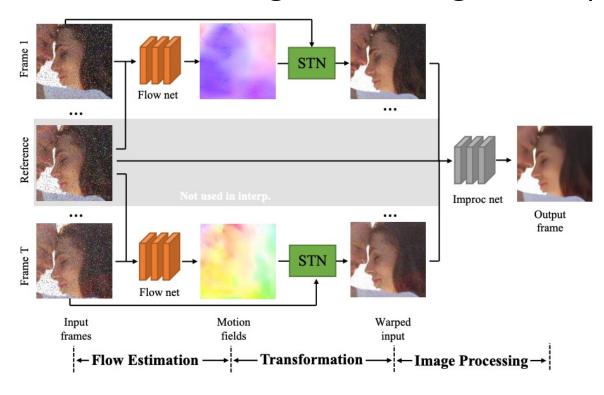


Two-stream architecture for video classification

[Simonyan and Zisserman, 2014]

Video Restoration

Optical flow can be used to address a series video restoration tasks, such as denoising, deblocking, and super-resolution



- Flow net to estimate motion field between neighboring frames
- Stack warped frames as input for the image processing network to predict the high-quality frame

Video Restoration



https://www.youtube.com/watch?v=msC5GK9aV9Q

Visual Tracking



https://nanonets.com/blog/optical-flow/

Further Reading

FlowNet: Learning Optical Flow with Convolutional Networks, 2015 https://arxiv.org/abs/1504.06852

PWC-Net: CNNs for Optical Flow Using Pyramid, Warping, and Cost Volume, 2018 https://arxiv.org/abs/1709.02371

RAFT: Recurrent All-Pairs Field Transforms for Optical Flow, 2020

https://arxiv.org/abs/2003.12039