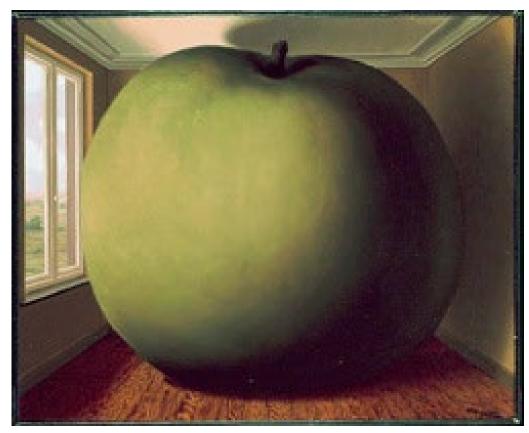
Video Magnification



Magritte, "The Listening Room"

Computational Photography Derek Hoiem, University of Illinois

This Class

- 1. Video Magnification
 - Lagrangian (point tracking) approach
 - Eulerian (signal within a pixel) approach
- 2. Video Microphone

Imperceptible Motions and Changes



[Wu et al. 2012]

MAGNIFIED Imperceptible Motions and Changes



[Wu et al. 2012]

Motion Magnification

Goal: exaggerate selected motions



Ideas?

Approach 1: Point Tracking

Motion Magnification (SIGGRAPH 2005)

Ce Liu Antonio Torralba William T. Freeman Frédo Durand Edward H. Adelson

Computer Science and Artificial Intelligence Laboratory

Massachusetts Institute of Technology

Following slides based on SG 2005 presentation: http://people.csail.mit.edu/celiu/motionmag/motionmag.html

Naïve Approach

- Magnify the estimated optical flow field
 Optical flow = displacement of each pixel between frames
- Render by warping



Original sequence

Magnified by naïve approach

Tracking-based Motion Magnification



(a) Registered input frame



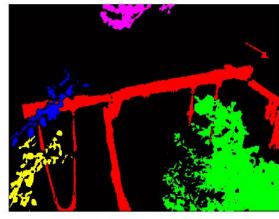
(d) Motion magnified, showing holes



(b) Clustered trajectories of tracked features



(e) After texture in-painting to fill holes



╋

(c) Layers of related motion and appearance



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

Liu et al. Motion Magnification, 2005

Robust Video Registration

- Find feature points with Harris corner detector on the reference frame
- Track feature points
- Select a set of robust feature points with inlier and outlier estimation (most from the rigid background)
- Warp each frame to the reference frame with a global affine transform

Feature tracking trick 1: Adaptive Region of Support

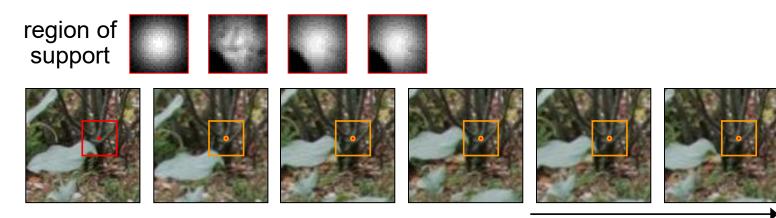
• SSD patch matching search

Confused by occlusion !



time

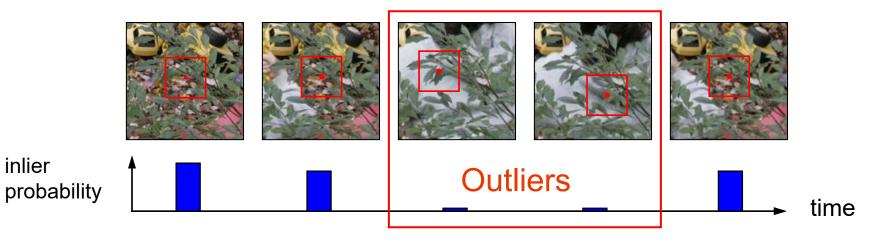
• Learn adaptive region of support using expectationmaximization (EM) algorithm



time

Feature tracking trick 2: trajectory pruning

- Tracking with adaptive region of support
- Nonsense at full occlusion!



• Outlier detection (by checking SSD error and distance to other parts of trajectory), and replacement by interpolation



Comparison

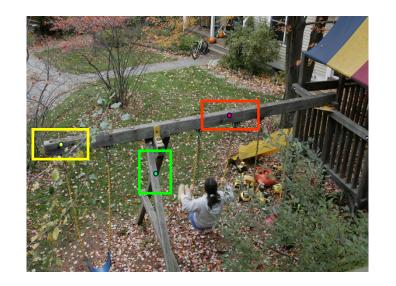


Without adaptive region of support and trajectory pruning

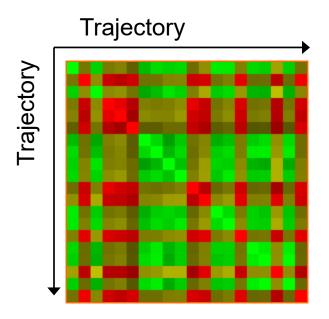
Cluster trajectories based on normalized complex correlation

- The similarity metric should be independent of phase and magnitude
- Normalized complex correlation

 $S(C_1, C_2) = \frac{\left|\sum_t C_1(t)\overline{C}_2(t)\right|^2}{\sqrt{\sum_t C_1(t)\overline{C}_1(t)}\sqrt{\sum_t C_2(t)\overline{C}_2(t)}}$



