



Massachusetts
Institute of
Technology



Thesis Defense:

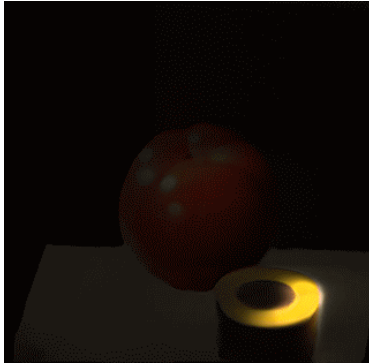
**Analysis and Visualization
of Temporal Variations in Video**

Nov 25 2013

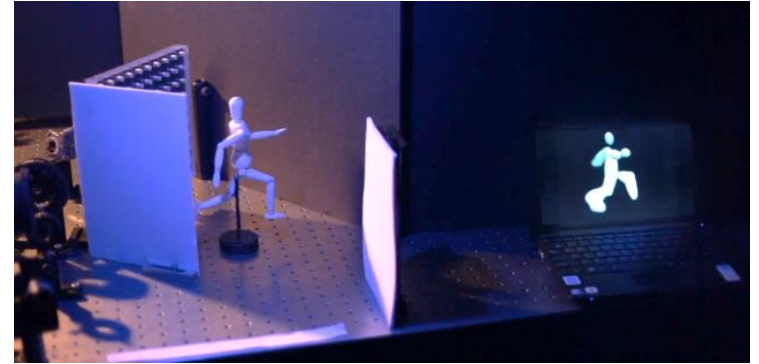
Michael Rubinstein

MIT CSAIL

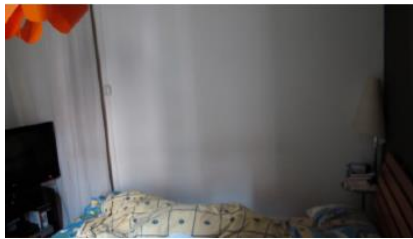
Seeing the Unseen in Images/Videos



[Velten et al., *Femto-Photography*, 2011]



[Velten et al., *CORNAR*, 2012]



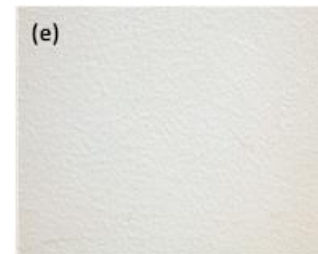
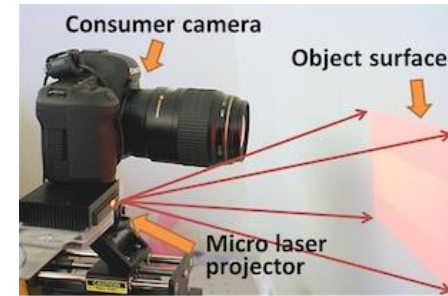
Input



Result

True image

[Torralba and Freeman, *Accidental Images*, 2012]



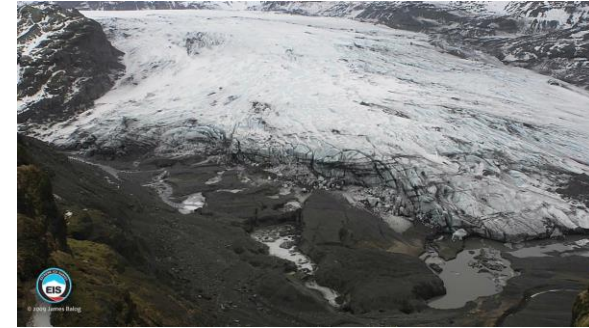
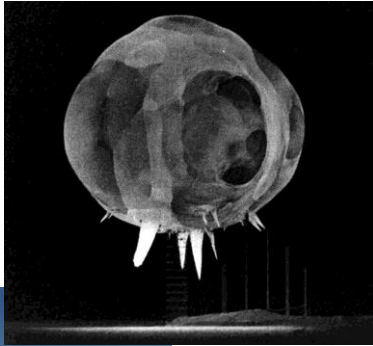
Input



Result

[Shih et al., *Laser Speckle Photography*, 2012]

Timescales in Imagery



Milliseconds
 10^4 fps (high-speed)

Seconds, Minutes
 10^1 fps (standard videos)

Months, years
 10^{-4} fps (time-lapse)

Timescale

Distracting Temporal Variation

- Too many confounding changes
 - Mixed changes at different timescales
 - Lighting changes, objects appearing/disappearing, ...



Source

Short-term changes attenuated

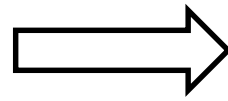
© Extreme Ice Survey

Remove changes to make long-term variation more visible!

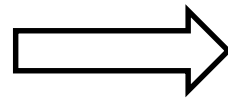
Imperceptible Temporal Variation

Magnify the variation to make it visible!

- Changes are too small



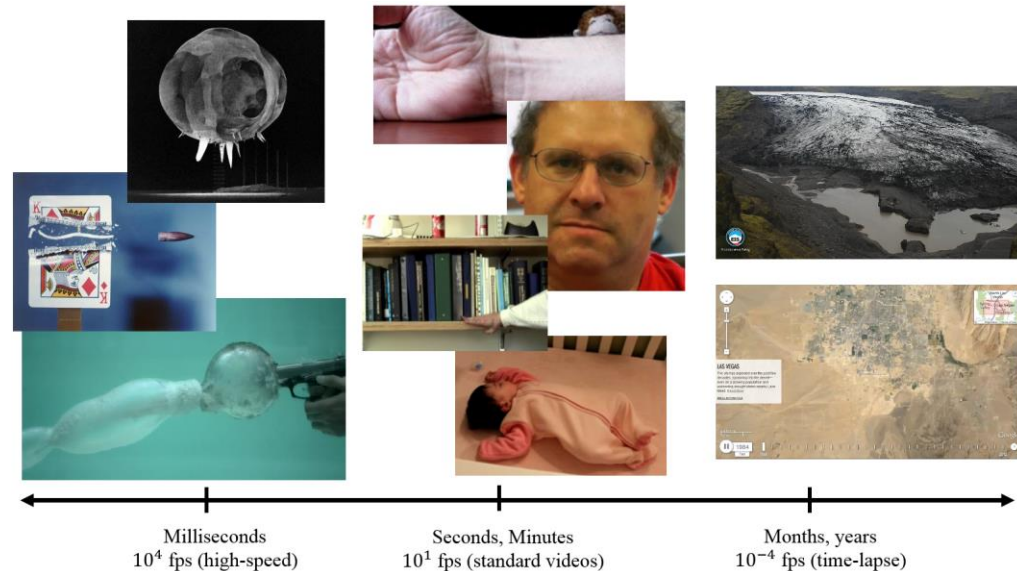
[Liu et al. 2005]



This Thesis

- Assist the analysis of temporal phenomena captured by imagery
- Reveal interesting temporal signals that may not be easily visible in the original data

- Leverage available imagery
 - Regular video, natural setting



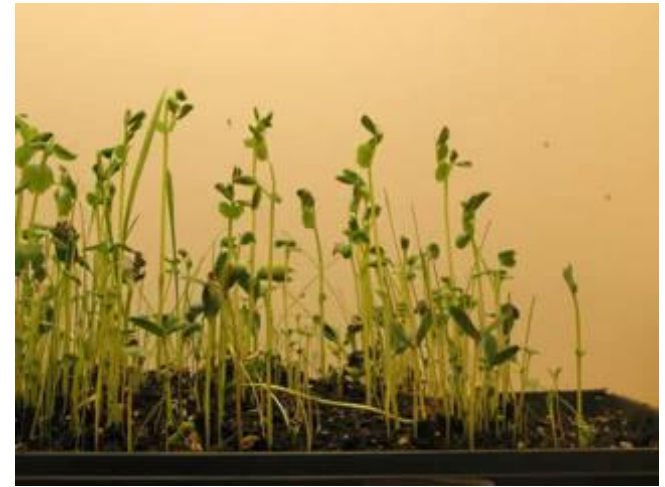
Our approach: analyzing images/videos and **re-rendering** changes in them such that the interesting temporal signals are more apparent

Talk Outline

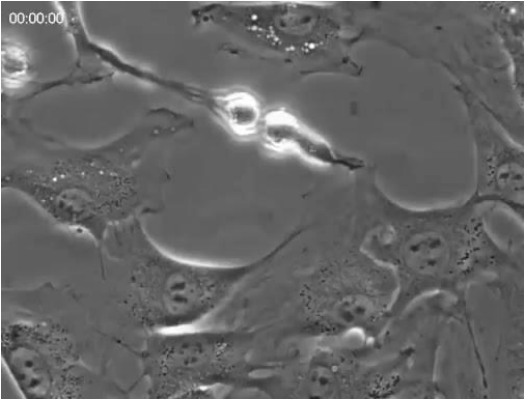
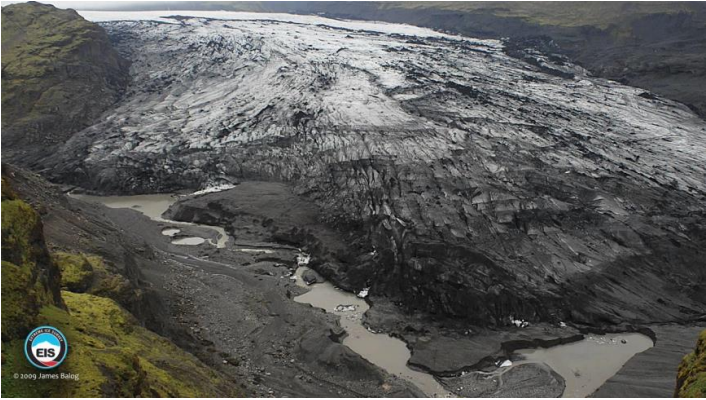
- Removing distracting variations
 - Motion Denoising

- Magnifying imperceptible variations
 - Eulerian Video Magnification
 - Phase-based Video Motion Processing

- Ongoing research and future work



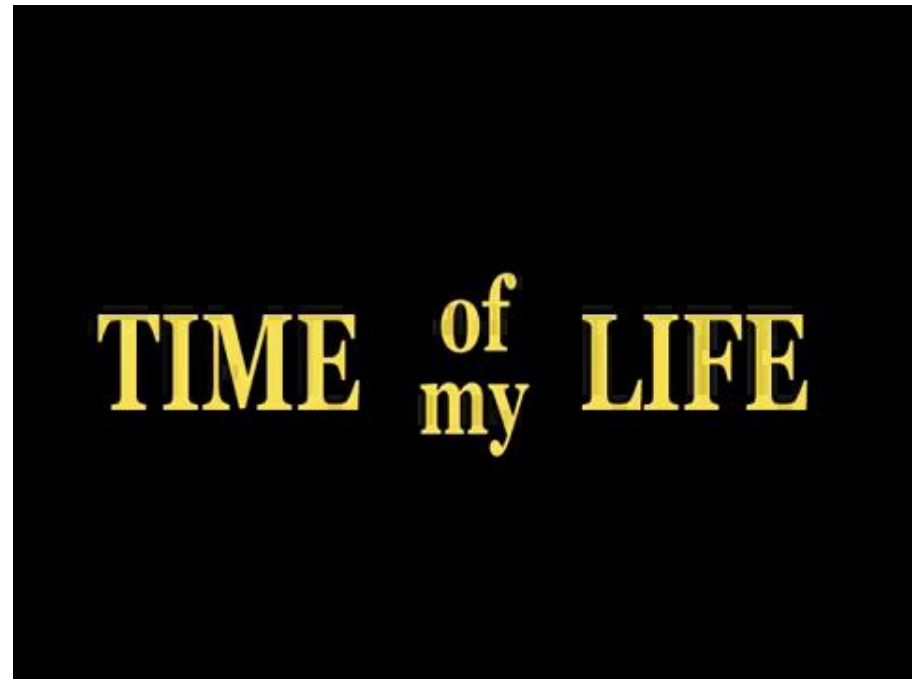
Time-lapse Videos



For Personal Use Too



9 months



16 years



10 years



1 year

something's not right...

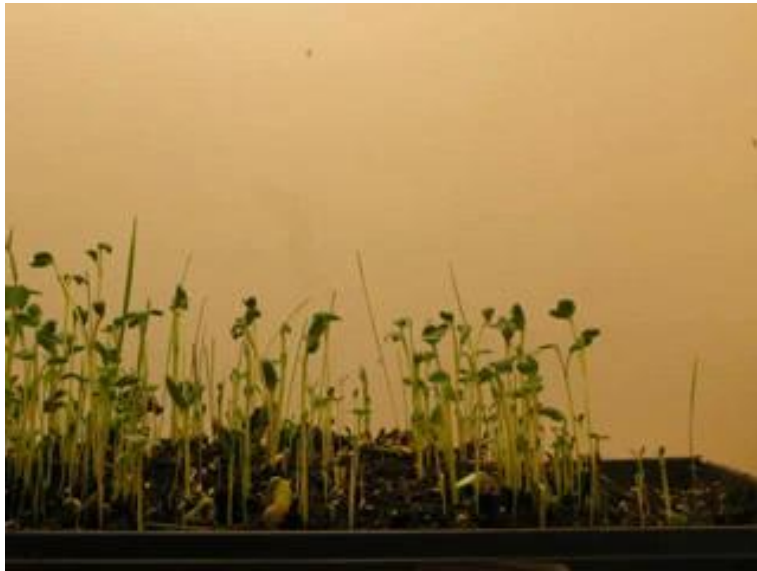
1566 km

Stylized Jerkiness

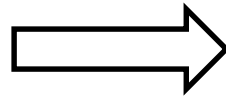


Timescale Separation

- Decompose the video into long-term and short-term changes



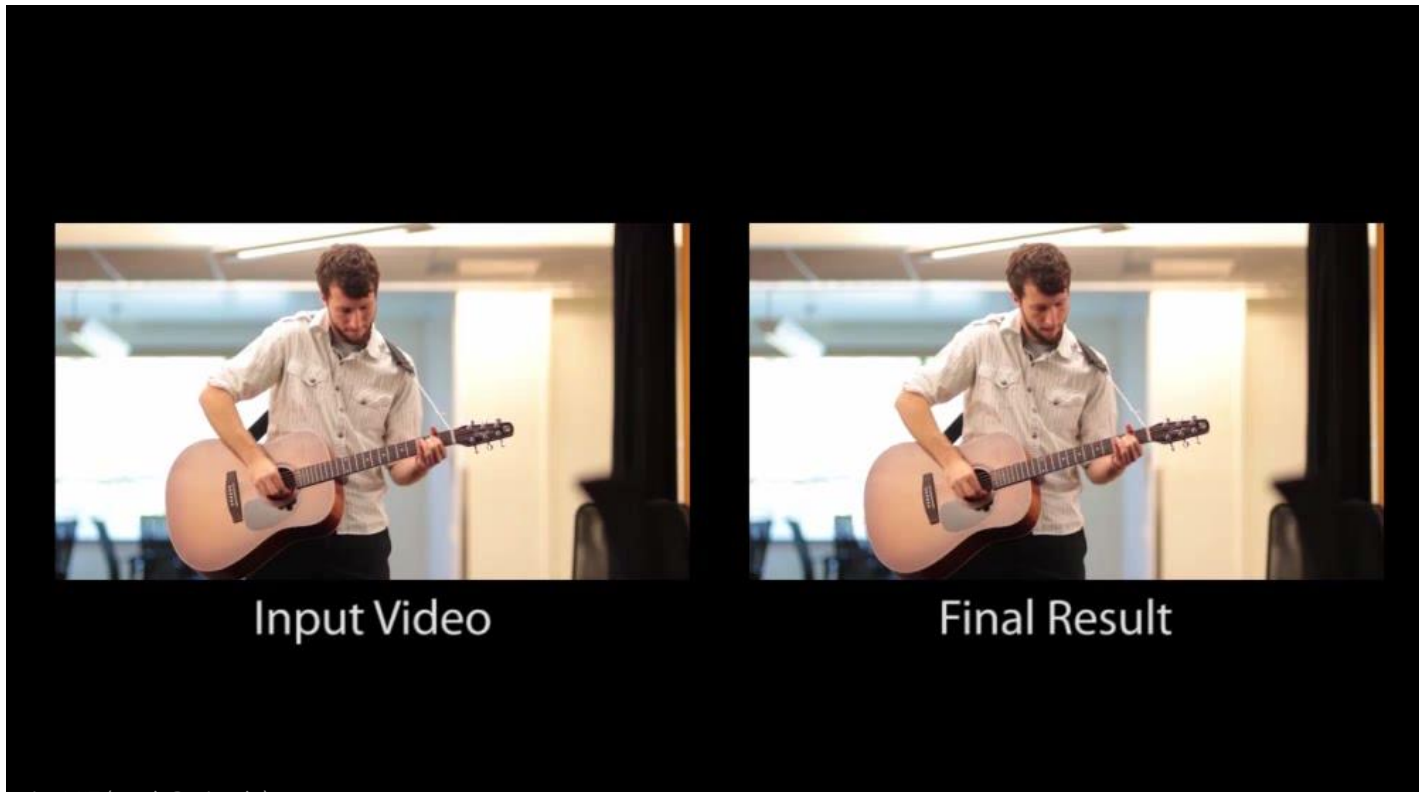
***Motion
Denoising***



Motion Denoising with Application to Time-lapse Photography, CVPR 2011
With Ce Liu, Peter Sand, Fredo Durand, William T. Freeman

Related Work

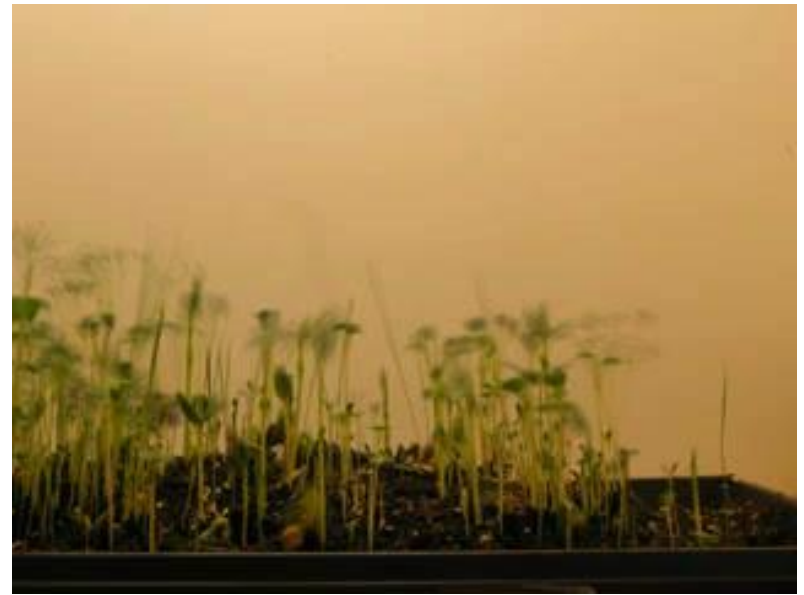
- **Video stabilization** [Matsushita et al. 2006], [Liu et al. 2011], [Grundmann et al. 2011]
 - Can denoise camera motions, but we need pixel-wise stabilization
- **Selectively De-Animating Video** [Bai et al. 2012]



How to Denoise Motion?

- Pixel-wise temporal low-pass filtering
 - Pixels of different objects are averaged

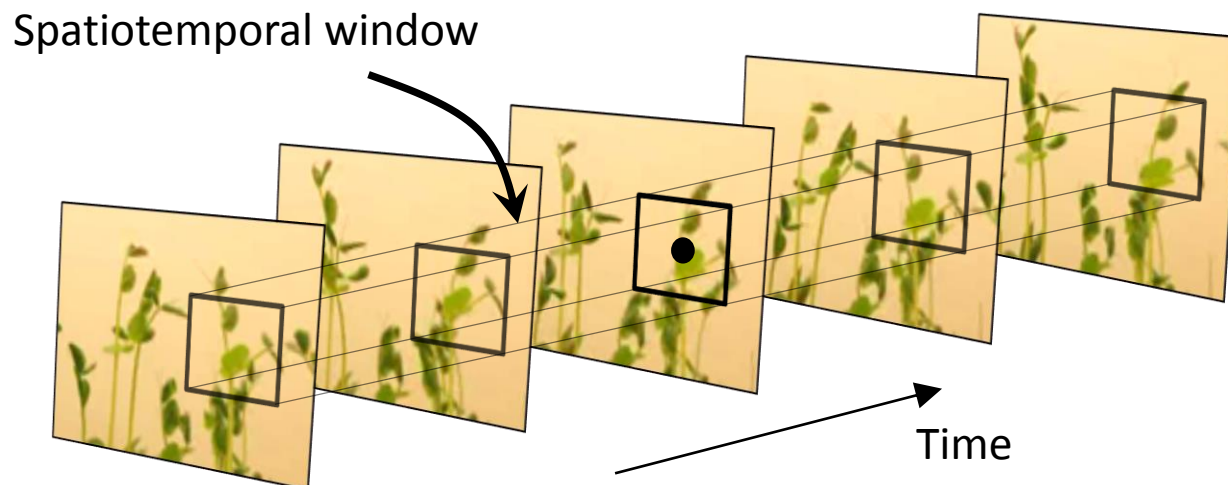
- Smoothing motion trajectories
 - Motion estimation in time-lapse videos is challenging:
 - Brightness *in*consistencies
 - Motion *dis*continuities



KLT tracks

Basic Idea

- **Assumption:** scene is changing slowly and perturbed by random motions (and color changes)
- **Approach:** reshuffle the pixels in space and time to reconstruct the smooth scene
 - Allow the filter to “look around” within local spatiotemporal windows



Formulation

- Solve for a spatiotemporal displacement (offset) field, \mathbf{w} :

$$E(\mathbf{w}) = \sum_p |I(p + \mathbf{w}(p)) - I(p)| +$$

Fidelity (to input video)

$$\alpha \sum_{p,r \in N_t(p)} \|I(p + \mathbf{w}(p)) - I(r + \mathbf{w}(r))\|^2 +$$