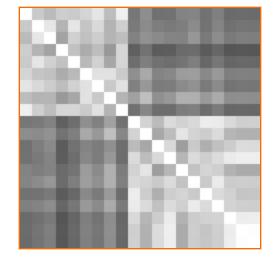
Spectral Clustering



Affinity matrix

Two clusters

Clustering



Reordering of affinity matrix

Clustering Results



From Sparse Feature Points to Dense Optical Flow Field

Interpolate dense optical flow field using locally weighted linear regression

> Demse explicabiliow fieldtofedusperse (swing) points

Cluster 1: leaves Cluster 2: swing

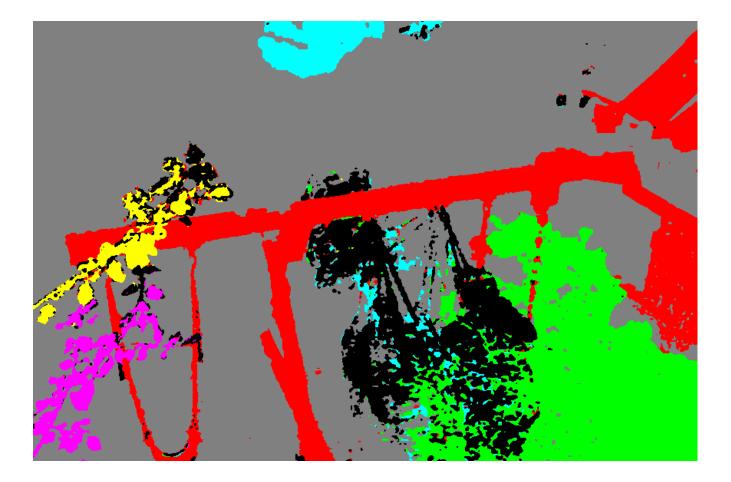


Motion Layer Assignment

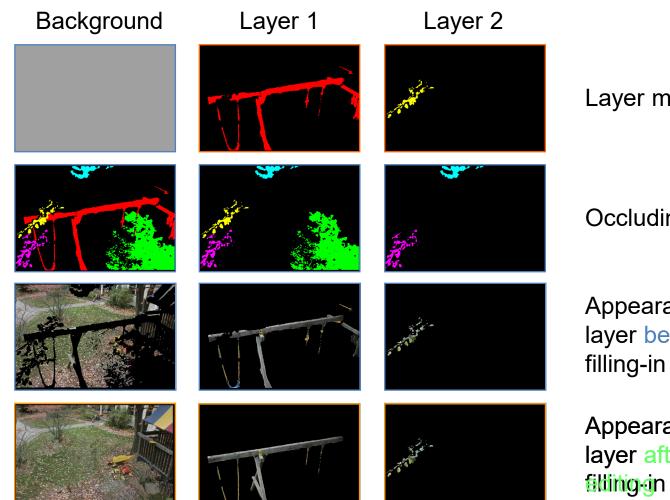
- Assign each pixel to a motion cluster layer, using four cues:
 - Motion likelihood—consistency of pixel's intensity if it moves with the motion of a given layer (dense optical flow field)
 - Color likelihood—consistency of the color in a layer
 - **Spatial connectivity**—adjacent pixels favored to belong the same group
 - Temporal coherence—label assignment stays constant over time
- Energy minimization using graph cuts

Segmentation Results

Two additional layers: static **background** and **outlier**



Layered Motion Representation for Motion Processing



Layer mask

Occluding layers

Appearance for each layer before texture

Appearance for each layer after texeture





Discussion of point tracking approach

• Good: applies to any motion

 Bad: requires accurate point tracking, clustering and texture synthesis, so likely to fail

Approach 2: pixelwise processing

Eulerian Video Magnification for Revealing Subtle Changes in the World

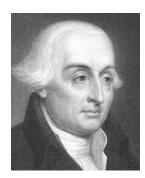
Hao-Yu Wu, Michael Rubinstein, Eugene Shih, John Guttag, Fredo Durand, William T. Freeman ACM Transactions on Graphics, Volume 31, Number 4 (Proc. SIGGRAPH) 2012

Phase-based Video Motion Processing

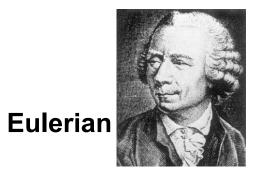
Neal Wadhwa, Michael Rubinstein, Fredo Durand, William T. Freeman ACM Transactions on Graphics, Volume 32, Number 4 (Proc. SIGGRAPH) 2013

> Following slides based on Siggraph presentations: http://people.csail.mit.edu/mrub/vidmag/ http://people.csail.mit.edu/nwadhwa/phase-video/

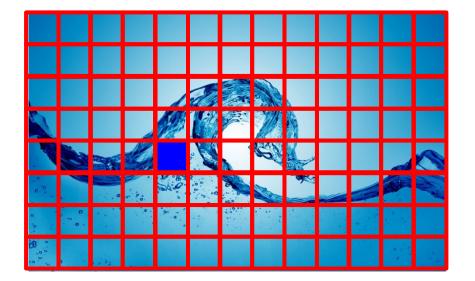
Lagrangian and Eulerian Perspectives: Fluid Dynamics



Lagrangian

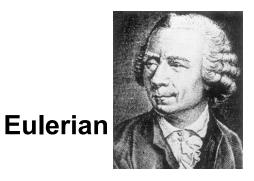


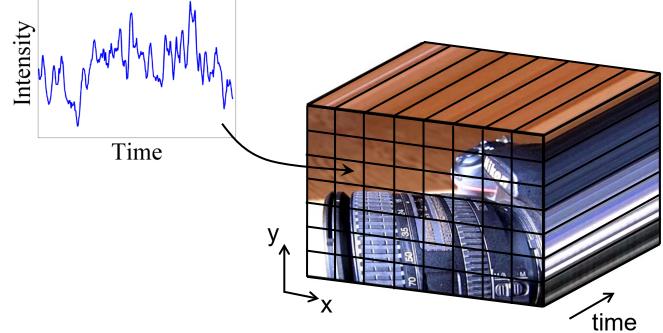




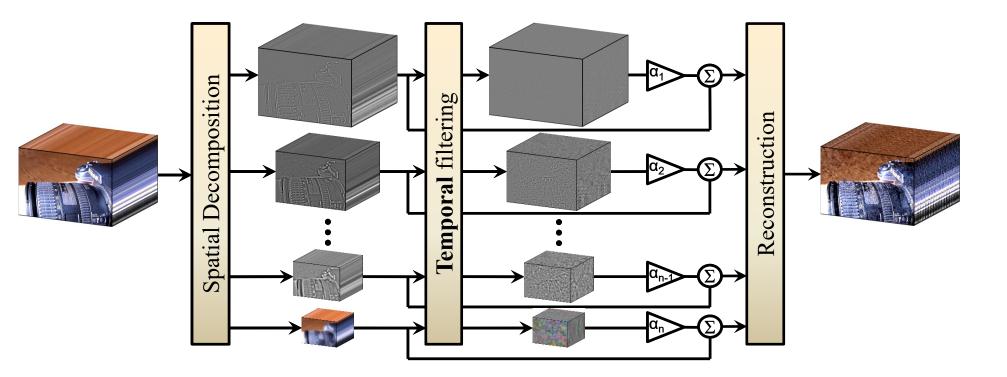
Eulerian Perspective: Videos

- Each pixel is processed independently
- Treat each pixel as a time series and apply signal processing to it