Temporal coherence
(of the output video)

$$\gamma \sum_{p,q \in N(p)} \lambda_{pq} |\mathbf{w}(p) - \mathbf{w}(q)|$$

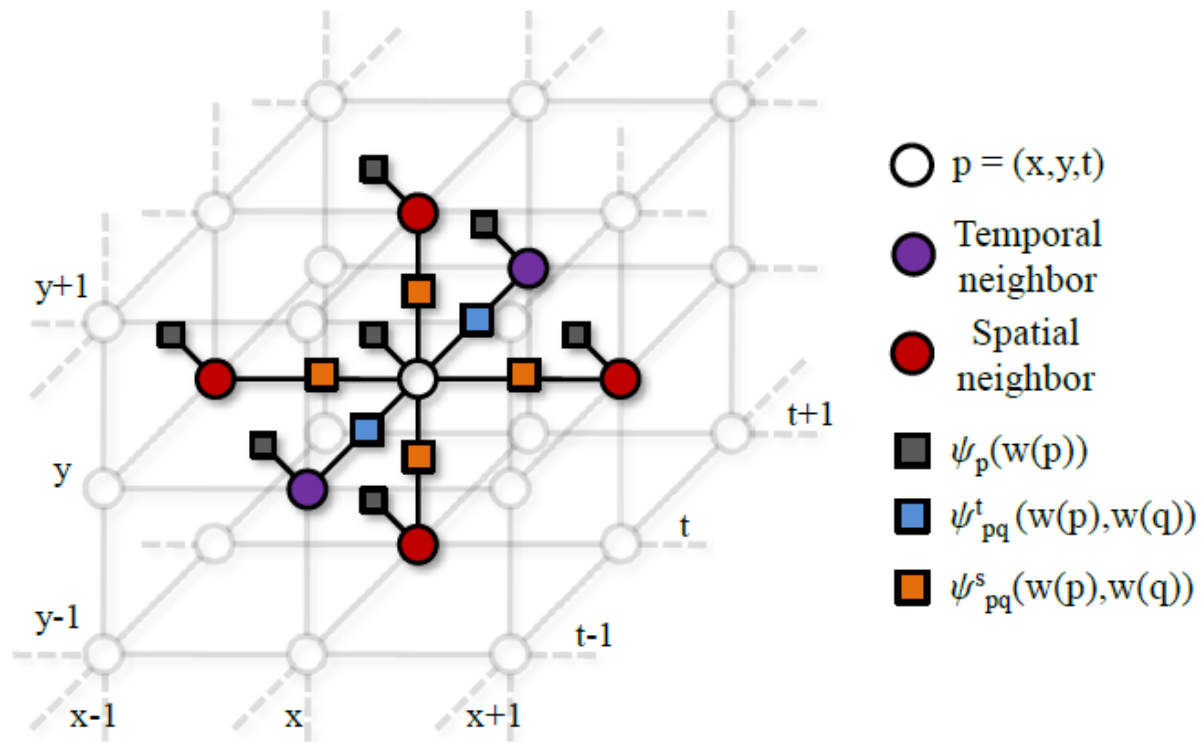
Regularization
(of the warp)

$I(p)$ - the input video

$I(p + \mathbf{w}(p))$ - the output video

Optimization

- Optimized discretely on a 3D MRF
 - Nodes represent pixels
 - state space of each pixel = volume of possible spatiotemporal offsets



Results



Comparison with Other Optimization Techniques



ICM



GCUT



LBP



Iterated conditional modes

© Michael Rubinstein, MIT (mrub@mit.edu)



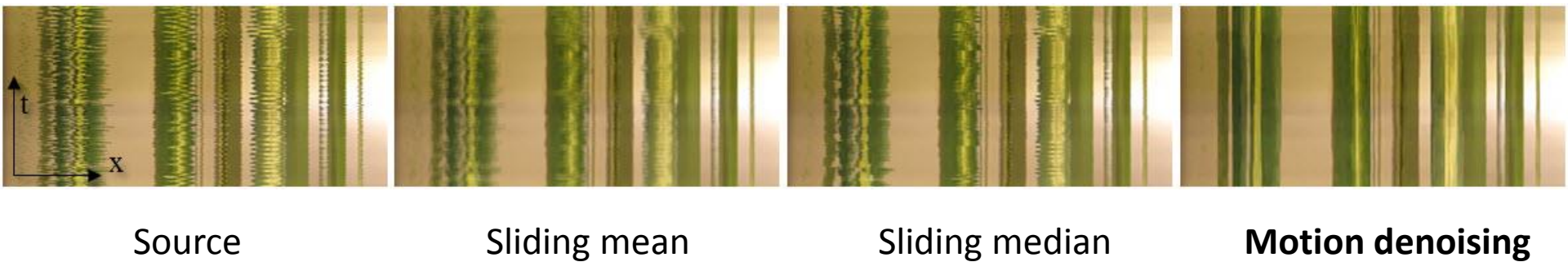
Graph Cut (α -expansion)

[Boykov et al. 2002]

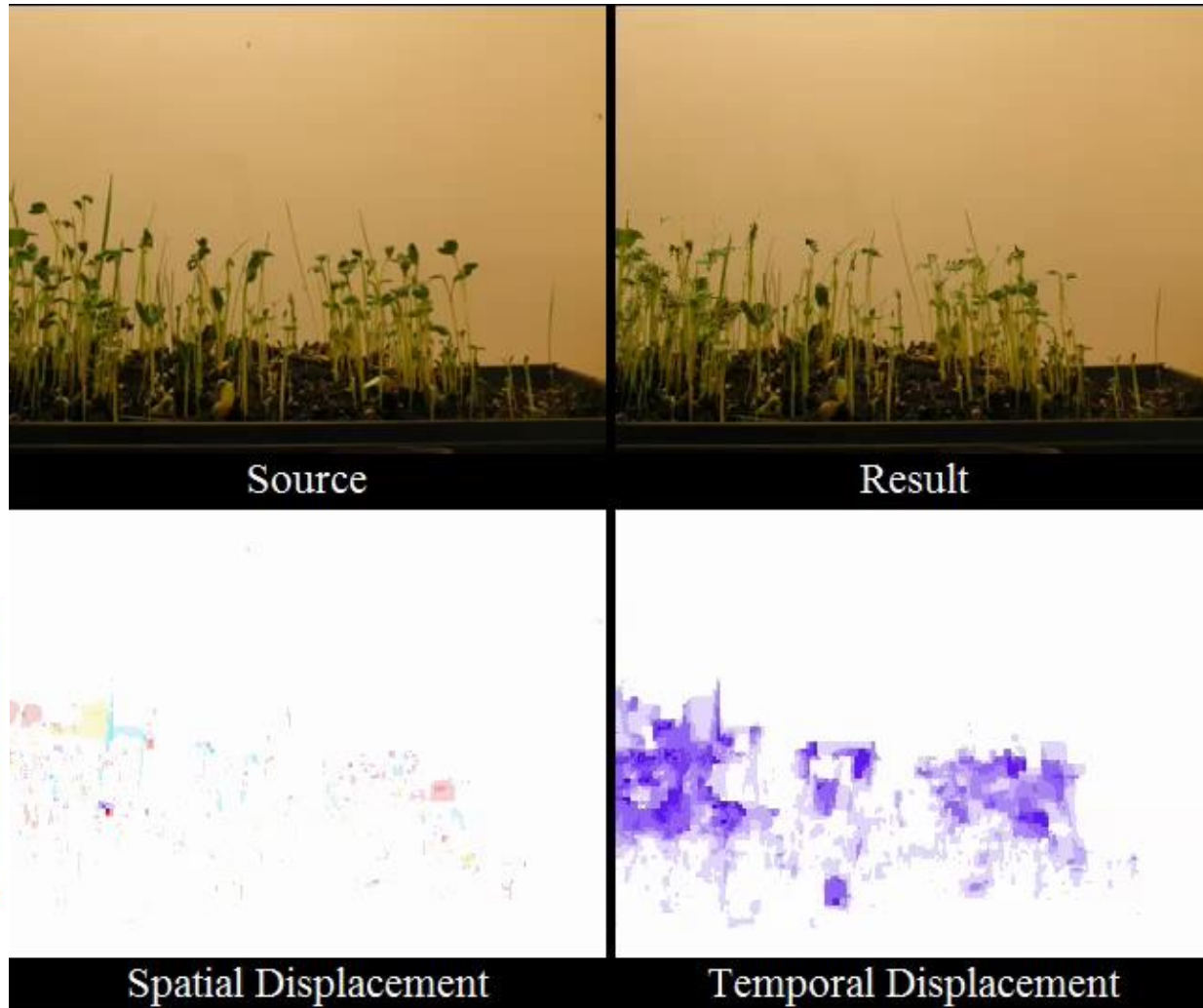


Belief Propagation

Comparison with Pixel-wise Temporal Filtering



Results



Support Size

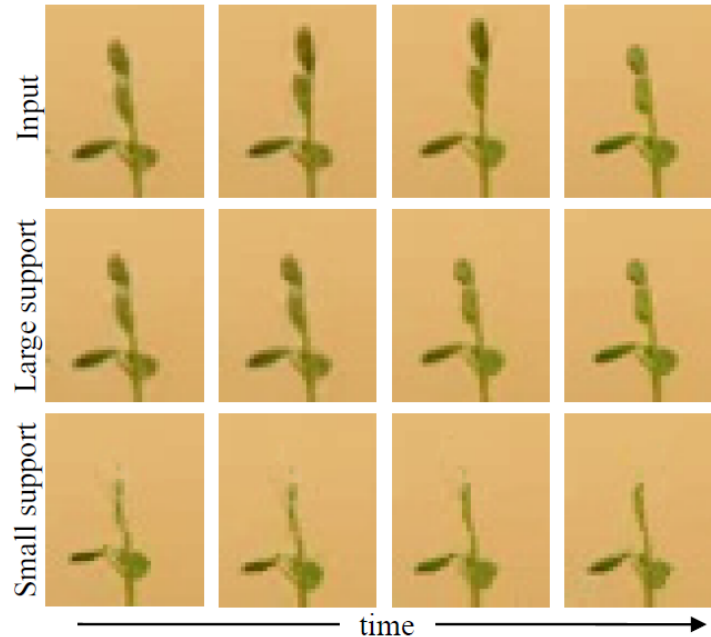
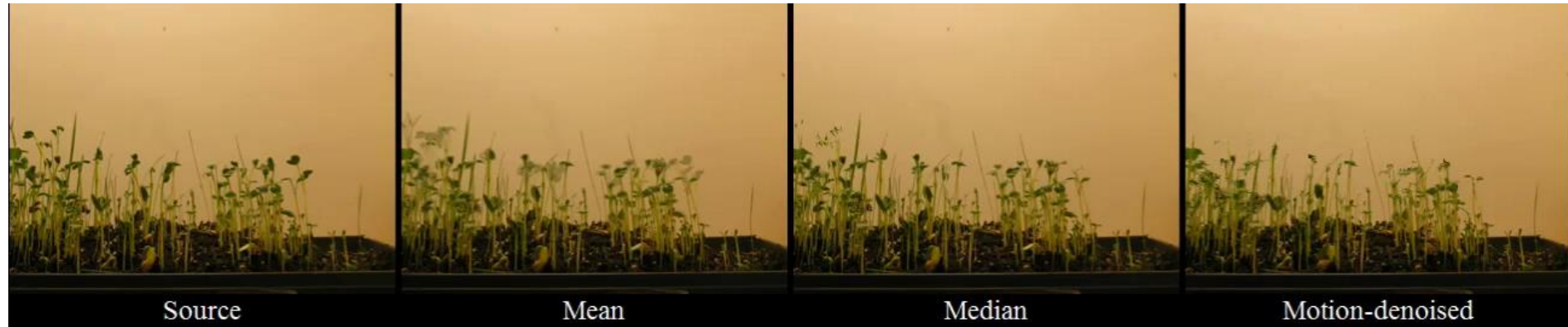


Figure 7. Zoom-in on the rightmost plant in the sprouts sequence in four consecutive frames shows that enlarging the search volume used by the algorithm can greatly improve the results. “Large support” corresponds to a $31 \times 31 \times 5$ search volume, while “small support” is the $7 \times 7 \times 5$ volume we used in our experiments.

Comparison with Pixel-wise Temporal Filtering



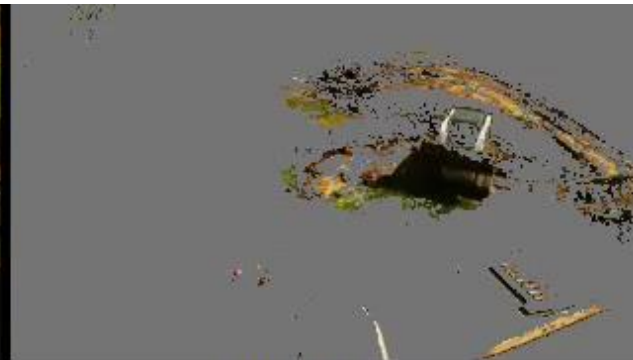
Timescale Decomposition



Source



Result (long-term)



Short-term



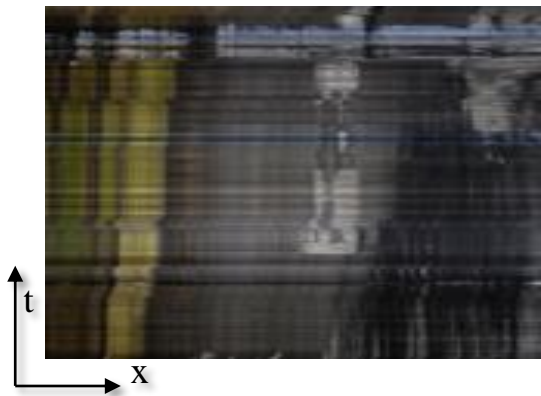
Source



Result (long-term)



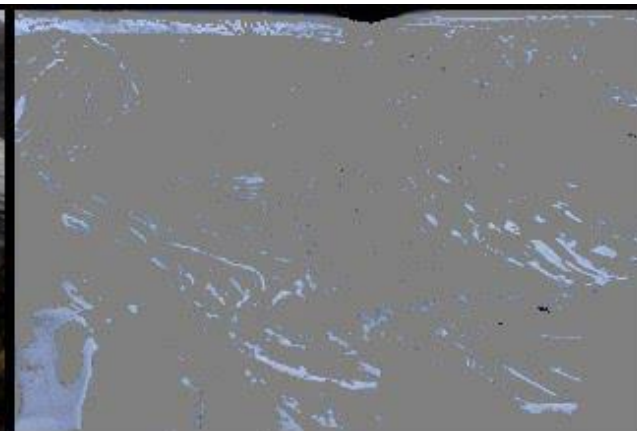
Short-term



Source



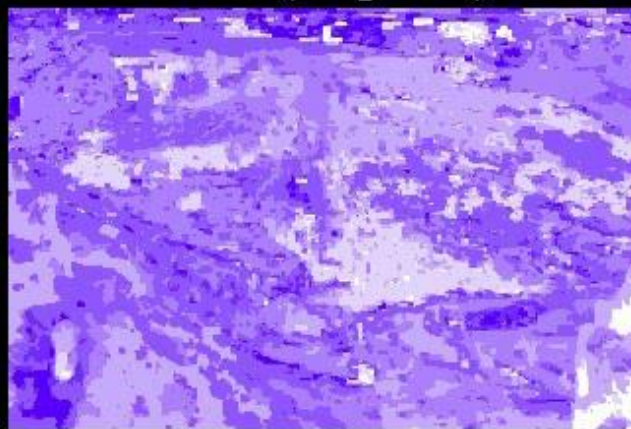
Result (long-term)



Short-term



Spatial Displacement



Temporal Displacement

Talk Outline

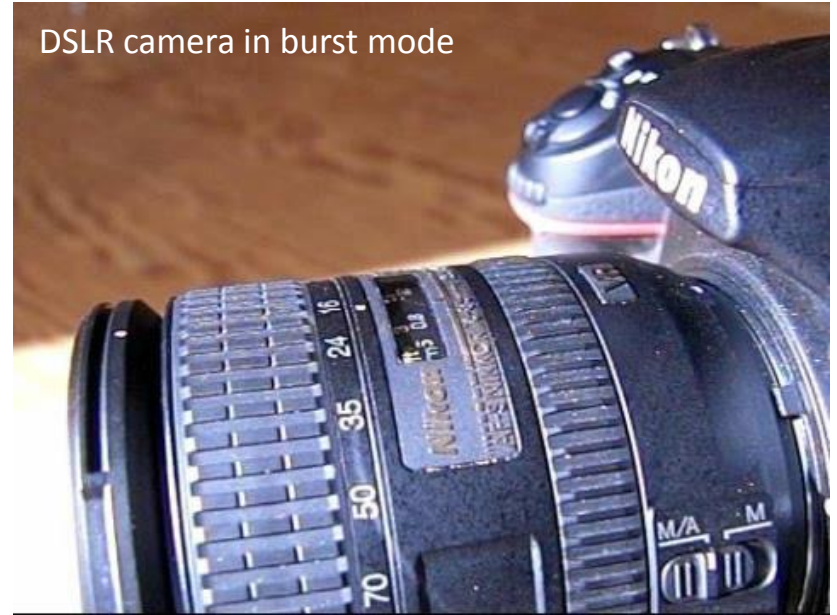
- Removing distracting variations
 - Motion Denoising

- Magnifying imperceptible variations
 - Eulerian Video Magnification
 - Phase-based Video Motion Processing

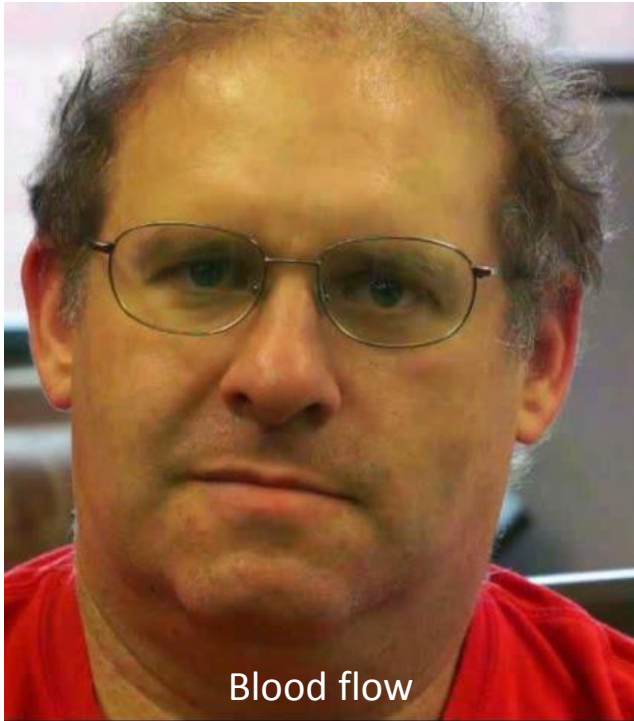
- Ongoing research and future work



Imperceptible Changes in the World



Imperceptible Changes in the World



Blood flow



Camera motion due to shutter and mirror



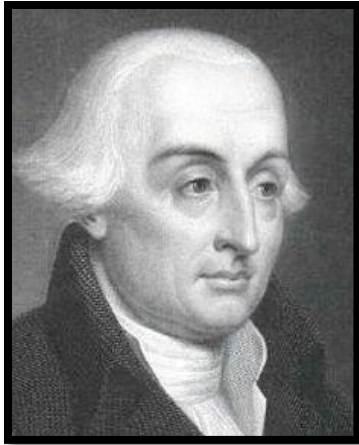
Breathing



Micro-expressions

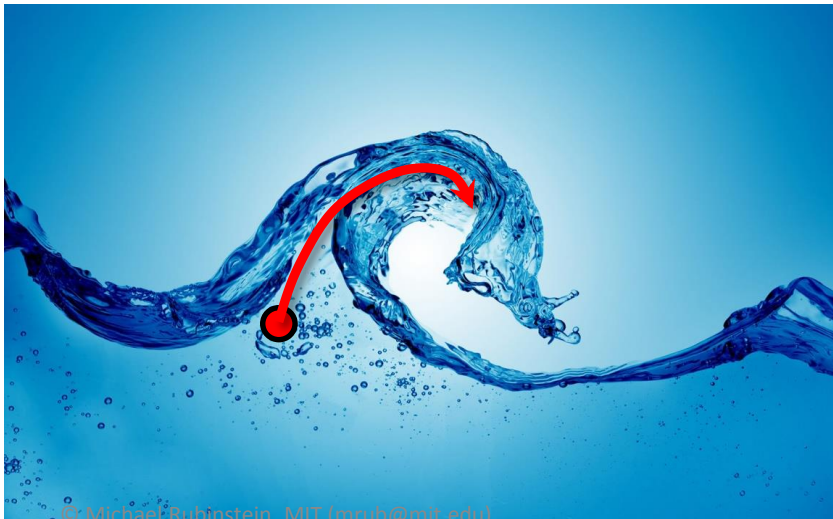
Lagrangian and Eulerian Perspectives (Fluid Dynamics)

- Specifications of physical measurements through space and time:



Lagrangian

Track particles

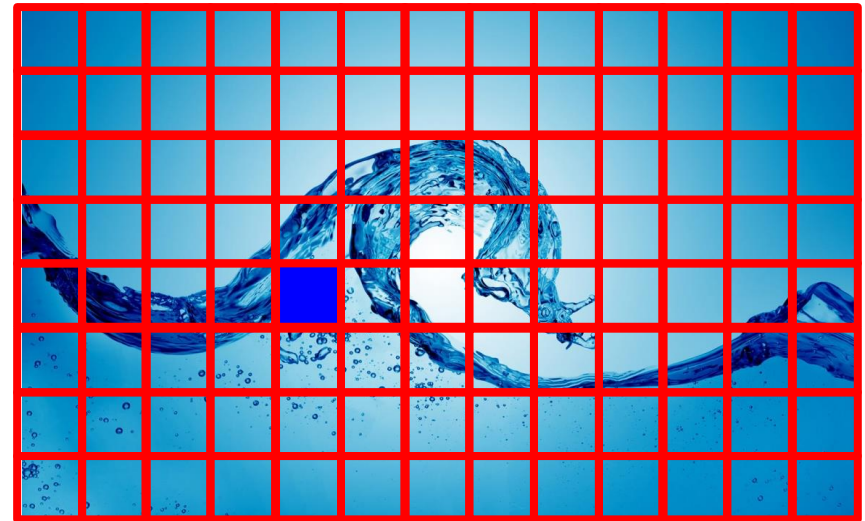


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Eulerian

Measure changes within fixed voxels in space



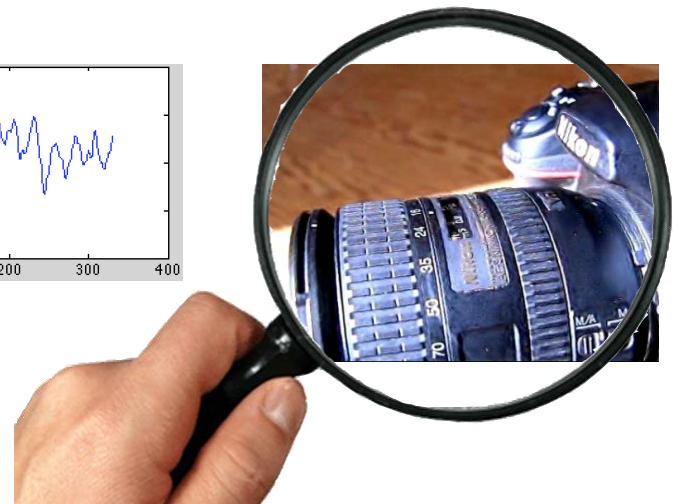
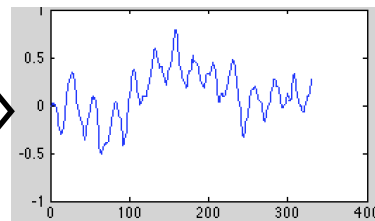
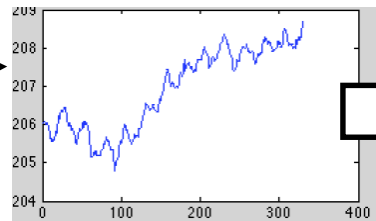
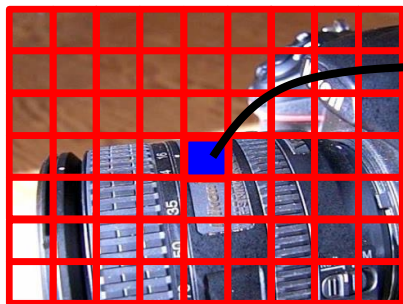
Basic Idea

- **Amplify temporal pixel color variations**
 - Each pixel processed independently
 - Treat each pixel as a time series
 - Apply standard 1D signal processing to it
 - Amplify particular temporal frequencies



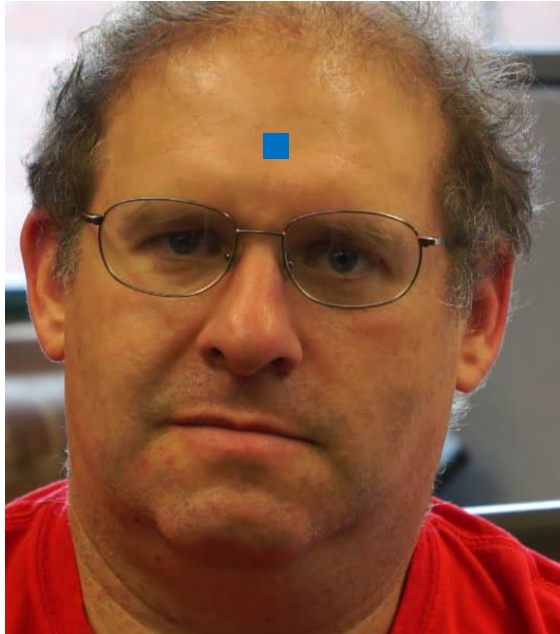
Eulerian Video Magnification (SIGGRAPH 2012)

With Hao-Yu Wu, Eugene Shih, John Guttag, Fredo Durand, Bill Freeman

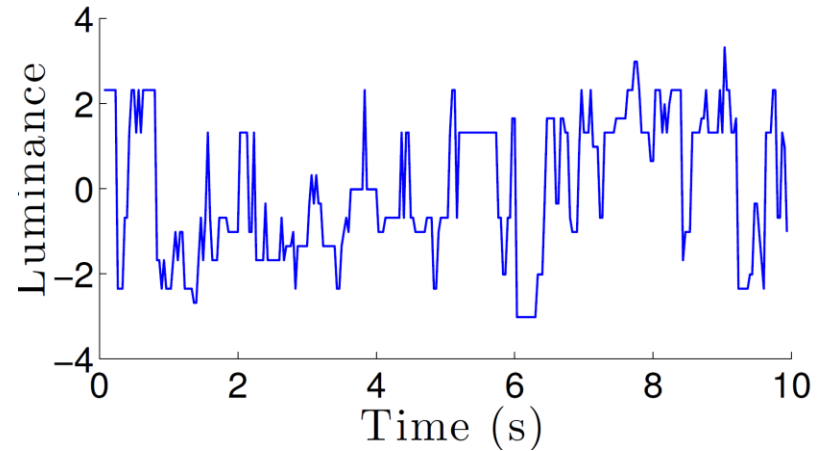


Subtle Color Variations

- The face gets slightly redder when blood flows
 - Very low amplitude: 0.5 intensity level in an 8-bit scale (0-255)



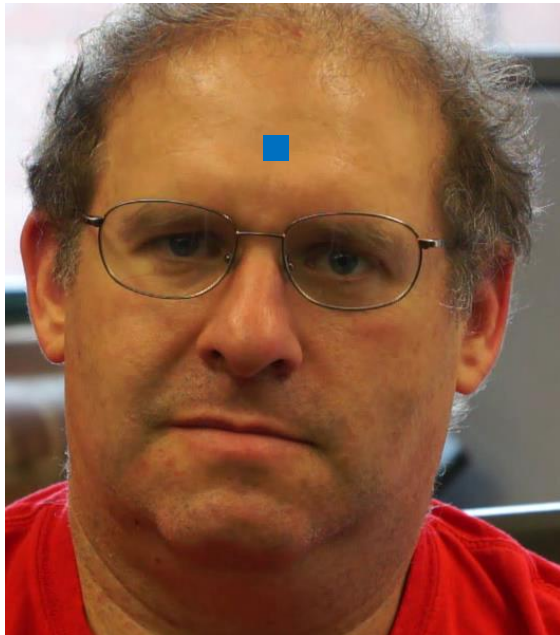
Input frame



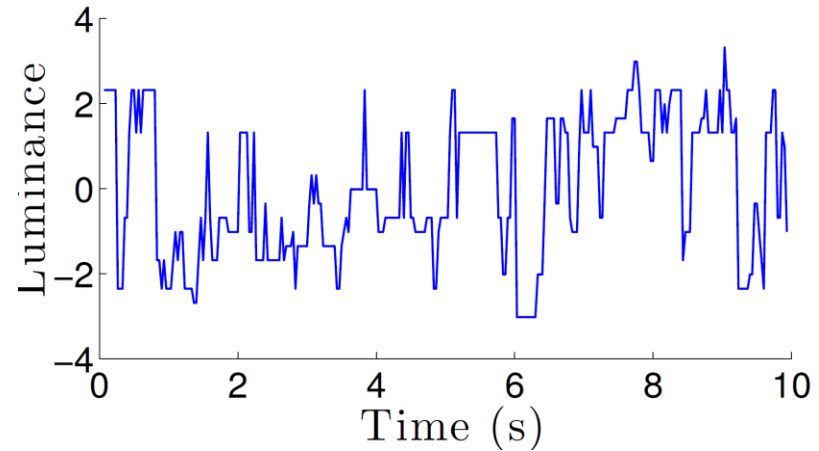
Luminance trace (zero mean)

Subtle Color Variations

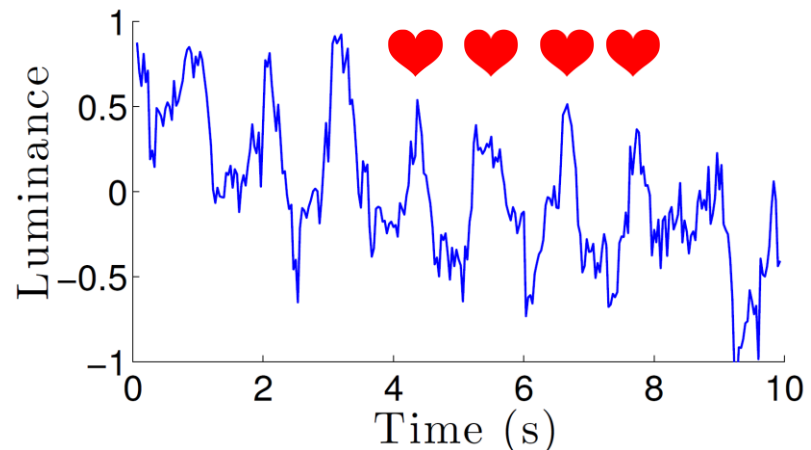
1. Average spatially to overcome sensor and quantization noise



Input frame



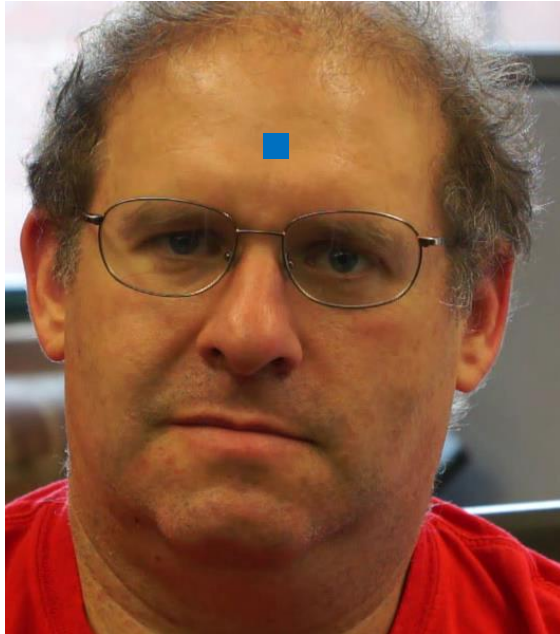
Luminance trace (zero mean)



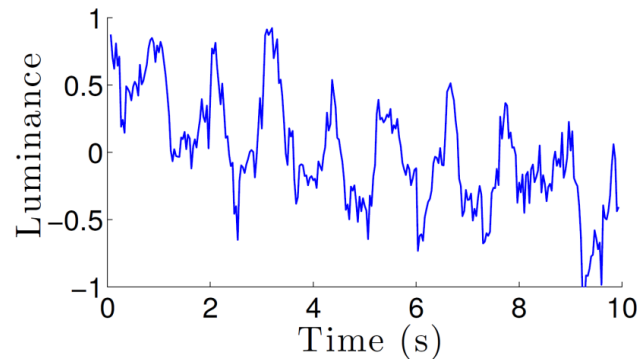
Spatially averaged luminance trace

Amplifying Subtle Color Variations

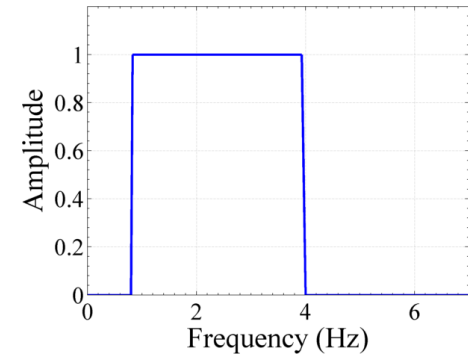
2. Filter temporally to extract the signal of interest



Input frame

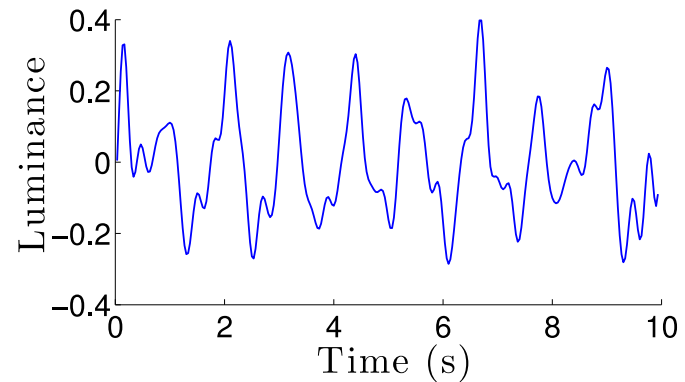


Spatially averaged luminance trace



Temporal filter

=



Temporally bandpassed trace

Color Amplification Results



Source



Color-amplified (x100)
0.83-1 Hz (50-60 bpm)

Bruce Wayne's Pulse

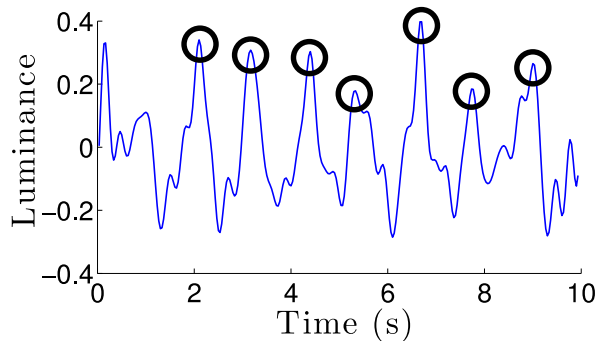


Christian Bale, *Batman Begins* (2005)

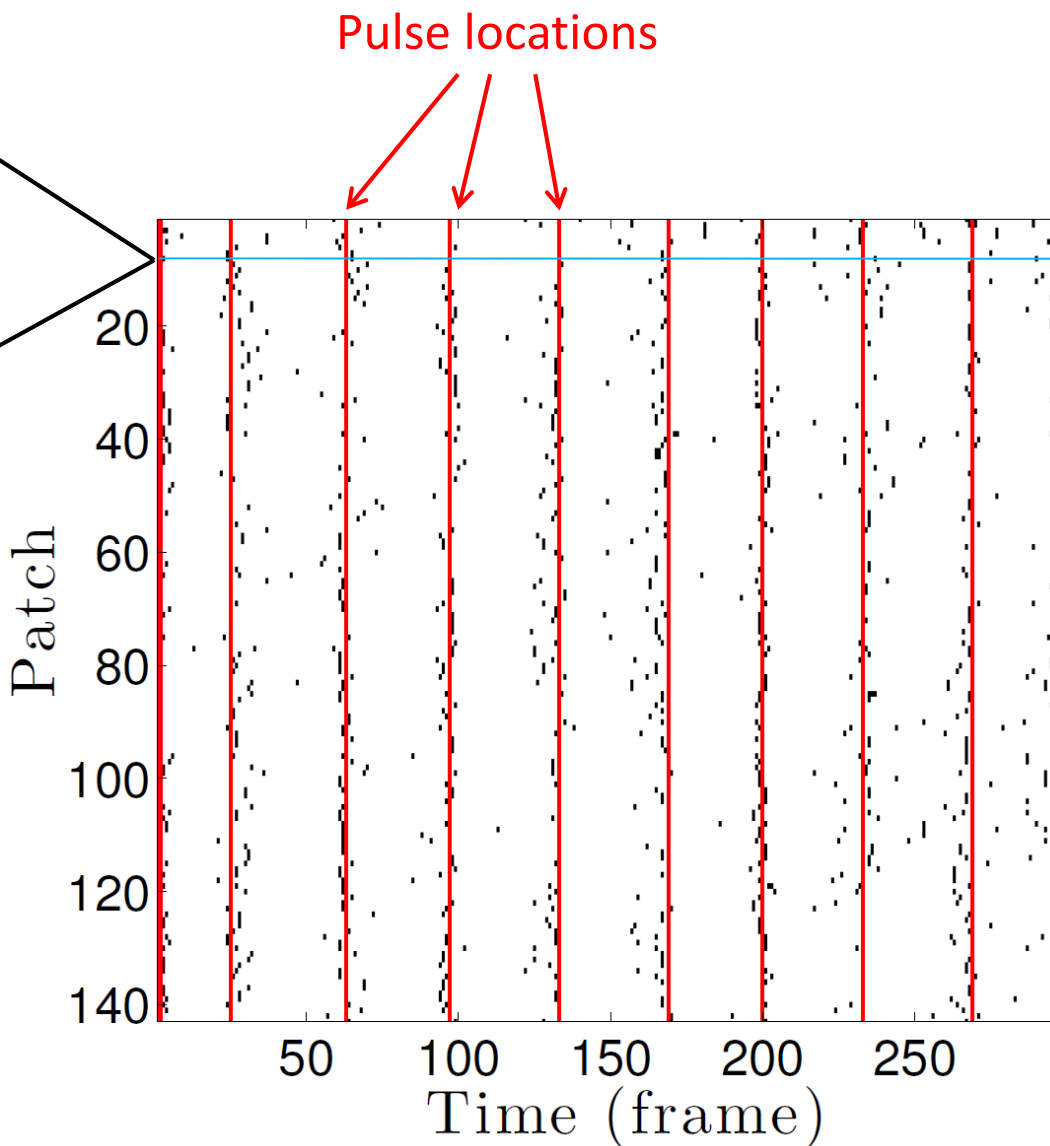
Courtesy of Warner Bros. Pictures

Heart Rate Extraction

Peak detection



Temporally bandpassed trace
(one pixel)



Extracting Heart Rate



Source

(Courtesy of Winchester Hospital. Do not copy)



← EKG

Hospital monitor

Bandpass signal +
peaks (pulse)

Estimated heart rate



● 146 bpm

35.2 w
Age: 19 d

With Dr. Donna Brezinski
and the Winchester
Hospital staff

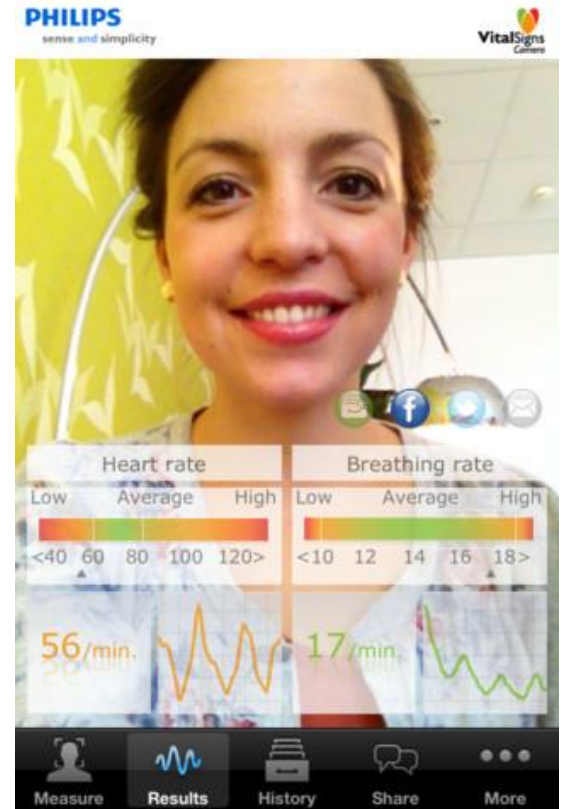
© Michael Rubinstein, MIT (mrub@mit.edu)

Color-amplified (x150) 2.33-2.67 Hz (140-160 bpm)

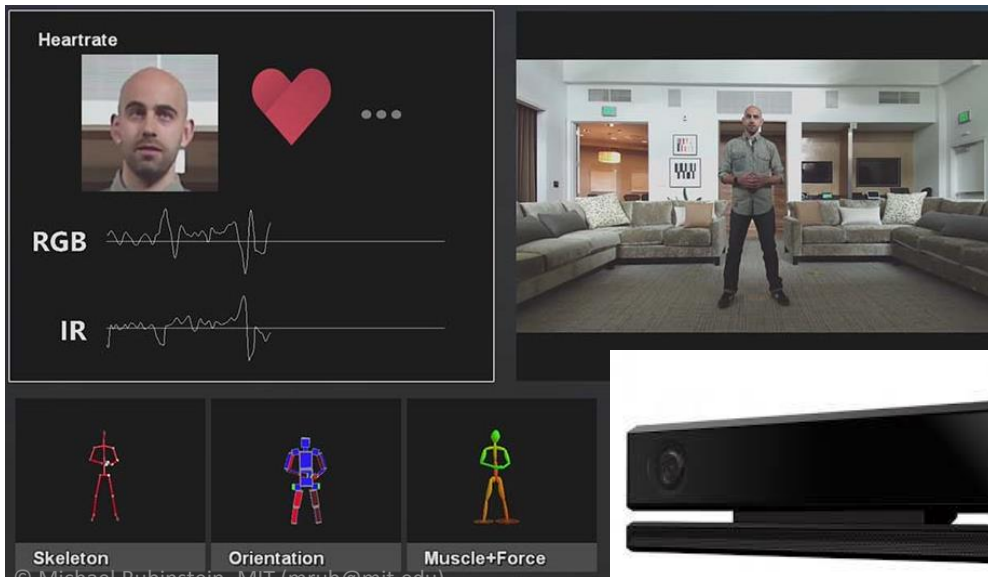
Related Work: Pulse Detection from Video



“Cardiocam” [Pho, Picard, McDuff 2010]



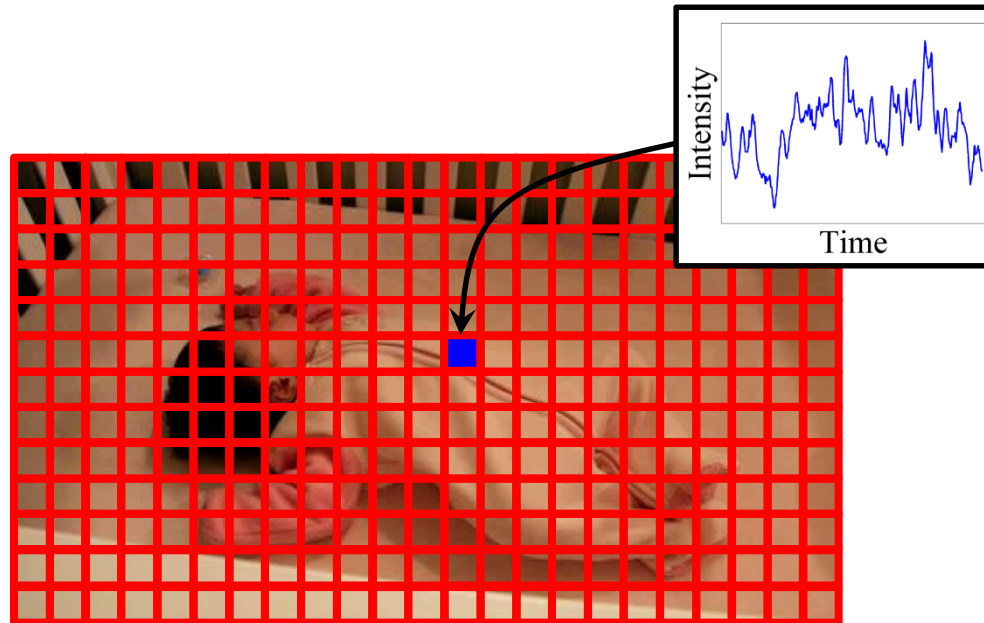
“Vital Signs Camera” – Philips
proprietary



Kinect (Xbox One)
proprietary

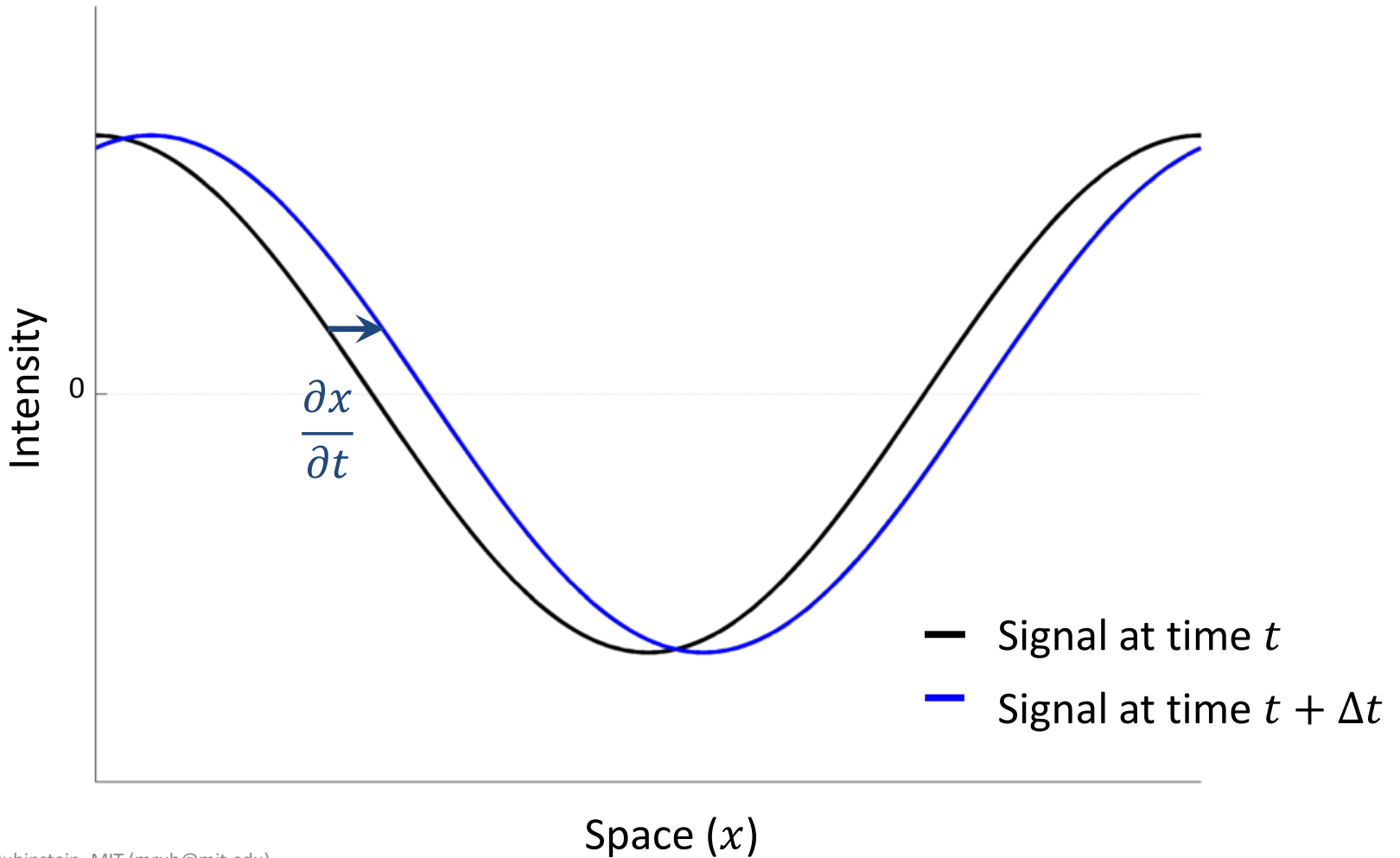
Why Does it Amplify Motion?