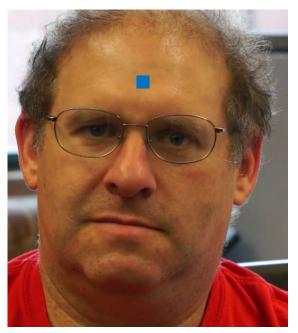
Method Overview



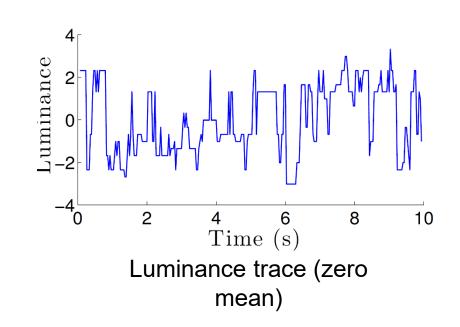
Laplacian Pyramid Bandpass filter intensity at each pixel over time Amplify bandpassed signal and add back to original

Subtle Color Variations

- The face gets slightly redder when blood flows
- Unfortunately usually below the per pixel noise level

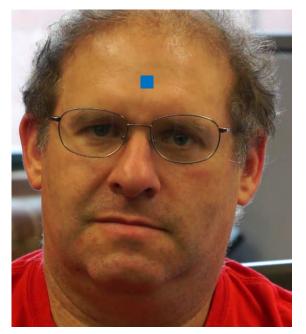


Input frame

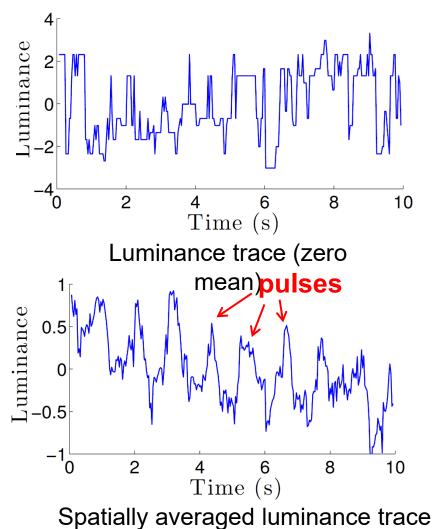


Subtle Color Variations

1. Average spatially to overcome sensor and quantization noise

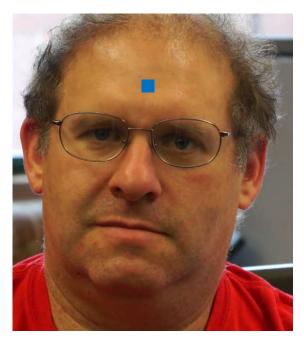


Input frame

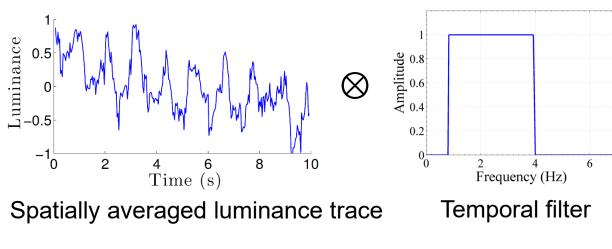


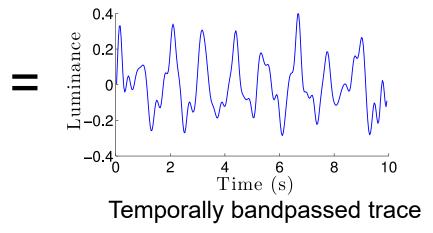
Amplifying Subtle Color Variations

2. Filter temporally to extract the signal of interest



Input frame





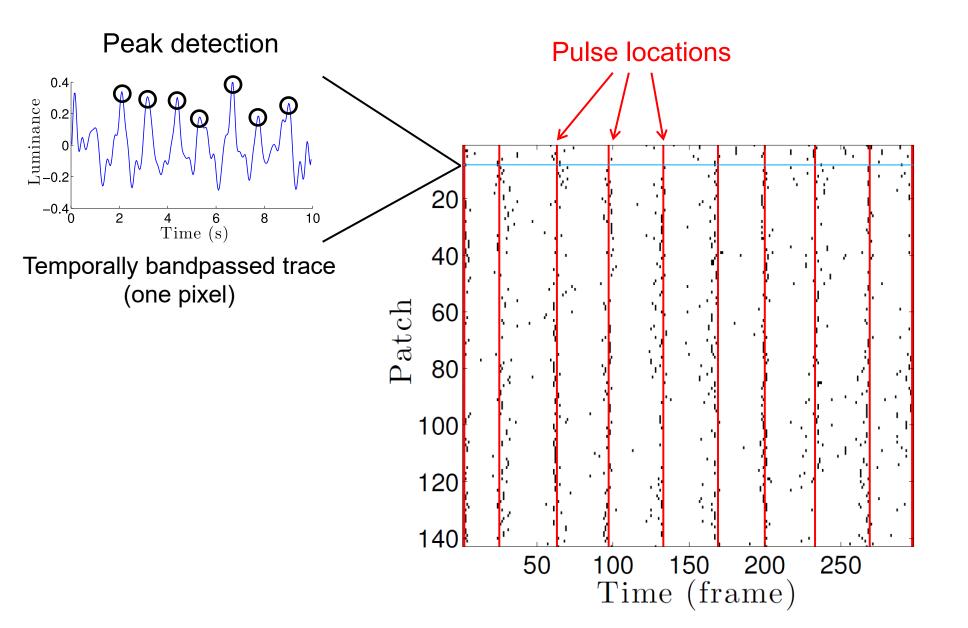
Color Amplification Results



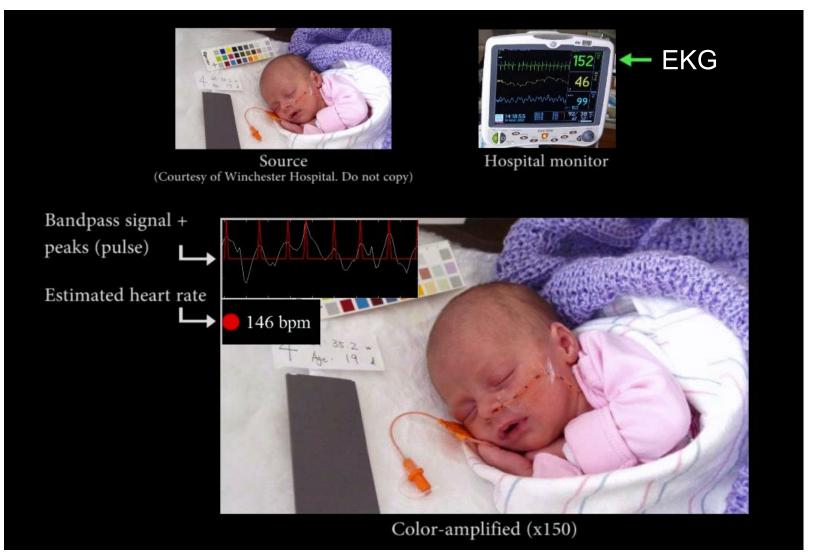
Color-amplified (x100) 0.83-1 Hz (50-60 bpm)

Source

Heart Rate Extraction

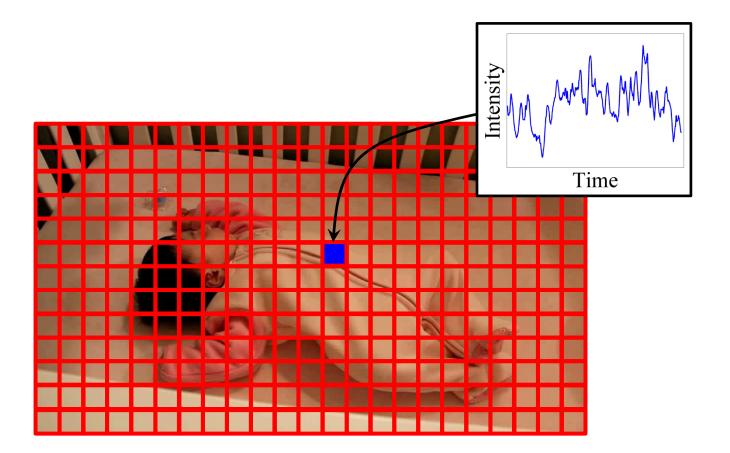


Heart Rate Extraction

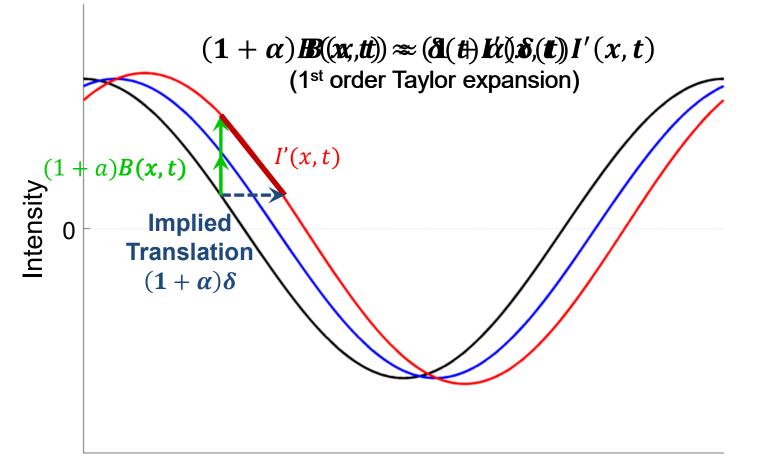


Thanks to Dr. Donna Brezinski and the Winchester Hospital staff 2.33-2.67 Hz (140-160 bpm)