- By increasing temporal variation – we can increase spatial motion!



Differential Brightness Constancy

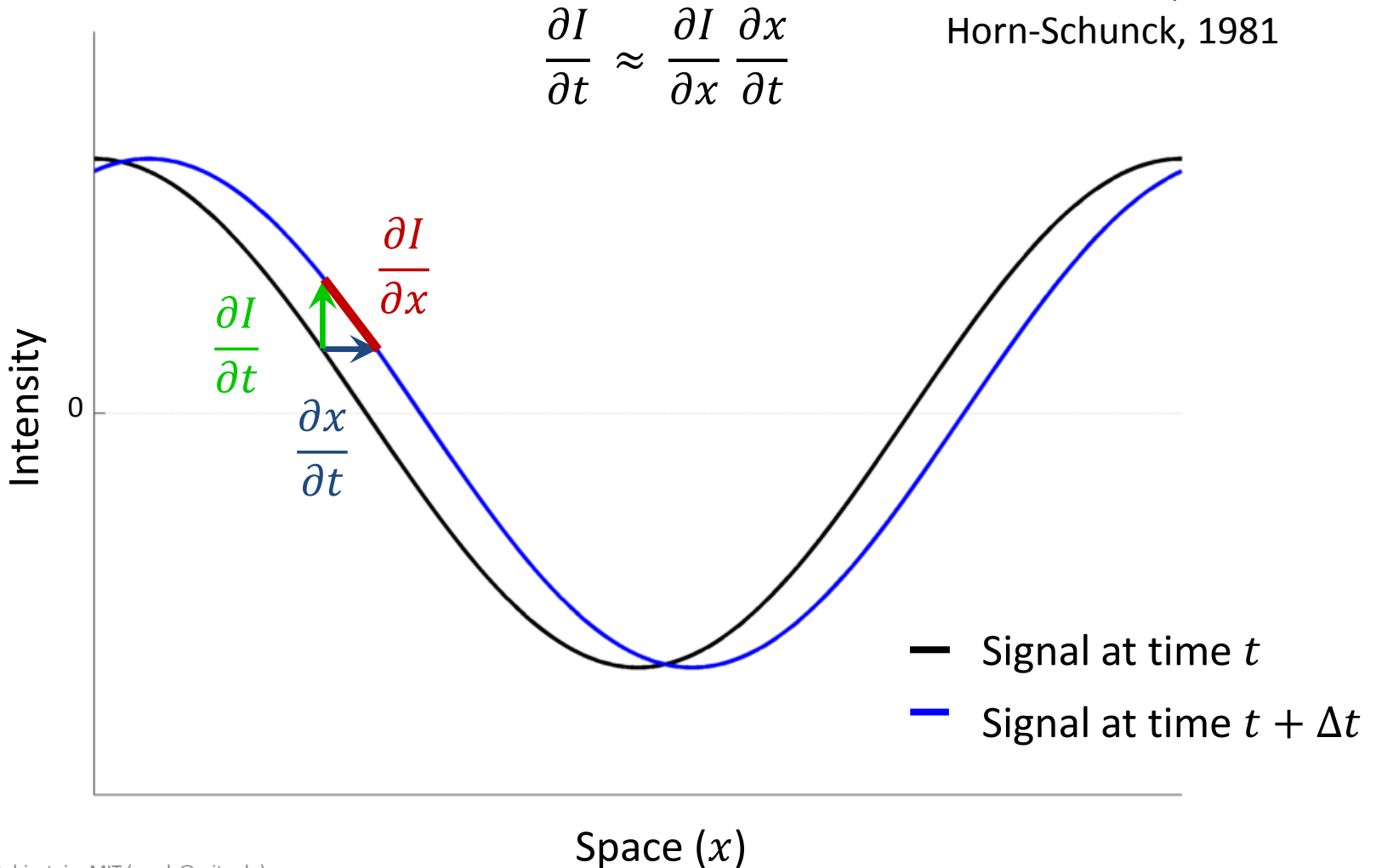
- Scenario: a 1D translating image profile



Differential Brightness Constancy

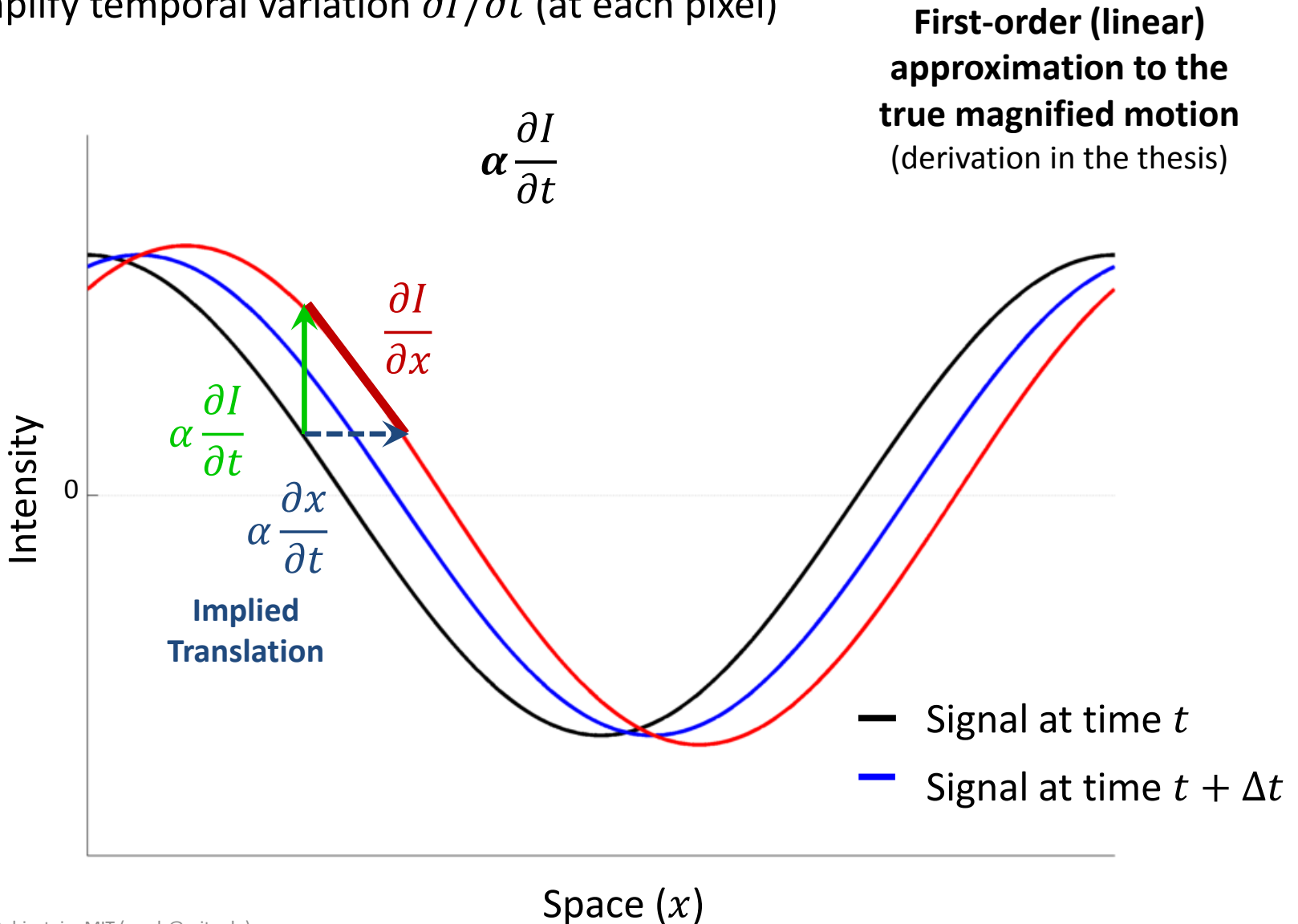
- Measure temporal variation $\partial I / \partial t$ (at each pixel)

Lucas-Kanade, 1981
Horn-Schunck, 1981

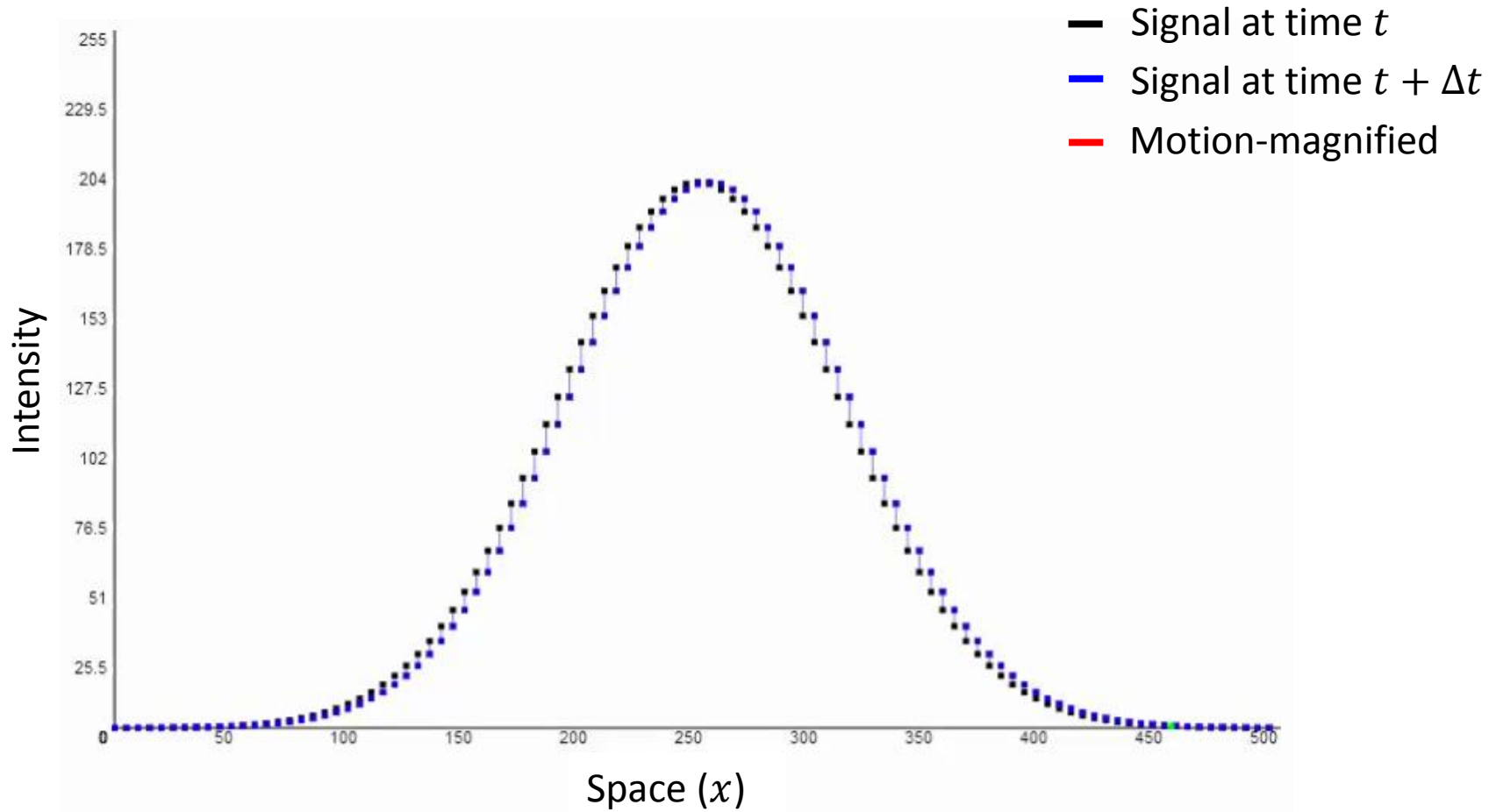


Eulerian Motion Magnification

- Amplify temporal variation $\frac{\partial I}{\partial t}$ (at each pixel)

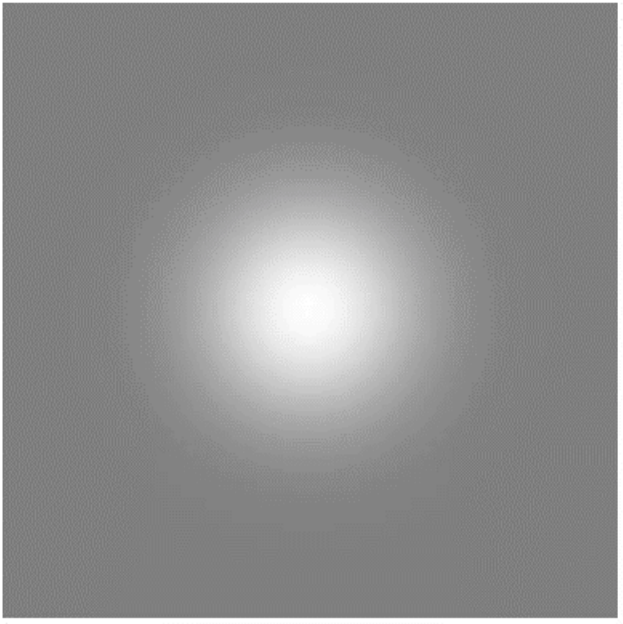


Relating Temporal and Spatial Changes



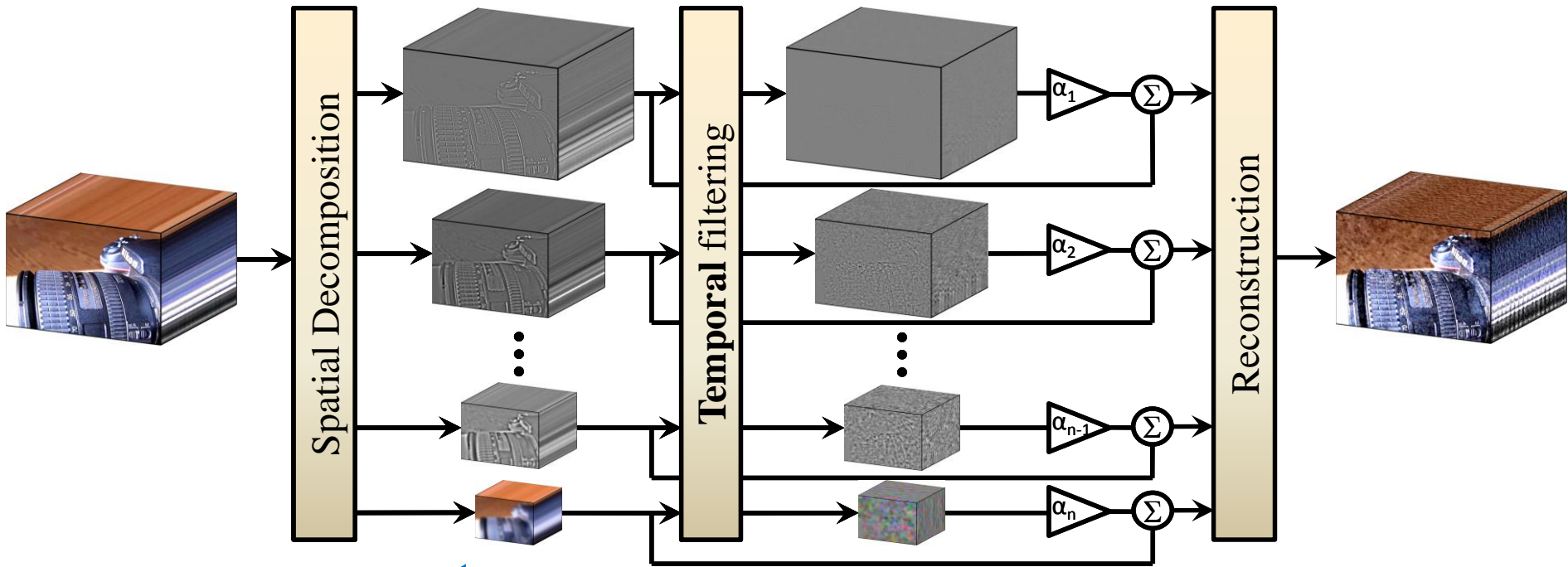
Courtesy of Lili Sun

Synthetic 2D Example



Source

Method Pipeline



Laplacian pyramid
[Burt and Adelson 1983]

Motion Magnification Results



Source



Selective Motion Magnification

Source
(600 fps)



72-92 Hz
Amplified



← Low E (82.4 Hz)

100-120 Hz
Amplified



← A (110 Hz)

Related Work: Motion Magnification [Liu 2005]



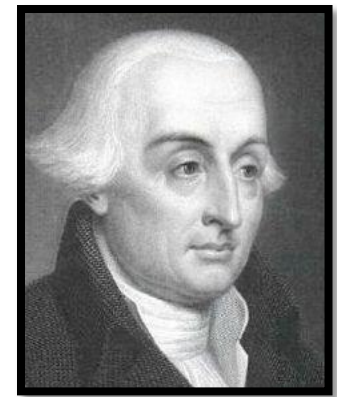
Source



Motion-magnified

Liu et al. *Motion Magnification*, SIGGRAPH 2005

Related Work: Motion Magnification [Liu 2005]



- Better for large motions, point features, occlusions, but...
- Requires motion analysis, motion segmentation, inpainting
 - Nontrivial to do artifact-free
 - Computationally intensive



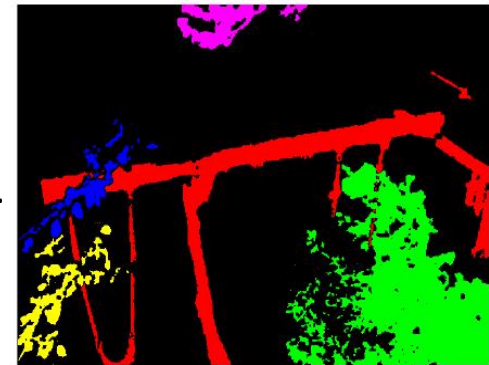
(a) Registered input frame

+



(b) Clustered trajectories of tracked features

+



(c) Layers of related motion and appearance

+



(d) Motion magnified, showing holes

+



(e) After texture in-painting to fill holes

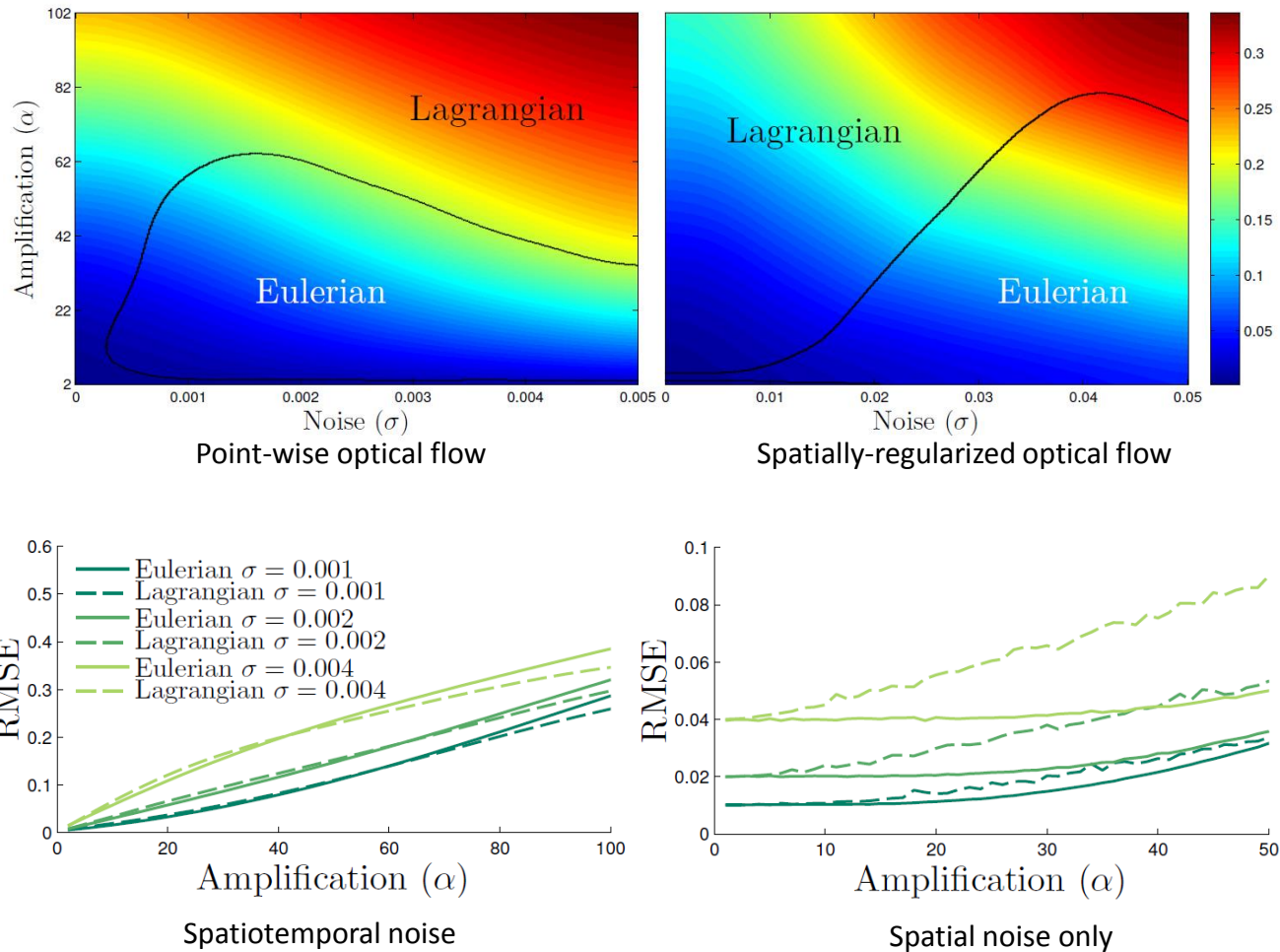
+



(f) After user's modifications to segmentation map in (c)

Lagrangian vs. Eulerian

- See my thesis for more details!