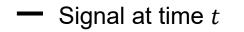
Why It Amplifies Motion



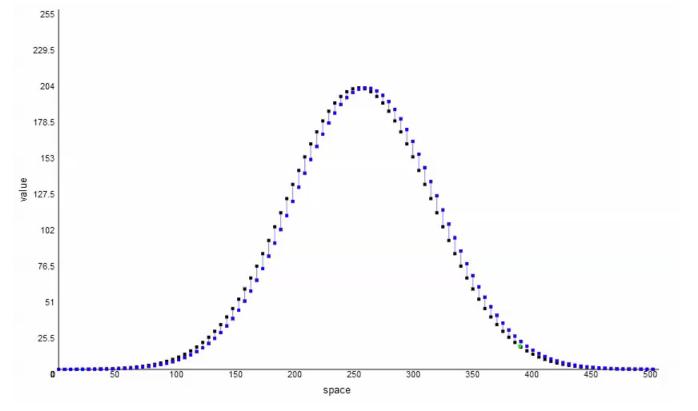
Relating Temporal and Spatial Changes



Relating Temporal and Spatial Changes

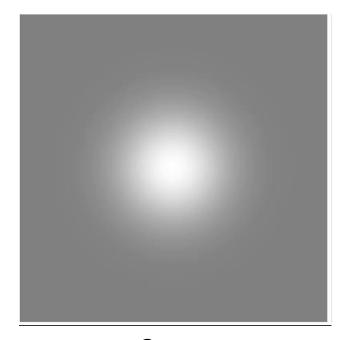


- Signal at time t + 1
- Motion-magnified



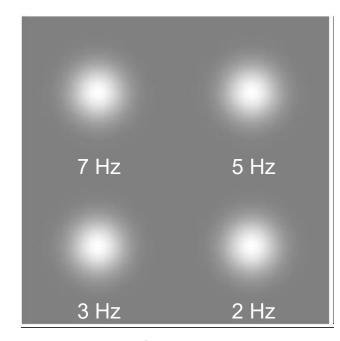
Courtesy of Lili Sun

Synthetic 2D Example

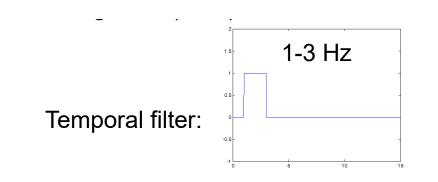


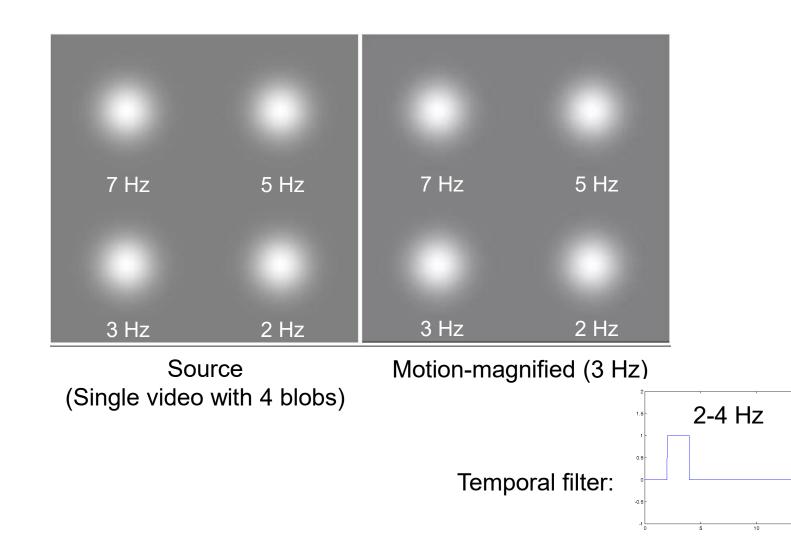
Source

-

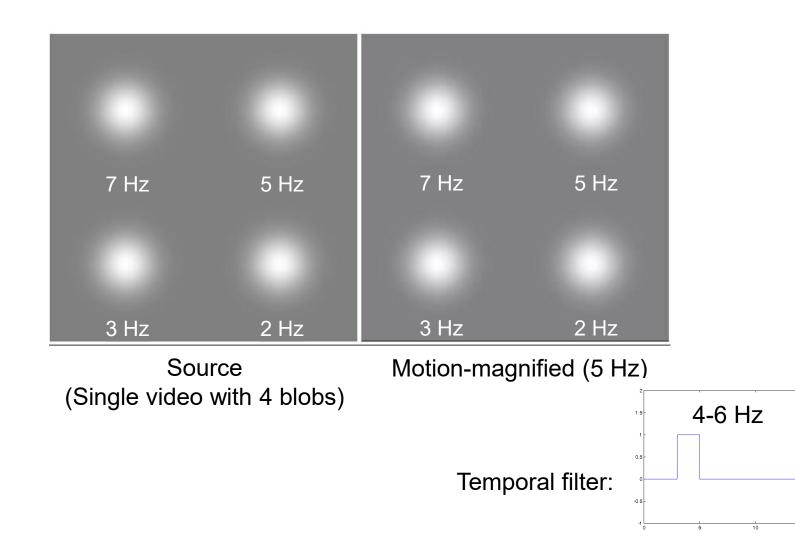


Source (Single video with 4 blobs)