Limitations

Amplified noise

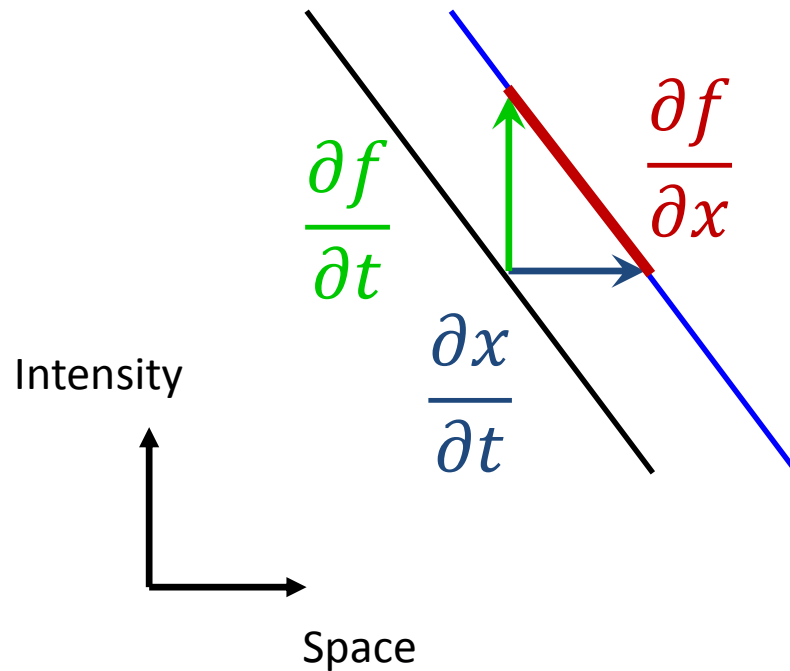


Source

Motion-magnified

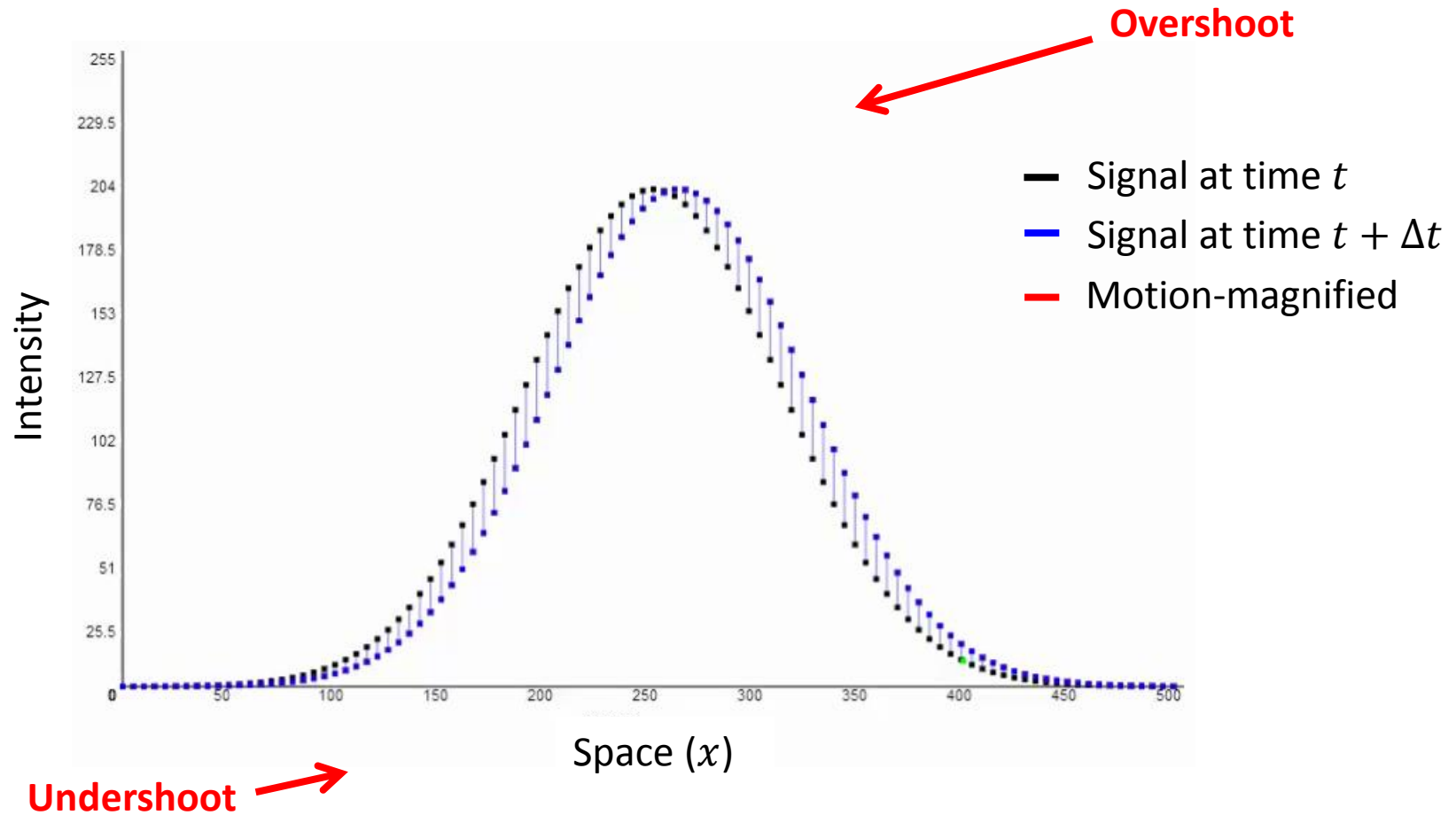
Limitations of Linear Motion Processing

- Assumes image intensity is locally linear



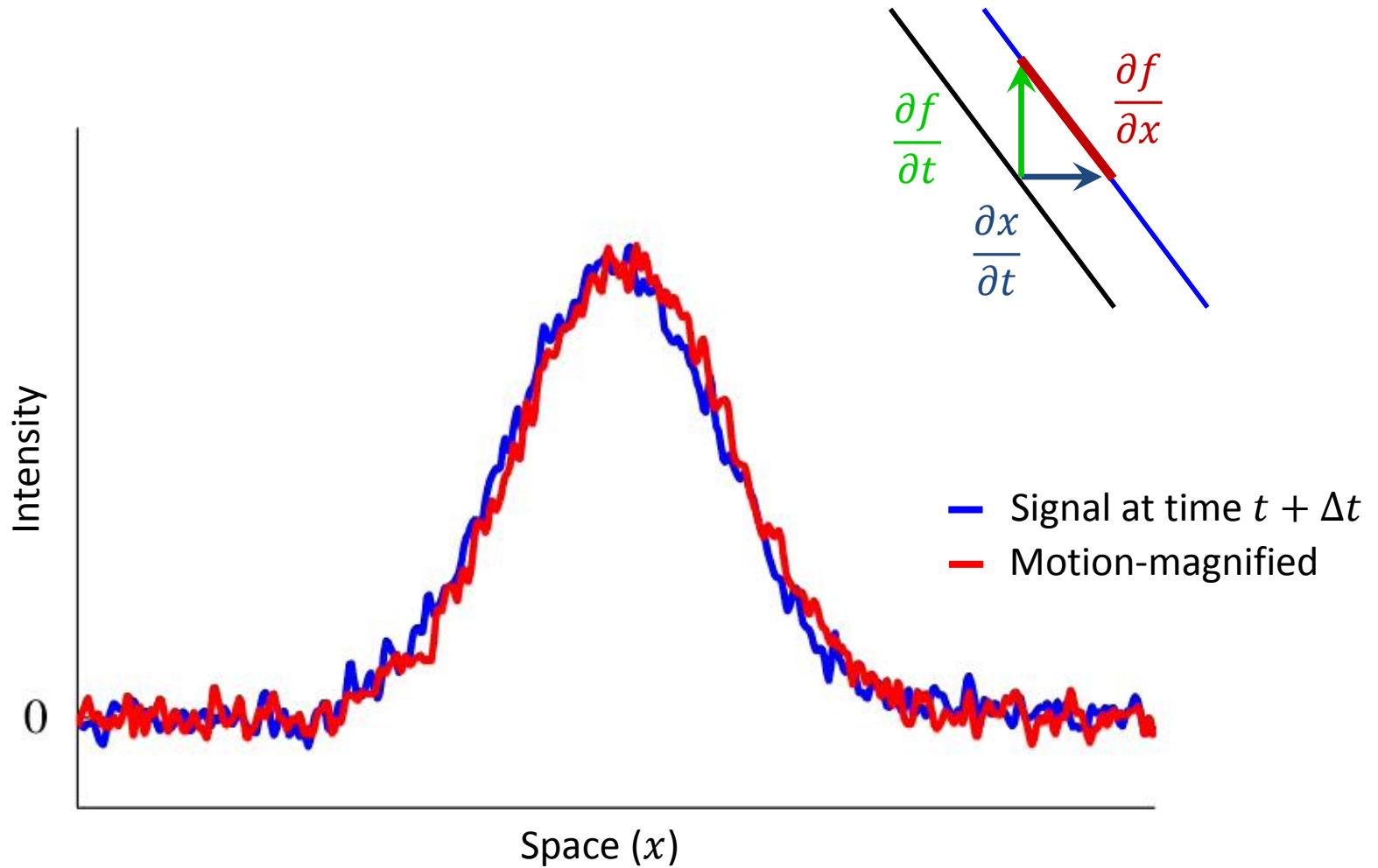
Limitations of Linear Motion Processing

- Breaks at high spatial frequencies and large motions



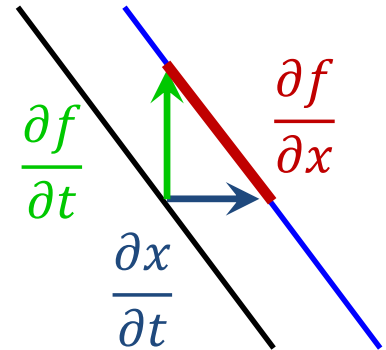
Limitations of Linear Motion Processing

- Noise amplified with signal

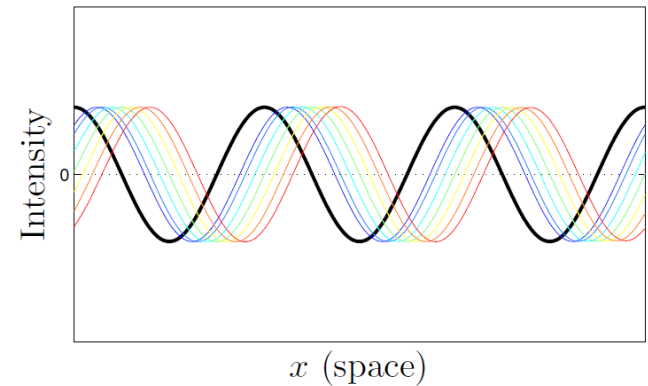


Linear vs. Phase-Based Motion Processing

- Linear motion processing
 - Assumes images are locally linear
 - Translate by **changing intensities**



- NEW phase-based motion processing
 - Represents images as collection of local sinusoids
 - Translate by **shifting phase**

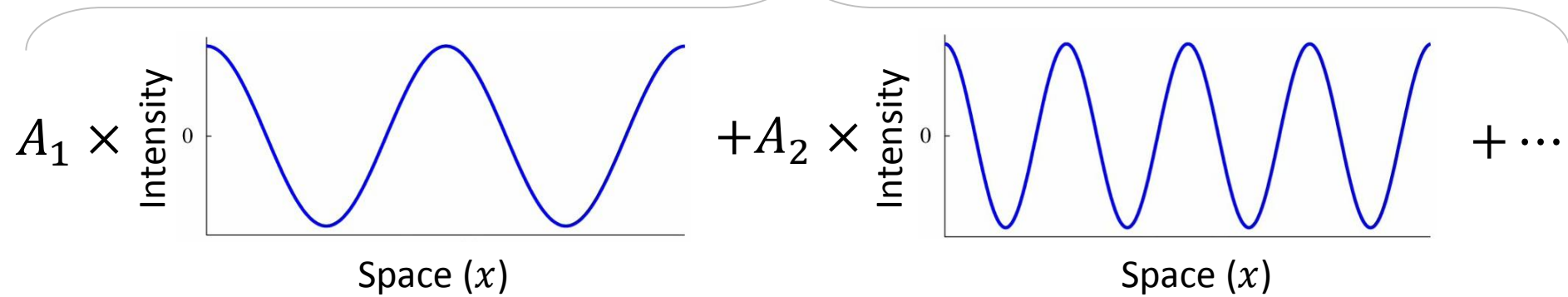
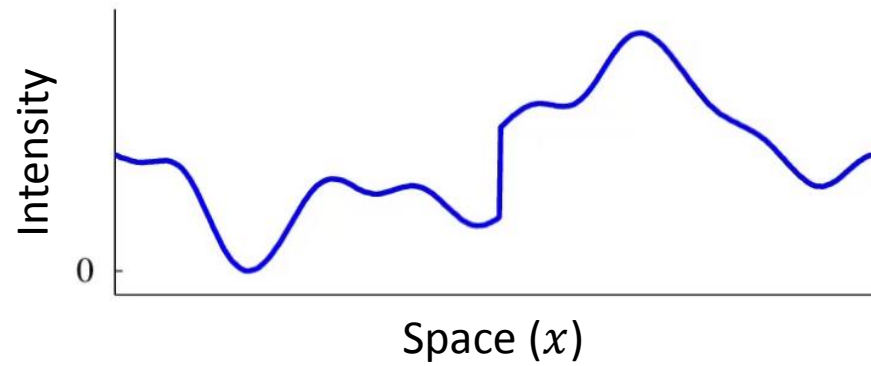


Phase-Based Video Motion Processing (SIGGRAPH 2013)

With Neal Wadhwa, Fredo Durand, Bill Freeman

Fourier Decomposition

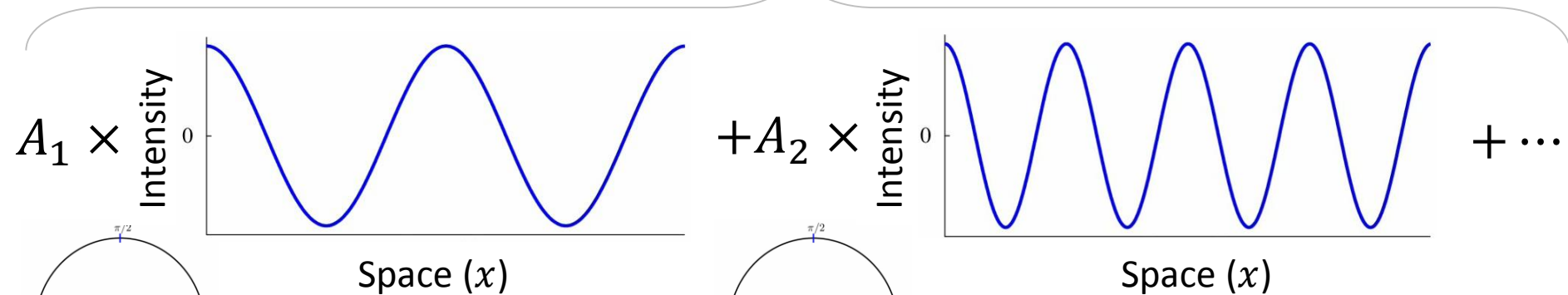
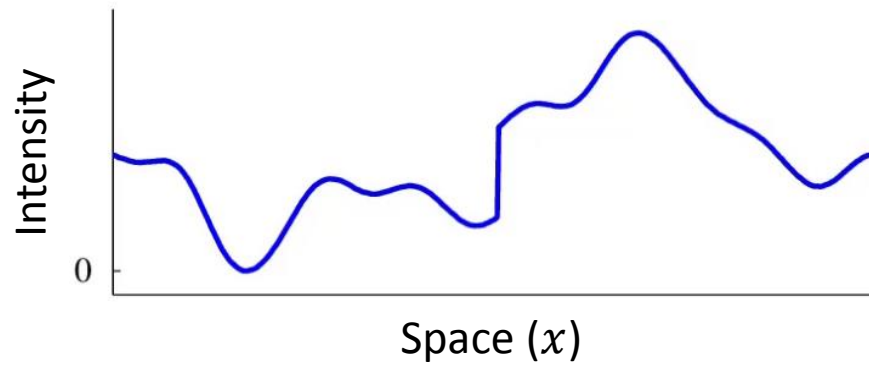
$$I(x) \quad \overset{\text{FFT}}{\rightleftharpoons} \quad \sum_{\omega=-\infty}^{\infty} A_{\omega} e^{i\omega x}$$



Fourier Shift Theorem

Phase shift \Leftrightarrow Translation

$$I(x - \delta) \stackrel{\text{FFT}}{\Leftrightarrow} \sum_{\omega=-\infty}^{\infty} A_{\omega} e^{i\omega x} e^{-i\omega\delta}$$

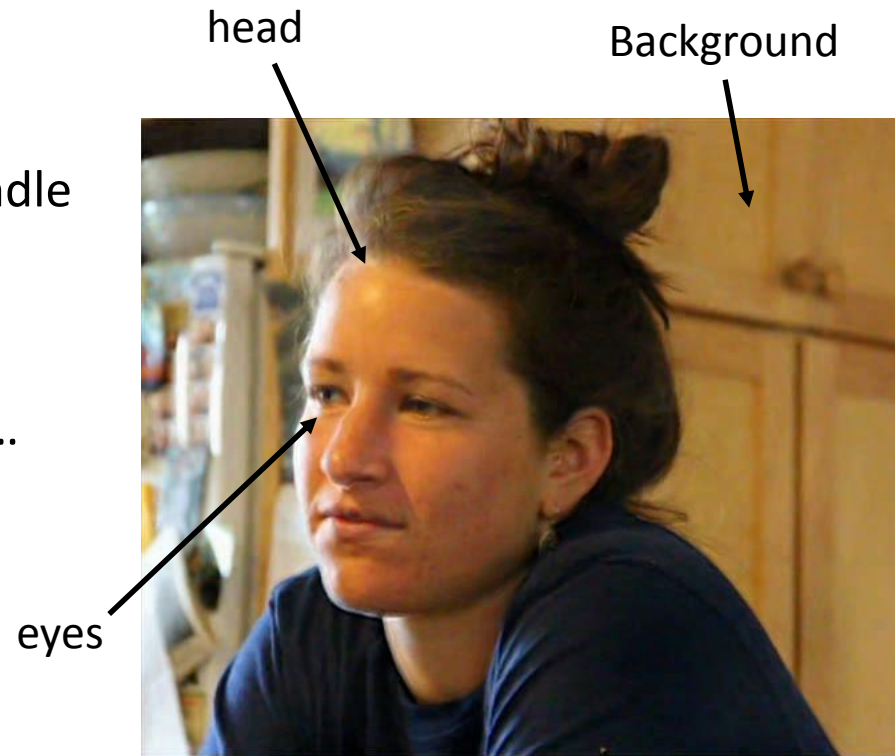


Phase ($e^{-i\delta}$)

Phase ($e^{-i2\delta}$)

Local Motions

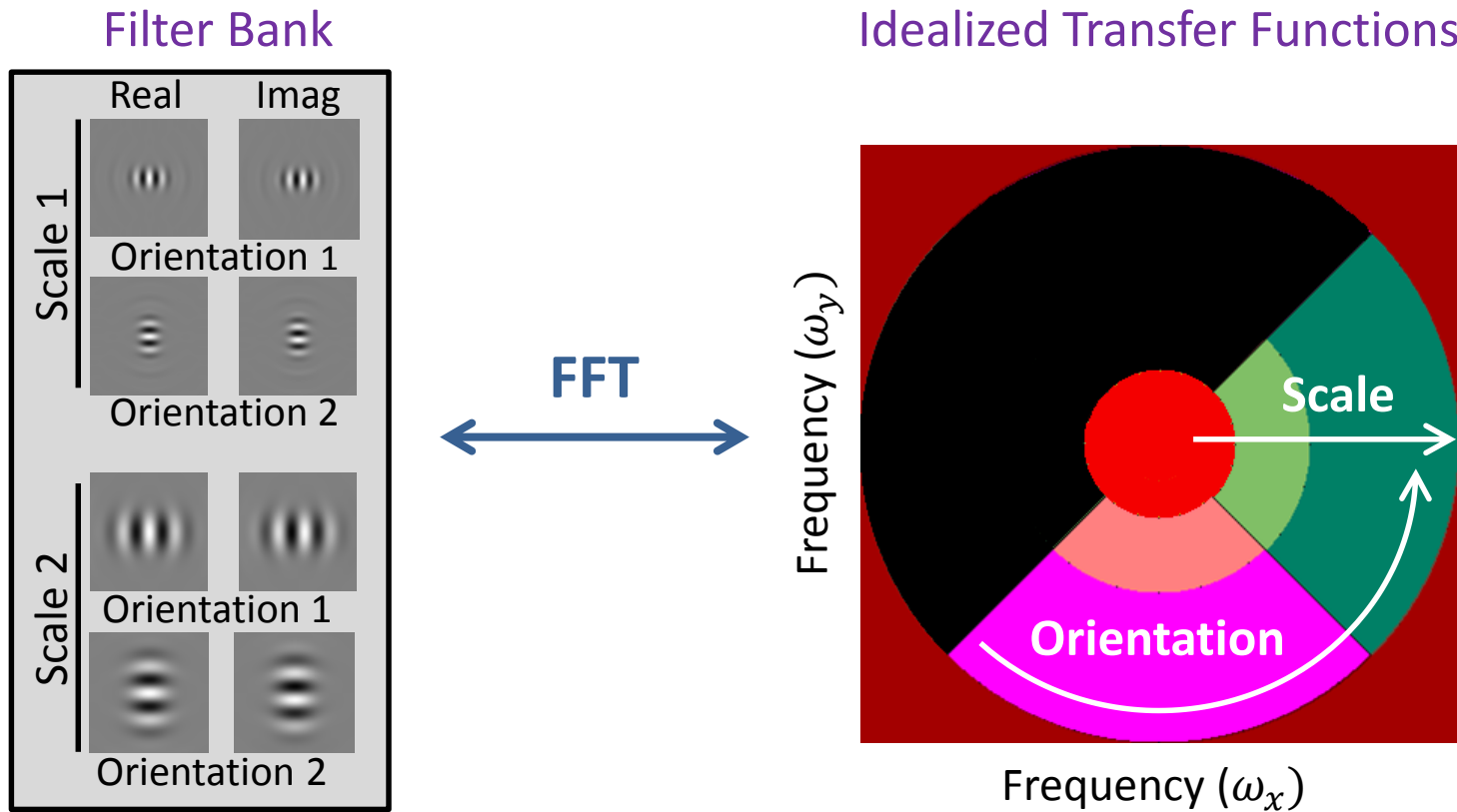
- Fourier shift theorem only lets us handle **global** motion
- But, videos have many local motions...