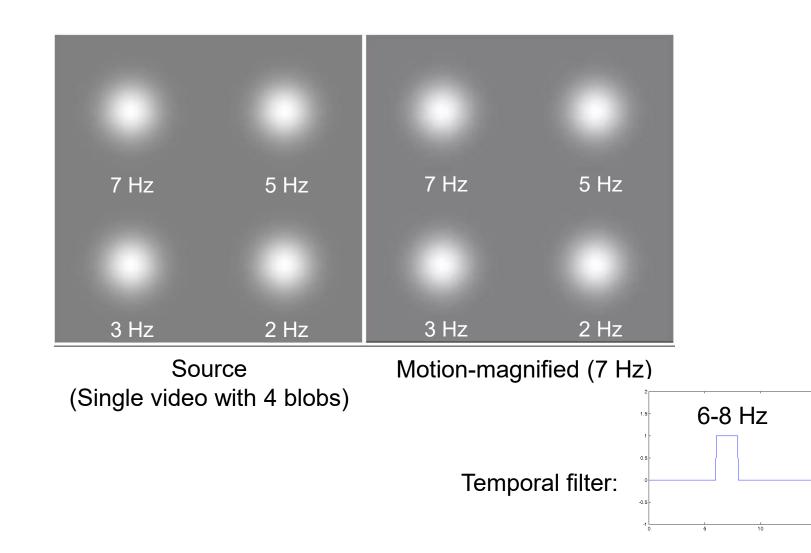




15

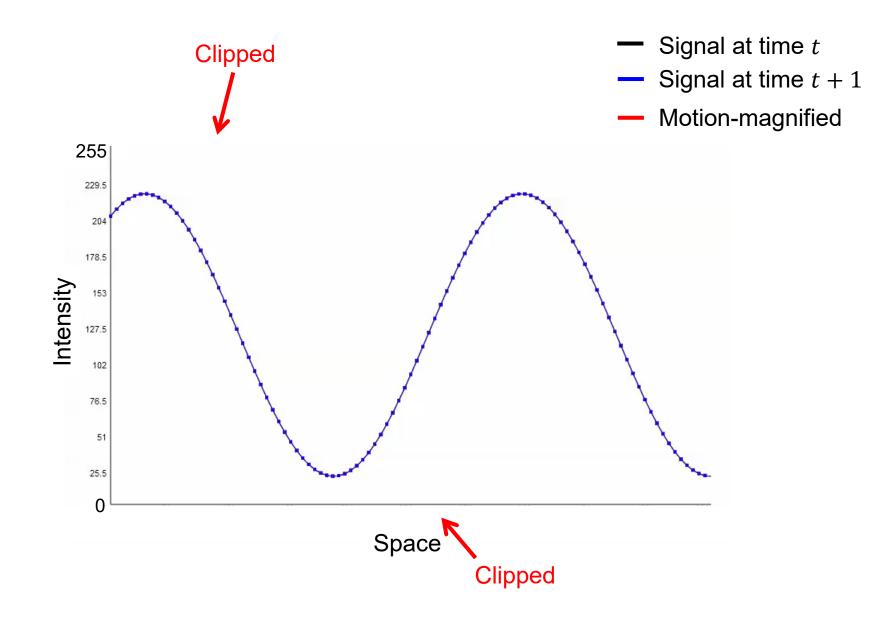


15



15

When Does It Break?



Motion Magnification Artifacts

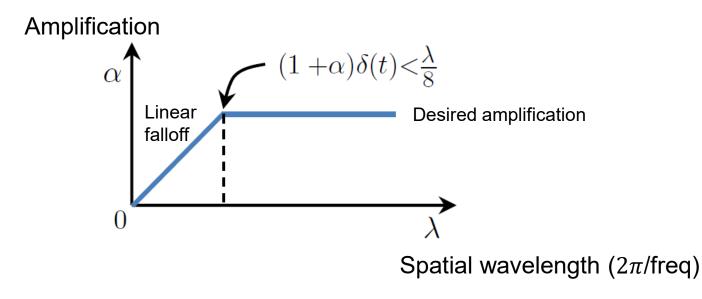


Motion-magnified (3.6-6.2 Hz, x6)

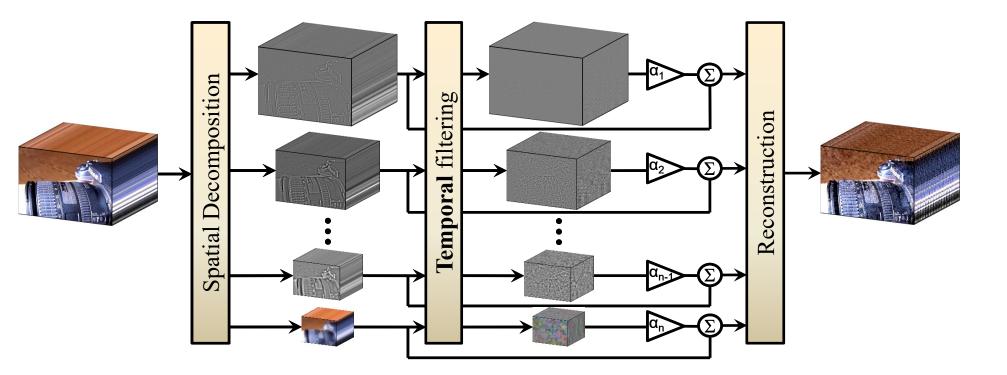
Artifact

Scale-varying Amplification

- The amplification is more accurate for low spatial frequencies
 - Images are smoother
 - Motions are smaller
- Use the desired α for lower spatial frequencies, and attenuate for the higher spatial frequencies



Method Recap



Laplacian Pyramid Bandpass filter intensity at each pixel over time Amplify bandpassed signal and add back to original

Motion Magnification Results



Source

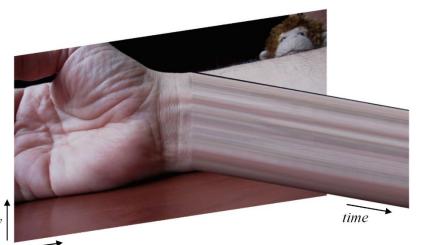
Motion-magnified (0.4-3 Hz, x10)

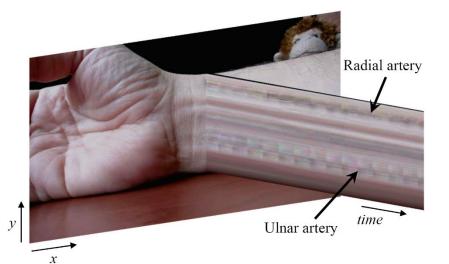
Motion Magnification



Source

Motion-magnified (0.4-3 Hz, x10)





J

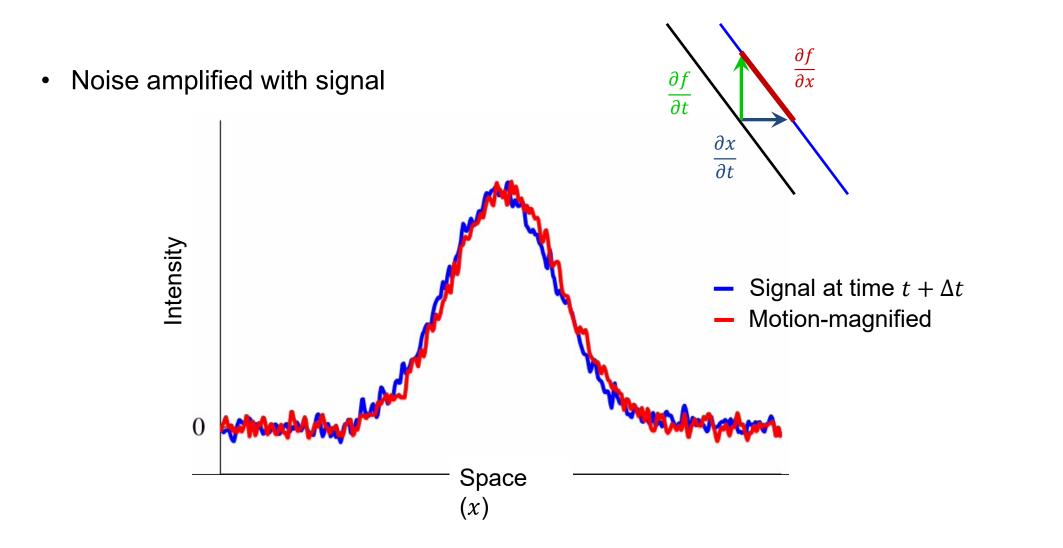
Cool illusion!

https://www.youtube.com/watch?v=x-mHnNDHDDE

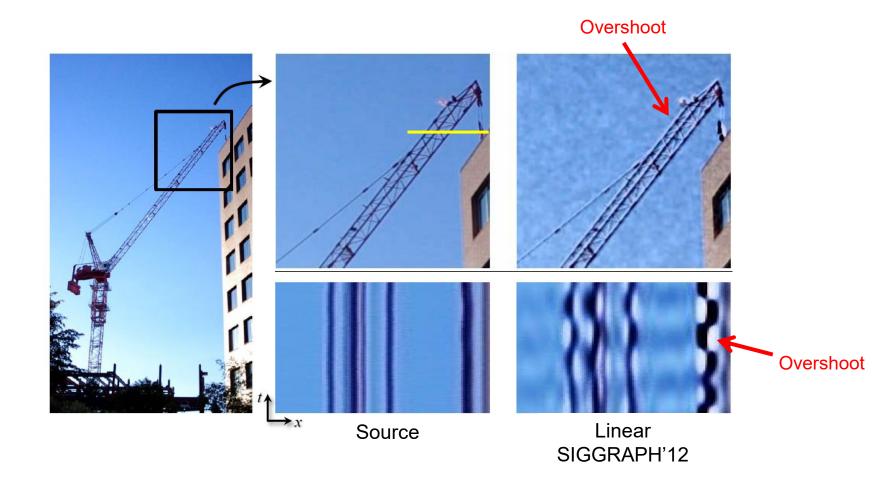
Discussion of pixelwise intensity amplification approach

- Good:
 - Does not require explicit motion estimation or texture synthesis (robust)
 - Very fast (real time)
- Bad:
 - Can only handle very small motions
 - Amplifies noise

Limitations of Linear Motion Processing



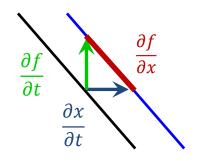
Limitations of Linear Motion Processing



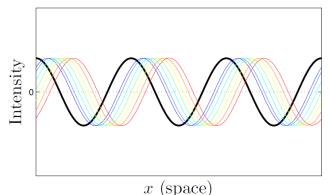
Eulerian approach part 2: shift phase instead of amplifying intensity

Translation in space is equivalent to a shift in phase

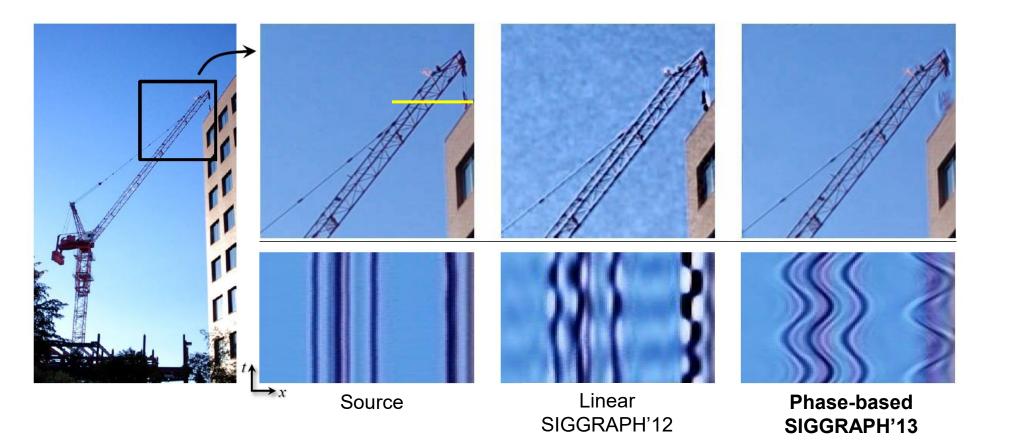
- Linear Motion Processing
 - Assumes images are locally linear
 - Translate by **changing intensities**



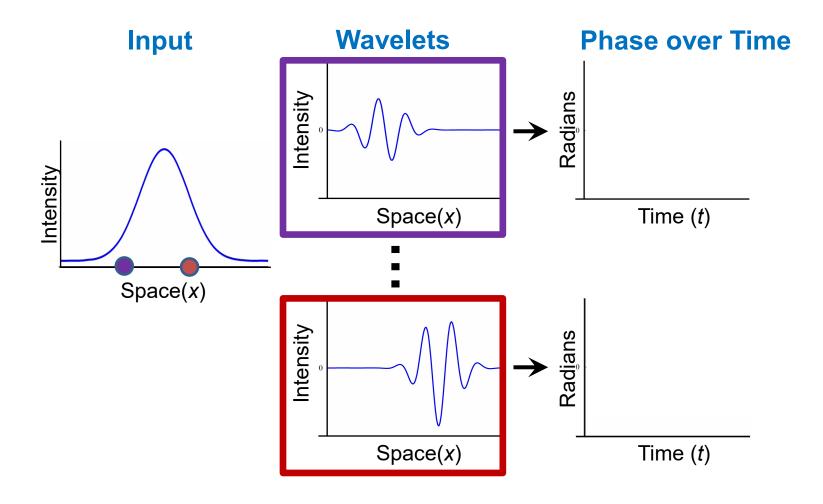
- Phase-Based Motion Processing
 - Represents images as collection of local sinusoids
 - Translate by **shifting phase**