→ Need a localized Fourier Series for **local** motion

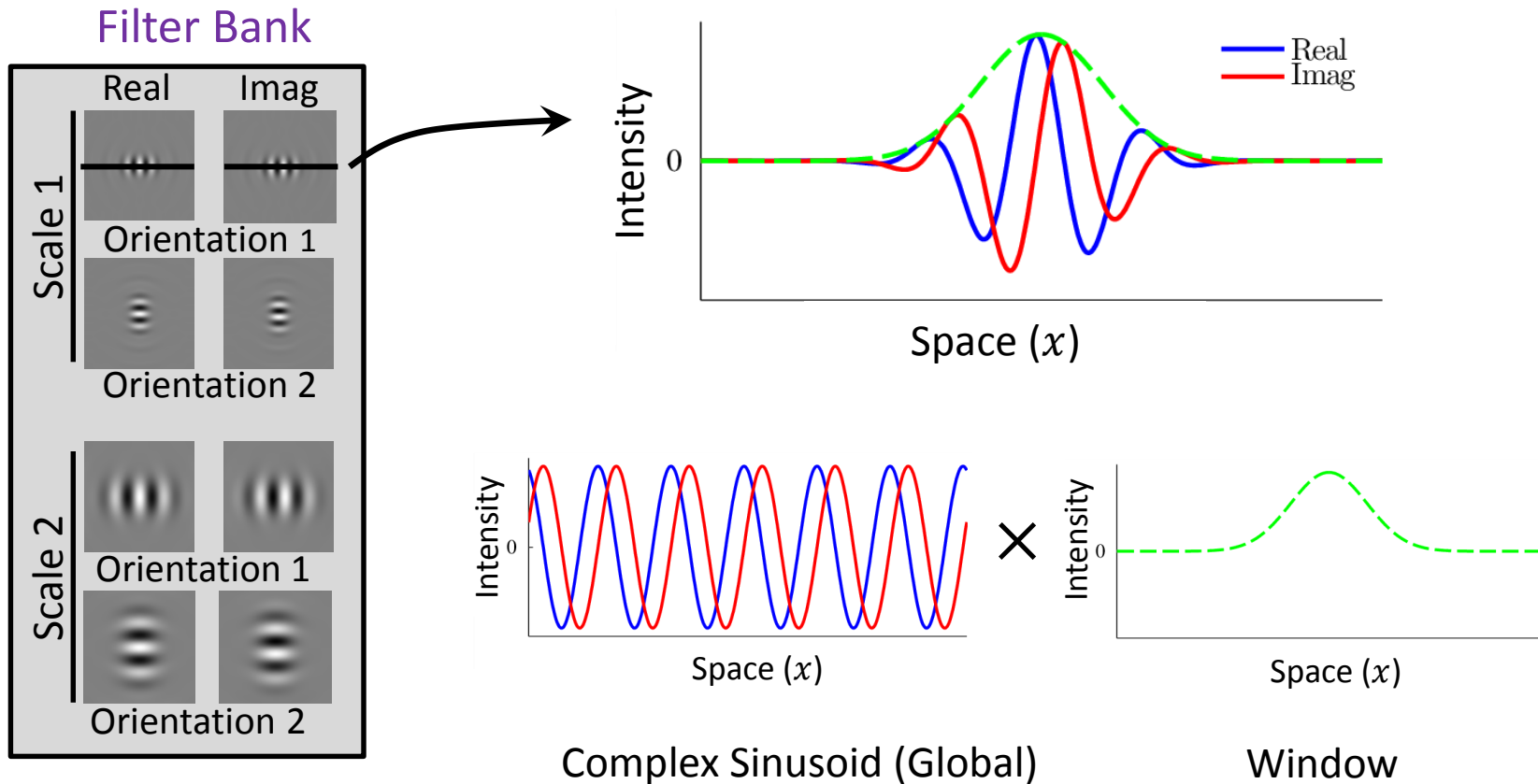
Complex Steerable Pyramid [Simoncelli, Freeman, Adelson, Heeger 1992]

- **Localized Fourier transform** that breaks the image into spatial structures at different scales and orientations



Complex Steerable Pyramid [Simoncelli, Freeman, Adelson, Heeger 1992]

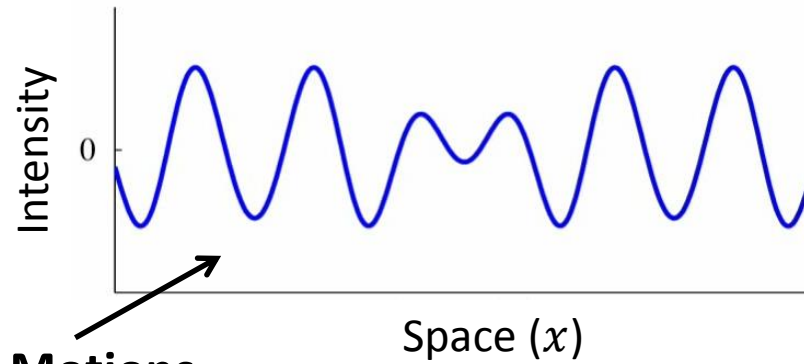
- Basis functions are wavelets with even (cosine) and odd (sine) components which give local amplitude and phase



Local Phase

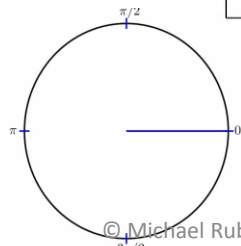
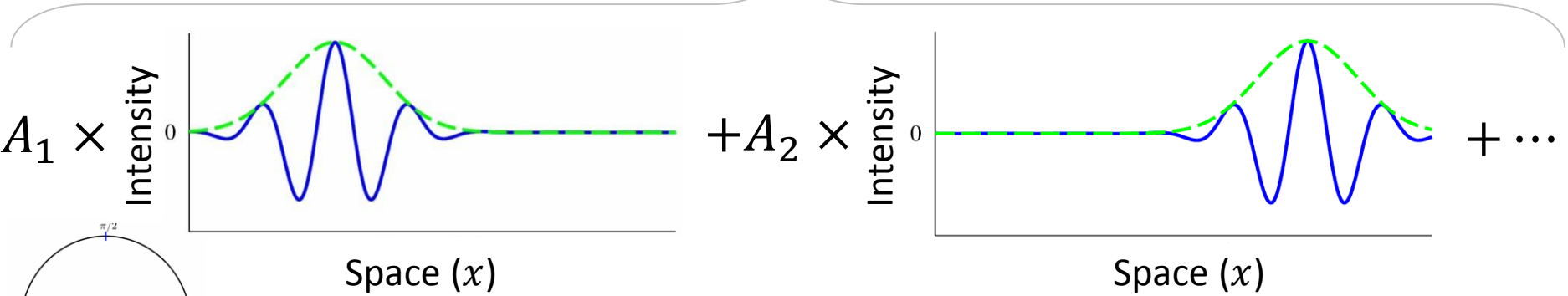
Local phase shift \leftrightarrow Local translation

- In a single subband, image is coefficients times translated copies of basis functions



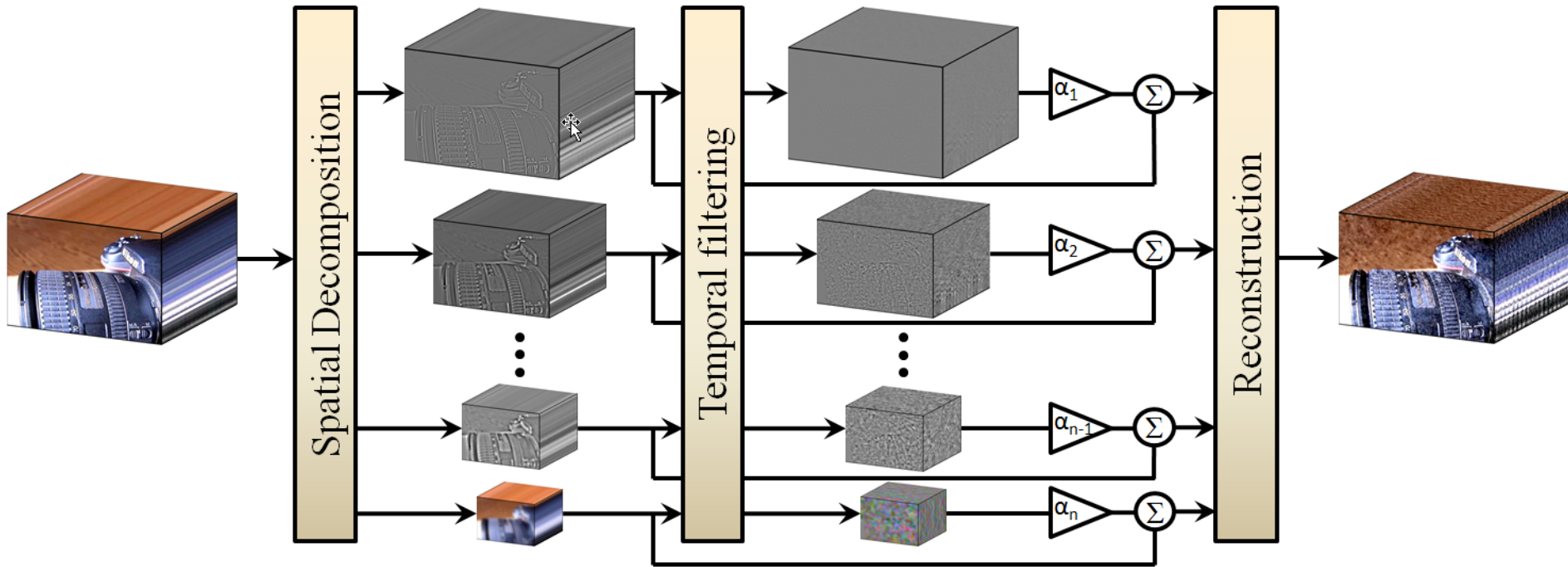
Local Motions

Space (x)



Local phase

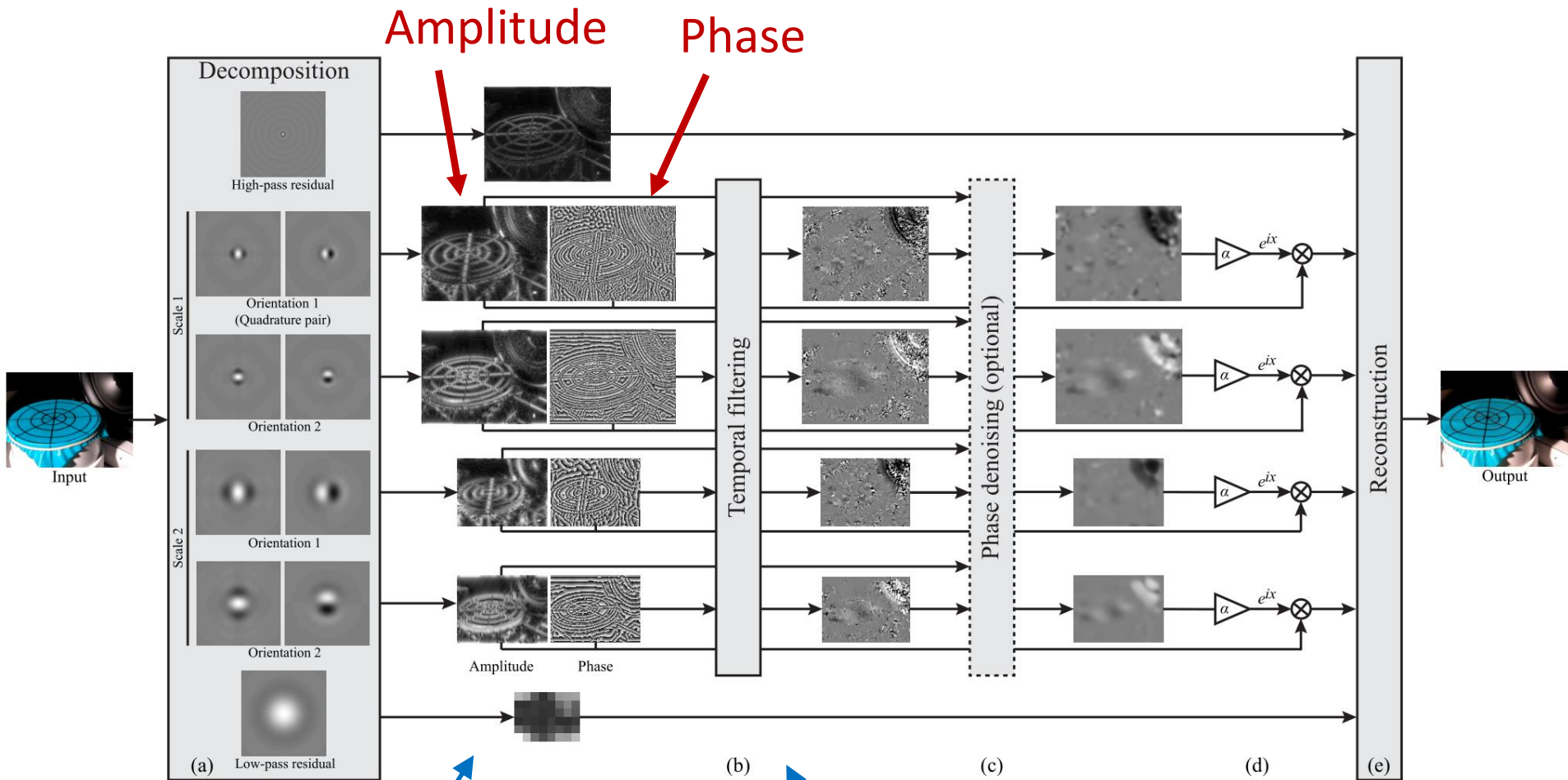
Linear Pipeline (SIGGRAPH'12)



Laplacian pyramid
[Burt and Adelson 1983]

Temporal filtering on intensities

Phase-based Pipeline (SIGGRAPH'13)



Complex steerable pyramid

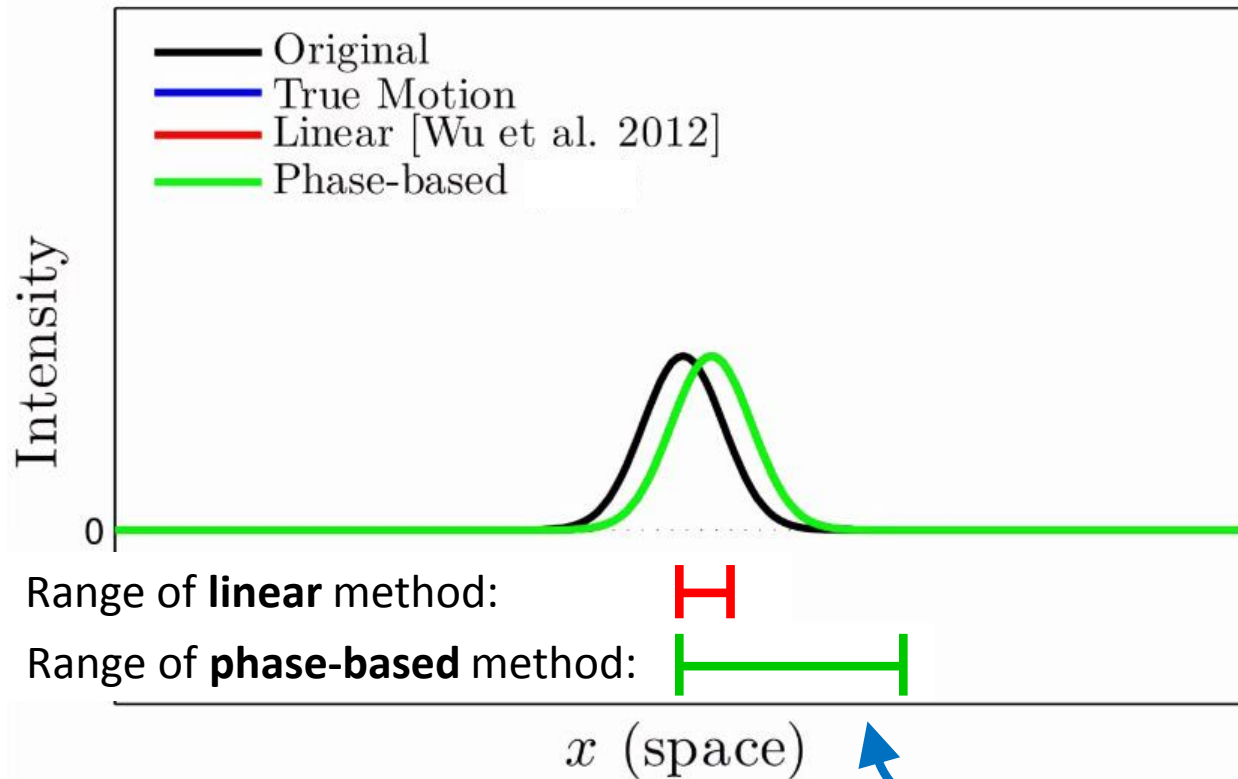
[Simoncelli, Freeman, Adelson, Heeger 1992]

[Portilla and Simoncelli 2000]

Temporal filtering on phases

Improvement #1: More Amplification

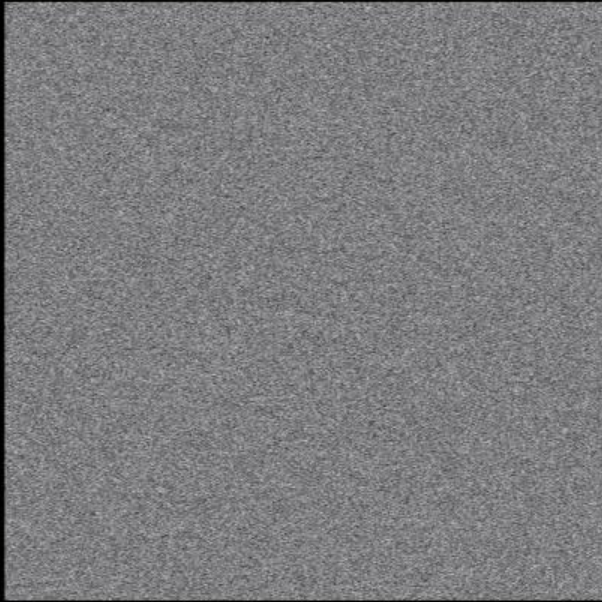
Amplification factor $\rightarrow \alpha = 0.0, \delta = 0.1 \leftarrow$ Motion in the sequence



4 times the amplification!
(derivation in the thesis)

Improvement #2: Better Noise Performance

Example of motion-magnifying Gaussian white noise



Source
(IID noise, std=0.1)

Noise amplified

Noise translated

Results: Phase-based vs. Linear

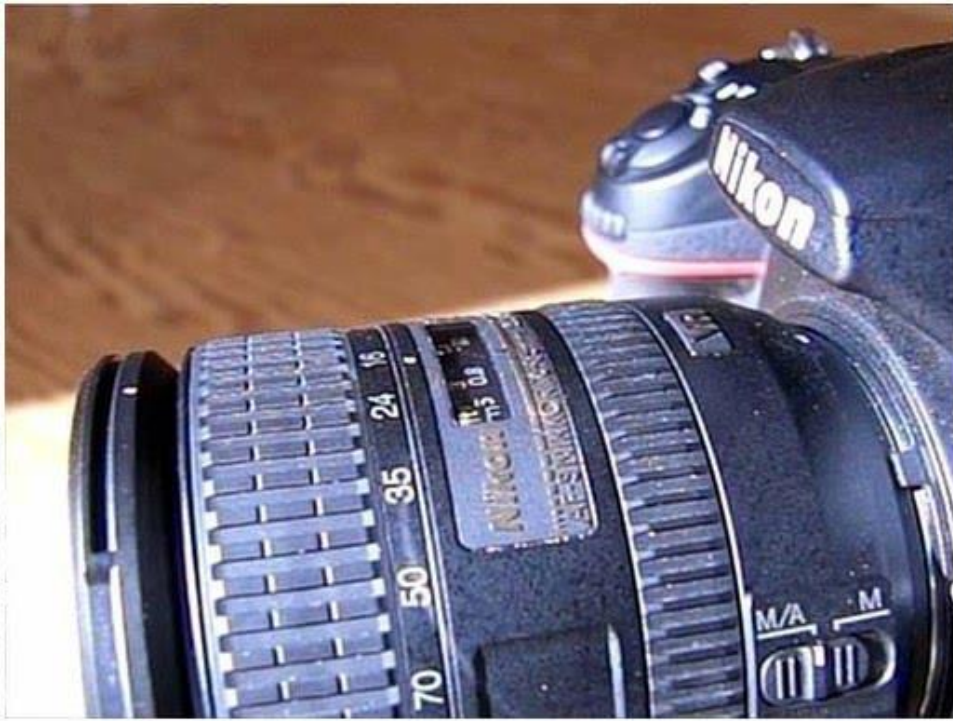
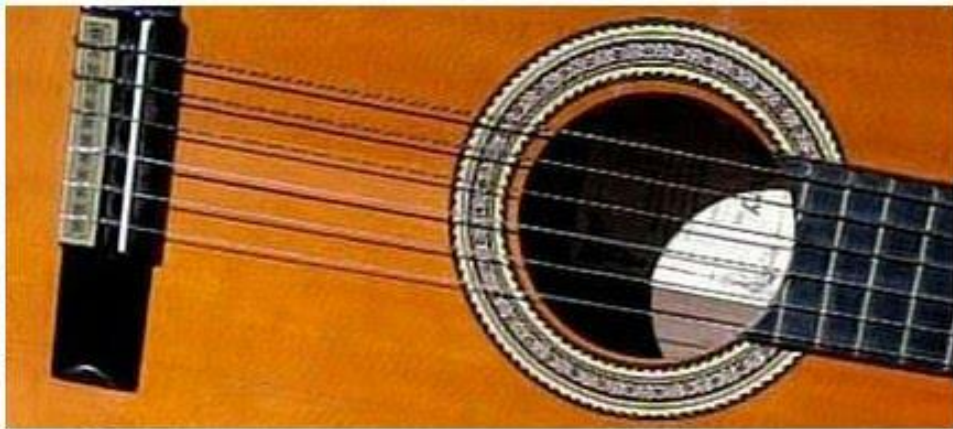


Linear (SIGGRAPH'12)
Motions amplified x10



Phase-based (SIGGRAPH'13)
Motions amplified x10

Results: Phase-based vs. Linear



Car Engine

Source



Car Engine

22Hz Magnified



Car Engine

Source



Car Engine

22Hz Magnified



Vibration Modes



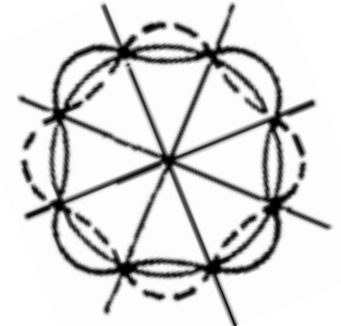
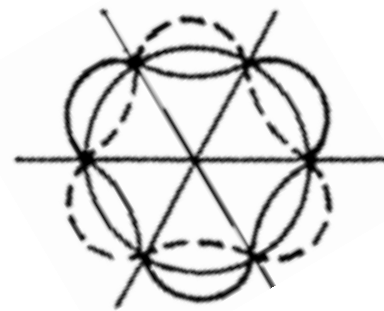
Source (20000 FPS)

480Hz (x200)

1200Hz (x400)

2400Hz (x1200)

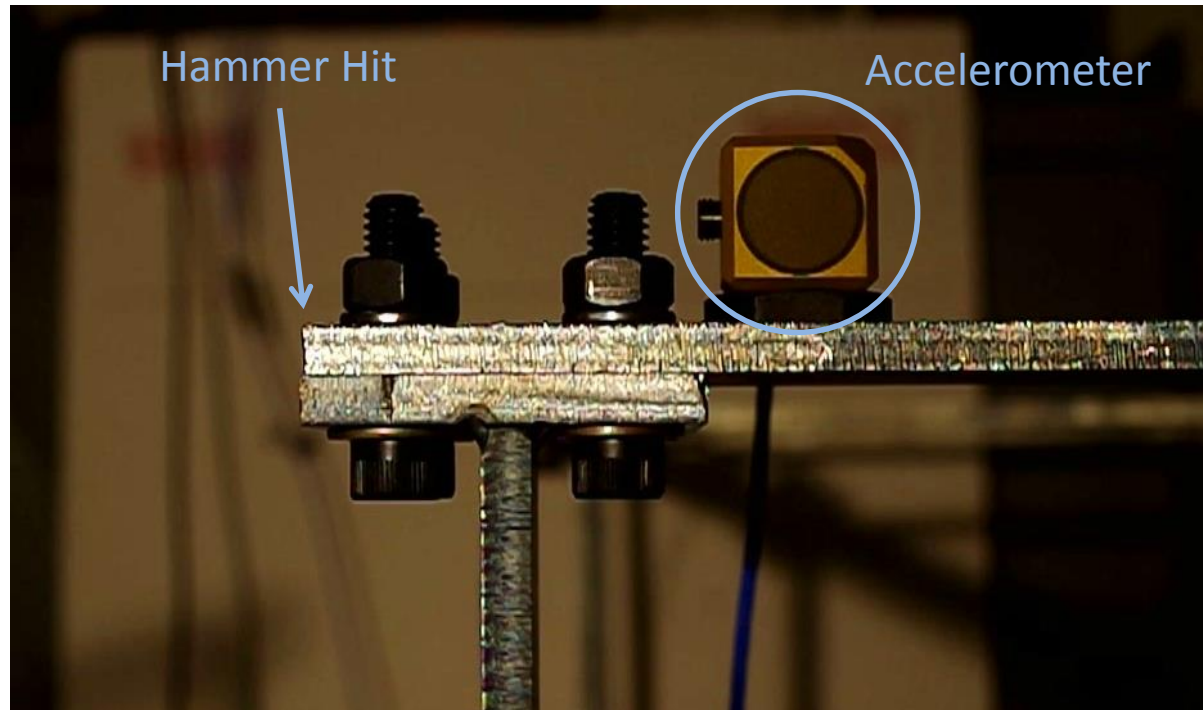
Sequence courtesy of Justin Chen



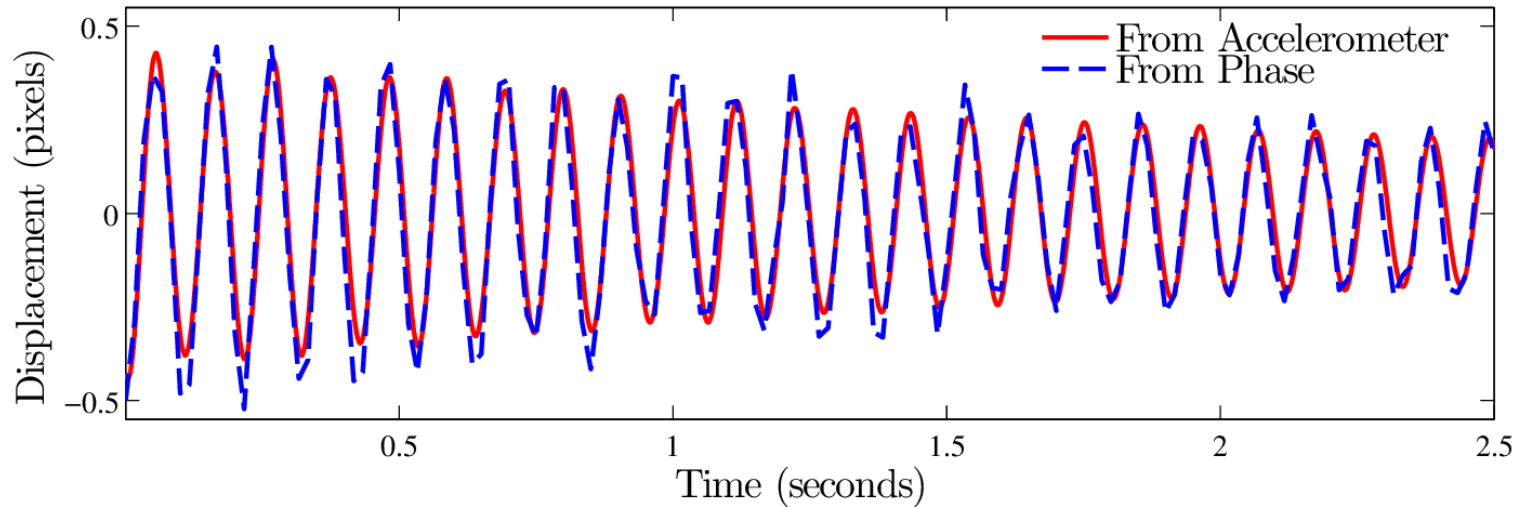
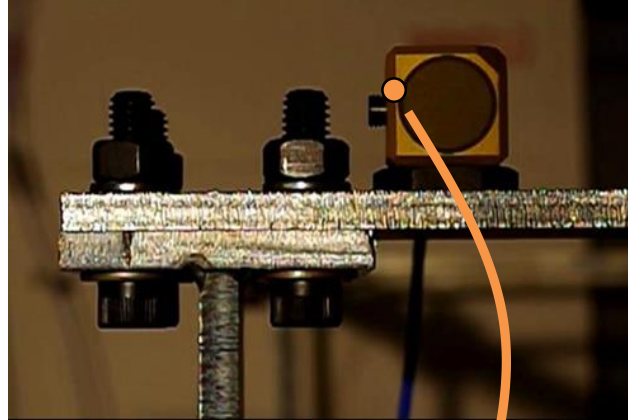
“Piping Vibration Analysis”
[Wachel et al. 1990]

Ground Truth Validation

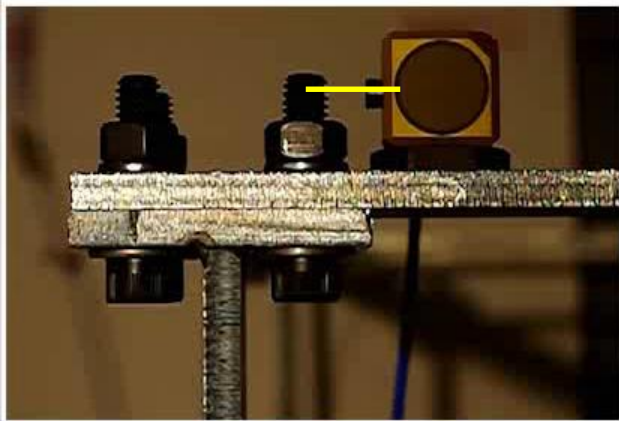
- Induce motion (with hammer)
- Record true motion with accelerometer



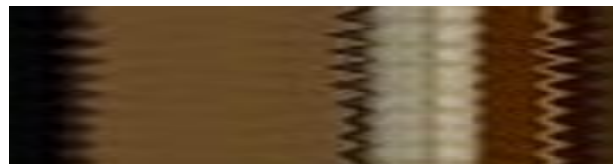
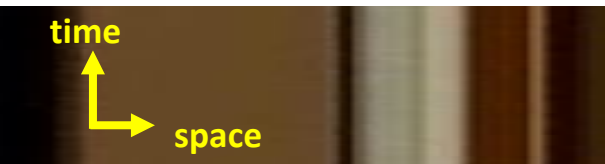
Ground Truth Validation



Qualitative Comparison



Input
(motion of 0.1 px)



Revealing Invisible Changes in the World

- NSF International Science and Engineering Visualization Challenge (SciVis), 2012
- Science Vol. 339 No. 6119 Feb 1 2013



Revealing Invisible Changes In The World

Created for the NSF International Science & Engineering
Visualization Challenge 2012

Talk Outline

- Removing distracting variations
 - Motion Denoising

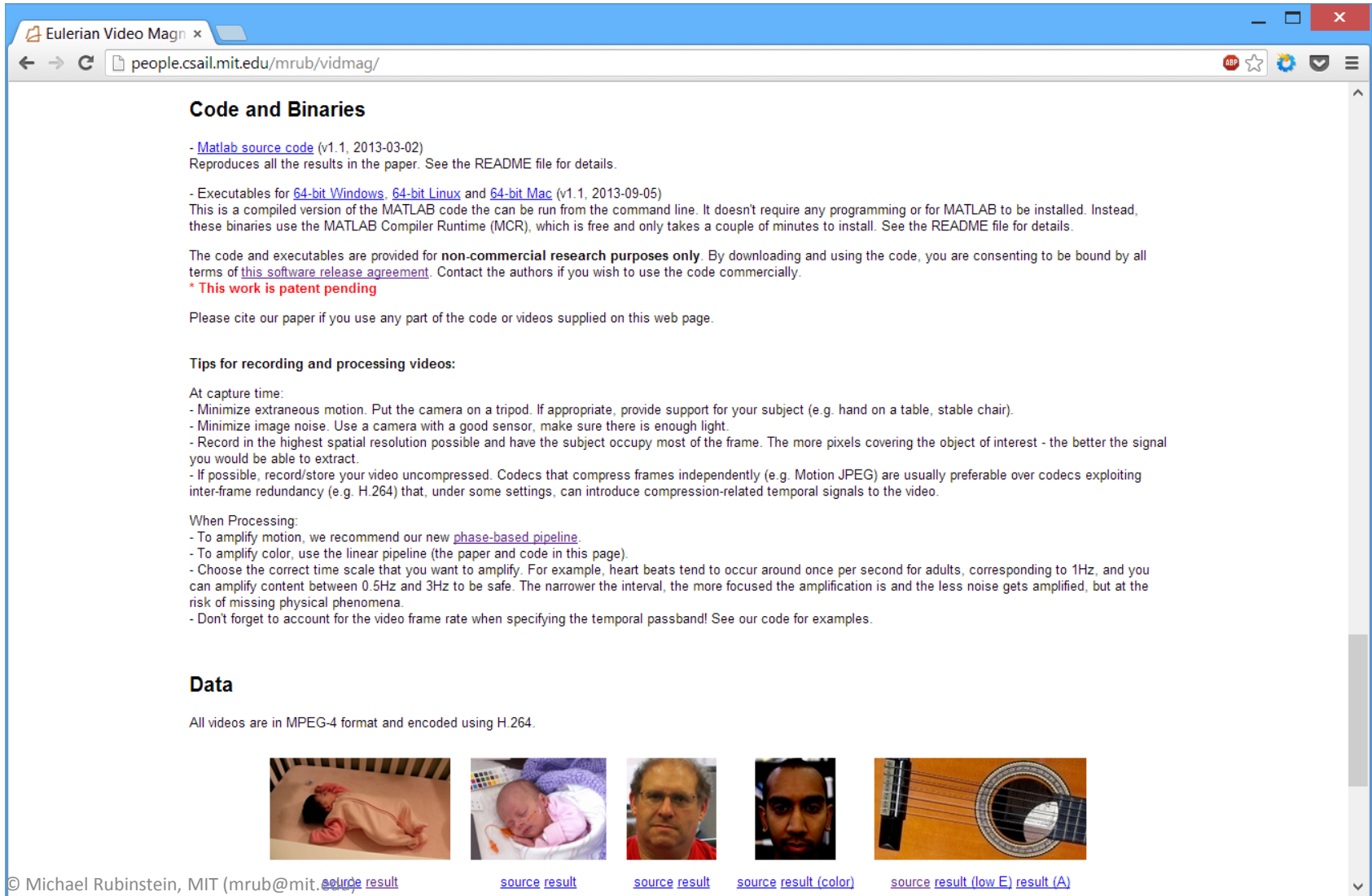
- Magnifying imperceptible variations
 - Eulerian Video Magnification
 - Phase-based Video Motion Processing

- Ongoing research and future work



Code Available

- Matlab code + executables <http://people.csail.mit.edu/mrub/vidmag/>



The screenshot shows a web browser window with the address bar containing `people.csail.mit.edu/mrub/vidmag/`. The page title is "Eulerian Video Magnification". The main content is titled "Code and Binaries" and includes the following text:

Code and Binaries

- [Matlab source code](#) (v1.1, 2013-03-02)
Reproduces all the results in the paper. See the README file for details.
- Executables for [64-bit Windows](#), [64-bit Linux](#) and [64-bit Mac](#) (v1.1, 2013-09-05)
This is a compiled version of the MATLAB code that can be run from the command line. It doesn't require any programming or for MATLAB to be installed. Instead, these binaries use the MATLAB Compiler Runtime (MCR), which is free and only takes a couple of minutes to install. See the README file for details.

The code and executables are provided for **non-commercial research purposes only**. By downloading and using the code, you are consenting to be bound by all terms of [this software release agreement](#). Contact the authors if you wish to use the code commercially.

* **This work is patent pending**

Please cite our paper if you use any part of the code or videos supplied on this web page.

Tips for recording and processing videos:

At capture time:

- Minimize extraneous motion. Put the camera on a tripod. If appropriate, provide support for your subject (e.g. hand on a table, stable chair).
- Minimize image noise. Use a camera with a good sensor, make sure there is enough light.
- Record in the highest spatial resolution possible and have the subject occupy most of the frame. The more pixels covering the object of interest - the better the signal you would be able to extract.
- If possible, record/store your video uncompressed. Codecs that compress frames independently (e.g. Motion JPEG) are usually preferable over codecs exploiting inter-frame redundancy (e.g. H.264) that, under some settings, can introduce compression-related temporal signals to the video.

When Processing:

- To amplify motion, we recommend our new [phase-based pipeline](#).
- To amplify color, use the linear pipeline (the paper and code in this page).
- Choose the correct time scale that you want to amplify. For example, heart beats tend to occur around once per second for adults, corresponding to 1Hz, and you can amplify content between 0.5Hz and 3Hz to be safe. The narrower the interval, the more focused the amplification is and the less noise gets amplified, but at the risk of missing physical phenomena.
- Don't forget to account for the video frame rate when specifying the temporal passband! See our code for examples.

Data

All videos are in MPEG-4 format and encoded using H.264.



Below the text are five small images showing examples of video magnification results:

- 1. A baby in a crib.
- 2. A baby in a crib (magnified).
- 3. A man's face.
- 4. A man's face (magnified).
- 5. A guitar.

At the bottom of the page, there is a copyright notice: © Michael Rubinstein, MIT (mrub@mit.edu) and a row of links: [source result](#), [source result](#), [source result](#), [source result \(color\)](#), [source result \(low E\) result \(A\)](#).

“VideoScope” by Quanta Research Cambridge

<http://videoscope.qrclab.com/>




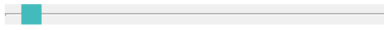
Home | Quanta | Projects | MIT CSAIL | People | Jobs | Press | Fun | Contact | Videoscope

User ID: **0b7f2be4-b8b6-464c-9ead-d68d10999661** | Current video: **baby2** [Return to chooser](#) | [Help](#)

Set frame rate (fps) [?](#)

Select magnification type [?](#) Color Motion

Set frequency range (Hz) [?](#) 

Set amplification [?](#) 

Select filter type [?](#) Ideal Butterworth IIR

Description (optional) [?](#)

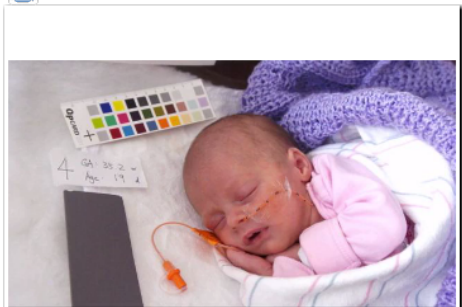
Show additional options...

Terms of Service [?](#) I agree to the [Terms of Service](#).

Process the video [?](#) Status: **Not running**.

We're interested in your feedback. If you have comments or suggestions, have an interesting application in mind, or would like to join our mailing list to learn about updates and new capabilities, [click here](#) or send email to videoscope@qrclab.com.

Original Video



© Michael Rubinstein, MIT (mrub@mit.edu)

EVM in the Wild: Pregnancy

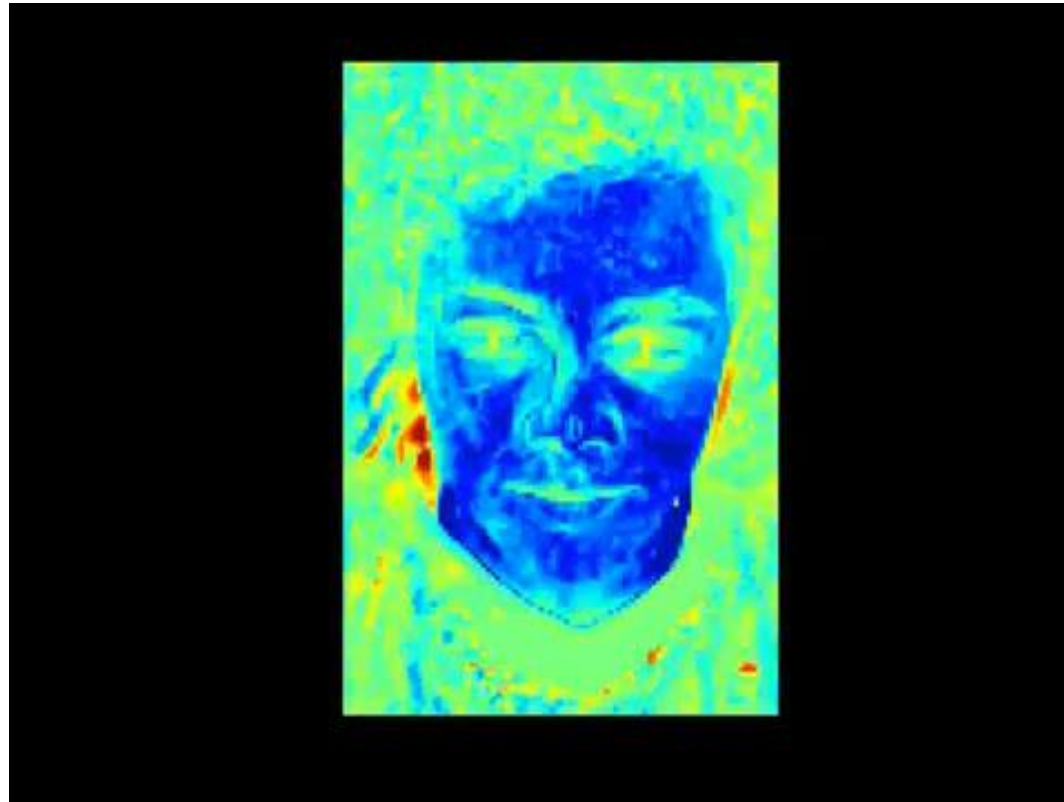


Original



Processed

EVM in the Wild: Blood flow Visualization



Red = high blood volume

Blue = low blood volume

Institute for Biomedical Engineering, Dresden Germany

<https://www.youtube.com/watch?v=Nb18CRVmXGY>

EVM in the Wild: Guinea Pig



Source

Motion-magnified

“SuperCreaturefan”: “Guinea pig Tiffany is the first rodent on Earth to undergo Eulerian Video Magnification.”

<http://www.youtube.com/watch?v=uXOSJvNwtIk>

EVM in the Wild: “Eulerian Profiles”

Source

Motion-amplified

Source

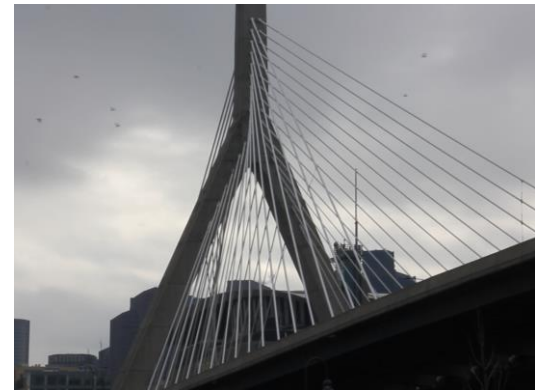
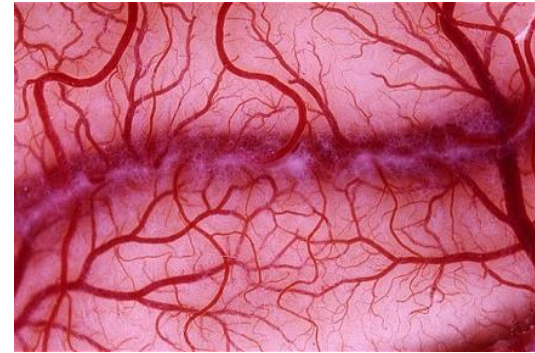
Motion-amplified



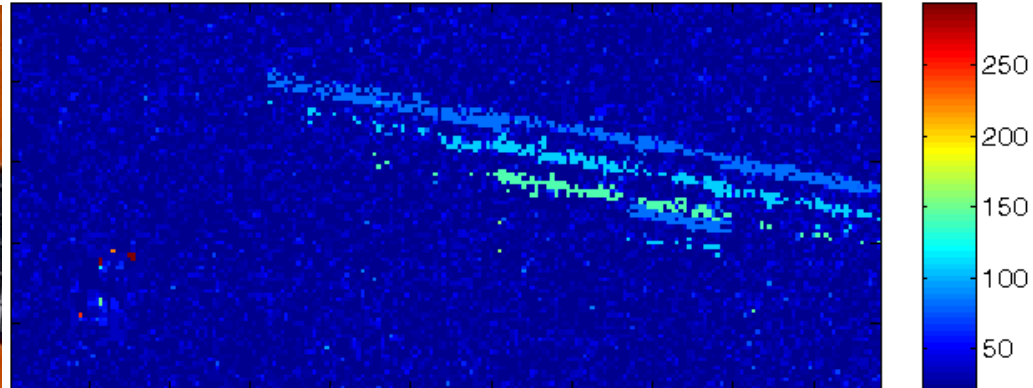
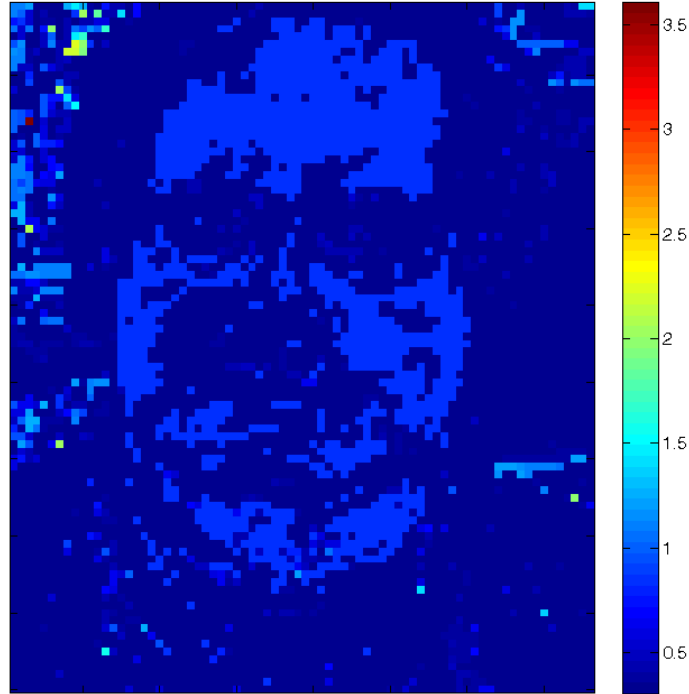
By Erin Knutson (Graphic Design student at Yale)

People Interested in...

- **Health care**
 - Contactless monitoring
 - Blood vessel identification
 - Tissue perfusion in plastic surgery
 - ...
- **Scientific analysis**
 - Changes in the earth's surface from satellite imagery
 - Seismic data
 - ...
- **Engineering**
 - Structural integrity of bridges, buildings
 - ...



Identifying Temporal Signals Automatically



Source frame

Dominant frequency

Seeing and Measuring Refractive Flow (hot air, gas)

- Small motions due to changes in refraction index (change in density, temp.)



Seeing and Measuring Sound

- Sound is fluctuations in air pressure traveling through space
- These pressure waves hit objects and make them vibrate
 - This is how we hear; this is how we record sound

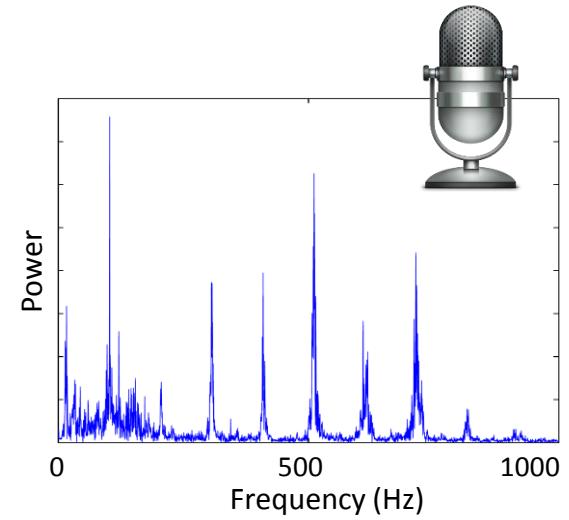
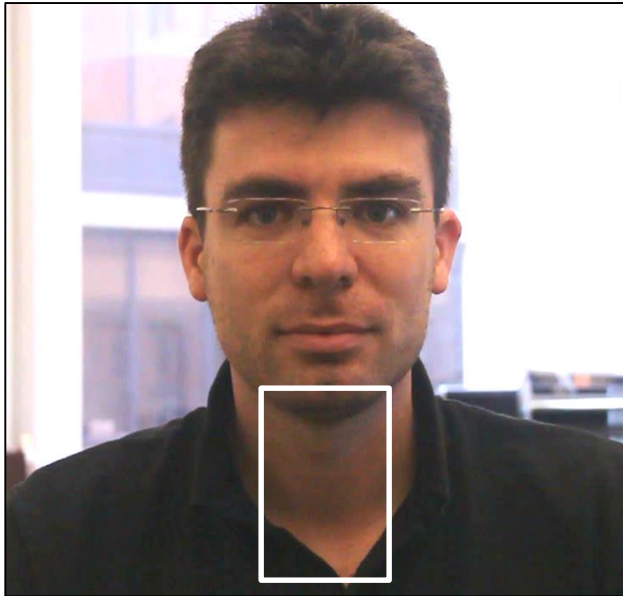


“water sound waves”

xsgianni, <http://www.youtube.com/watch?v=xPW3gihYnZE>

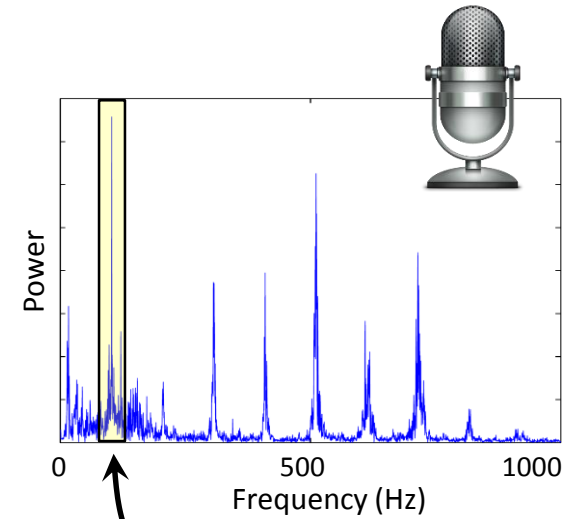
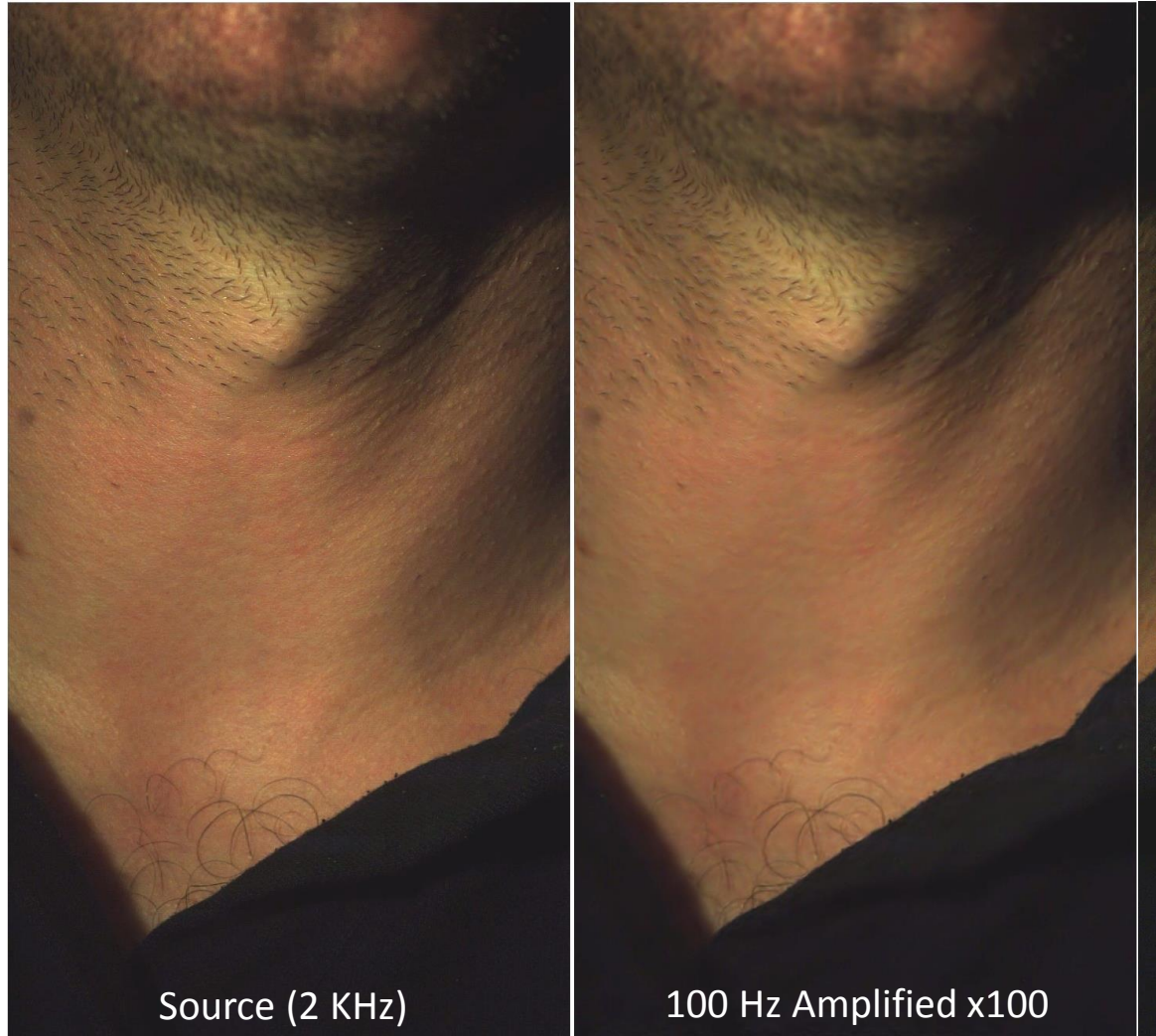
Neck Skin Vibrations

****Unpublished****



Neck Skin Vibrations

****Unpublished****



Fundamental
frequency: ~100Hz

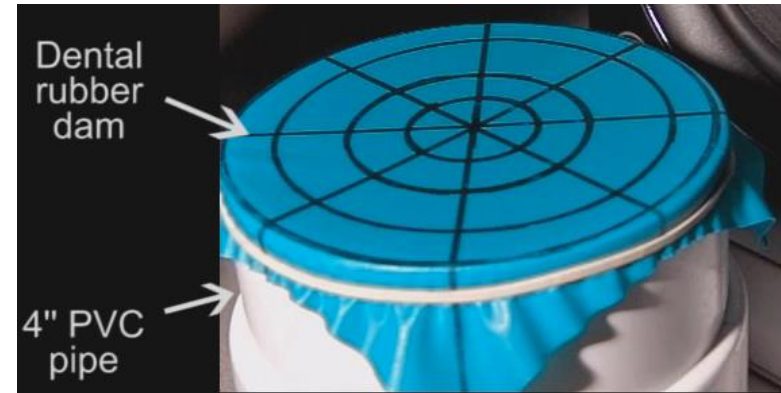
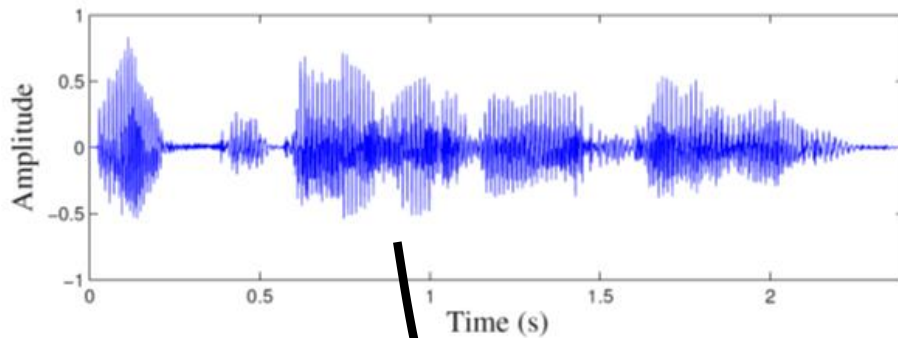
Can We Recover Sound From Video?

****Unpublished****

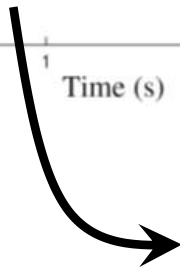


Recovering Sound from Video

- Assuming scene is static, motions should be well correlated with sound pressure waves

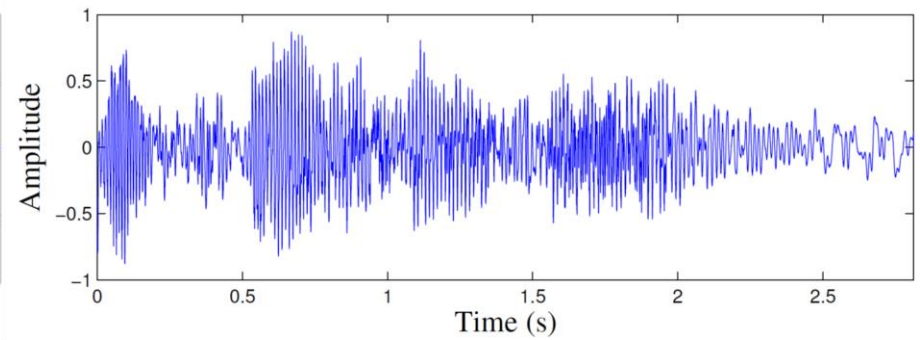
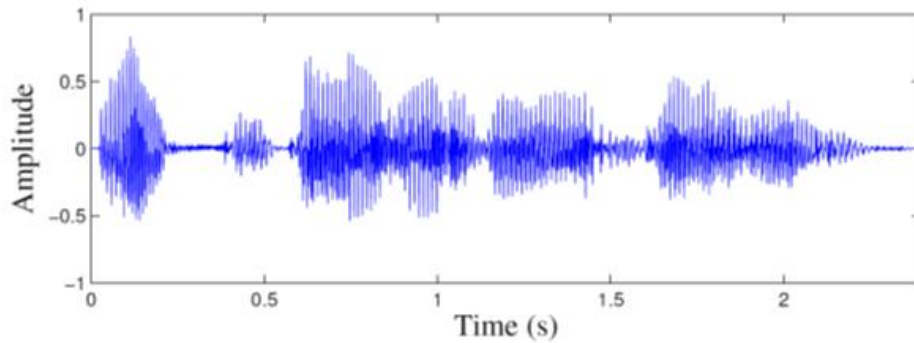


Low-passed 0-2KHz



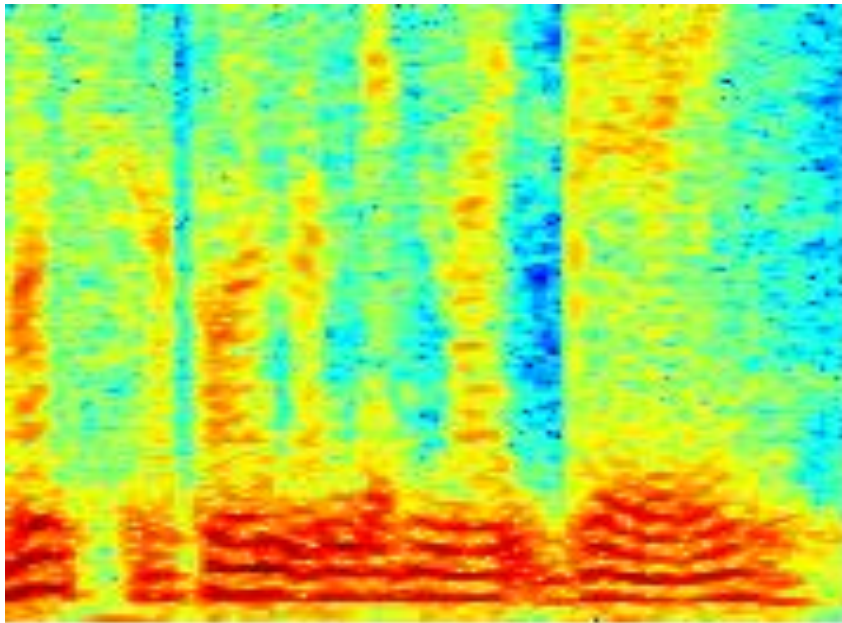
Recorded video (4 KHz)

Recovering Sound from Video



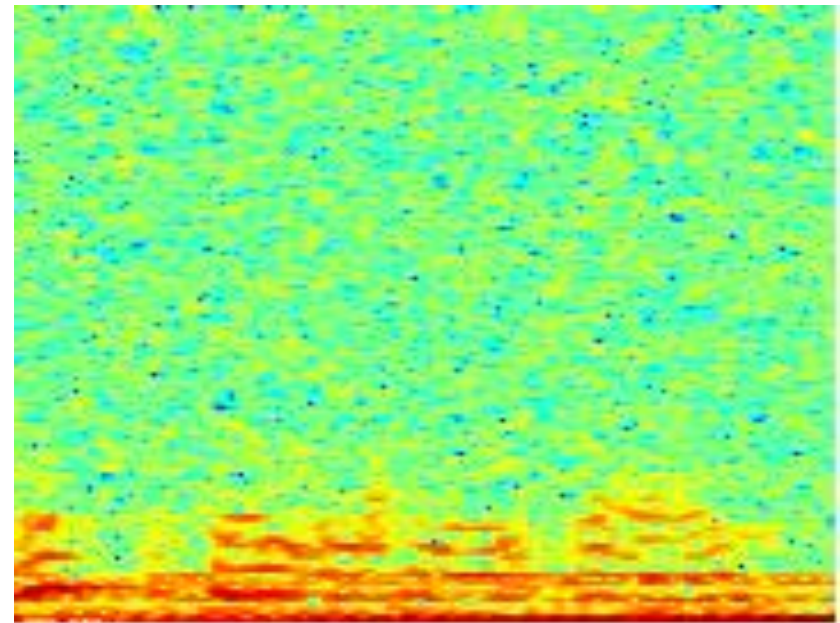
2Khz

Frequency



Time

Source



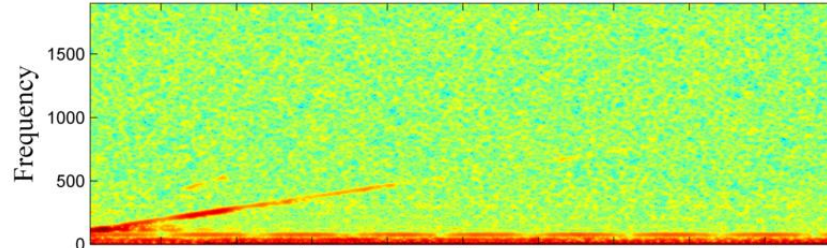
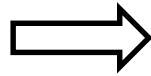
Time

Reconstructed

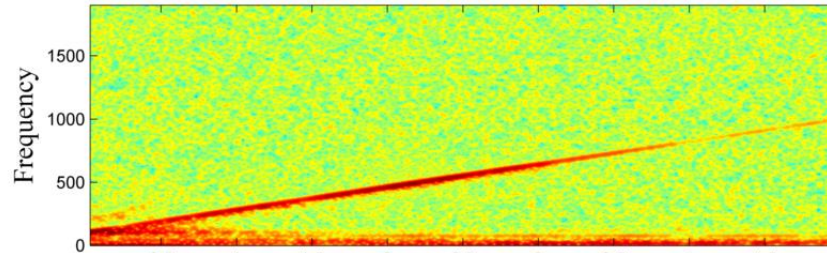
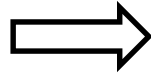
Natural Microphones

Reconstruction from:

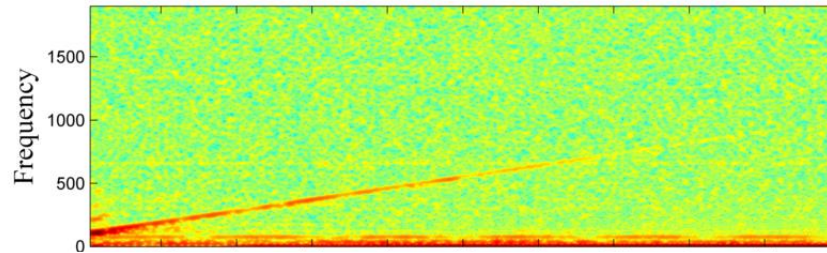
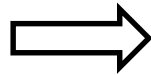
Water



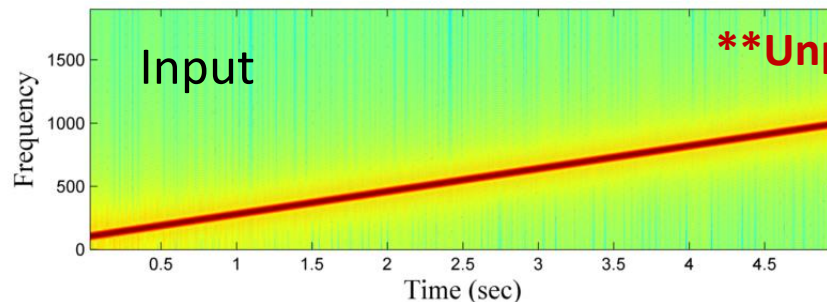
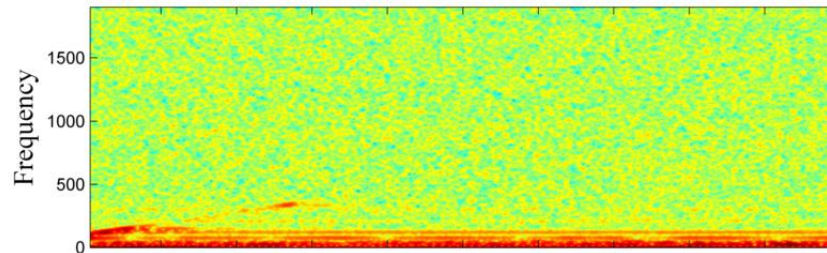
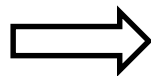
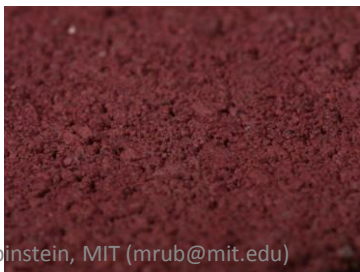
Latex membrane



Cardboard



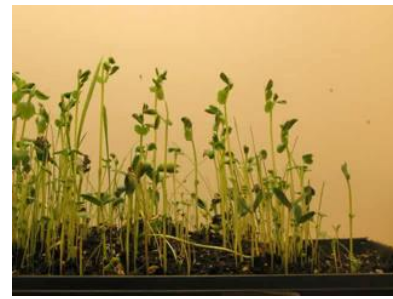
Brick



****Unpublished****

Conclusions

- We decompose temporal signals in videos into different components and re-render the video to analyze and visualize them separately
- **Removing distracting temporal variation**
 - **Motion denoising** – decomposition into long-term and short-term changes
 - No explicit motion analysis
- **Amplifying imperceptible temporal variation**
 - **Eulerian approaches** for representing, analyzing and visualizing small-amplitude temporal signals
 - No explicit motion analysis
 - The world is full of small, informative motions and changes we cannot normally see, and we can reveal them using regular video



Acknowledgements

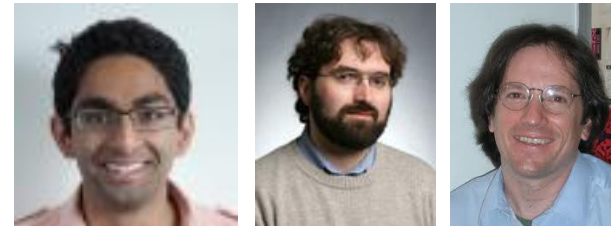
Motion Denoising
(CVPR 2011)



Eulerian Video Magnification
(SIGGRAPH 2012)



Phase-based Motion Processing
(SIGGRAPH 2013)



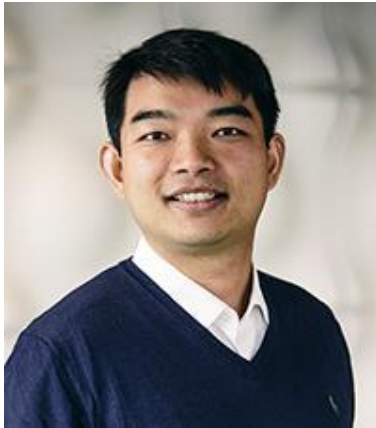
Refractive Flow



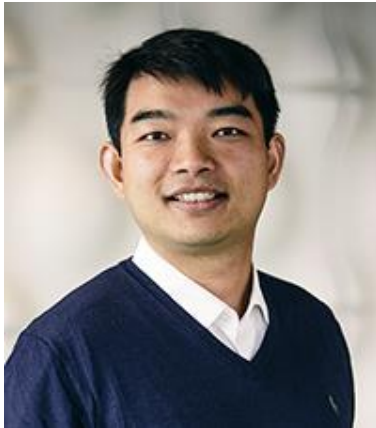
Visual Microphone



Acknowledgements



Acknowledgements



Acknowledgements



Acknowledgements



Acknowledgments



NSF CGV-1111415
"Images Through Time"





Massachusetts
Institute of
Technology



Thank you!

Michael Rubinstein

MIT CSAIL