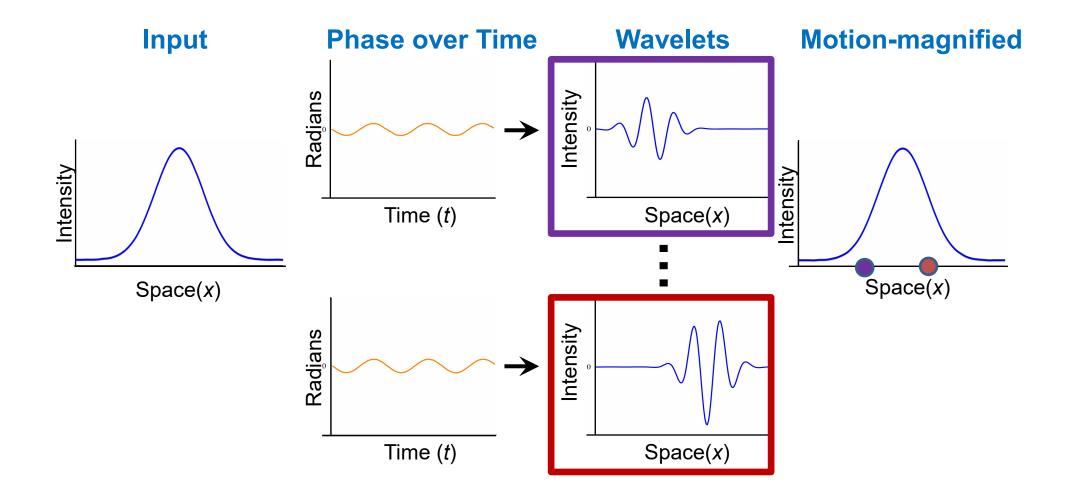
Linear vs. Phase-Based Motion Processing



Phase over Time

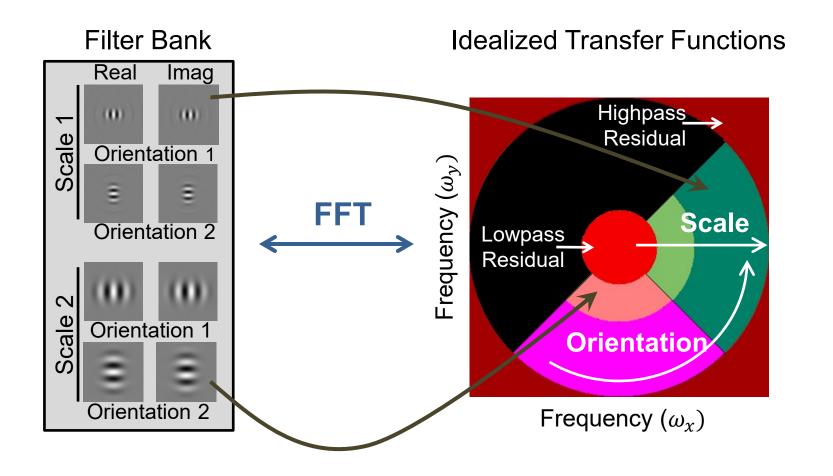


Phase over Time

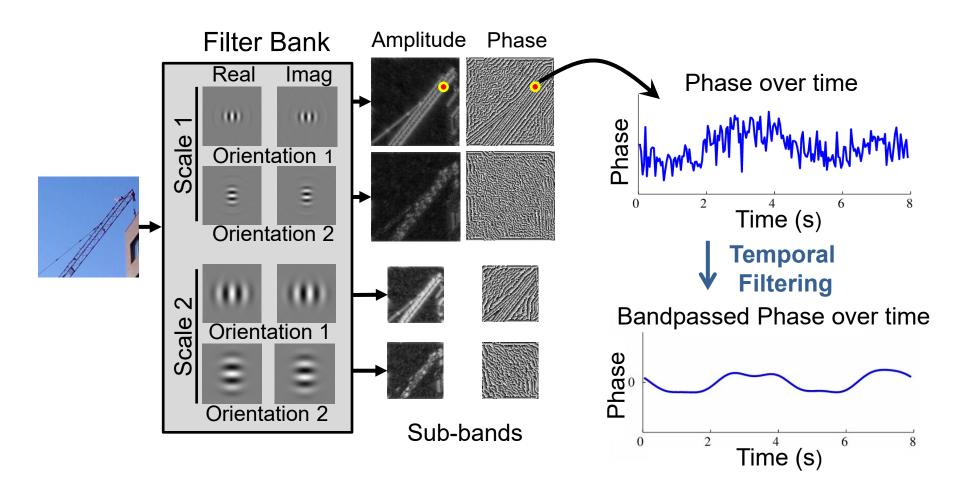


2D Complex Steerable Pyramid

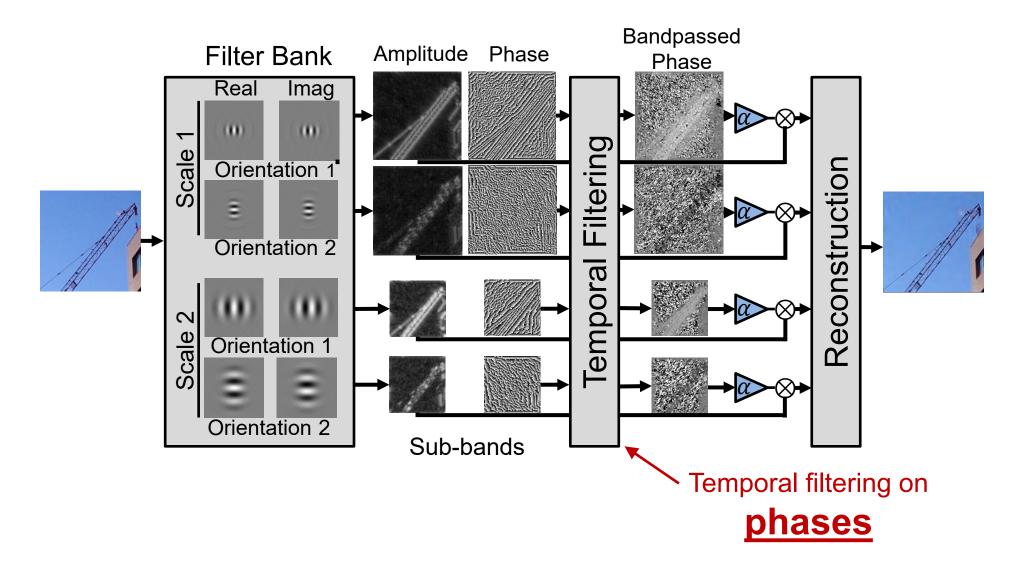
[Simoncelli et al. 1992]



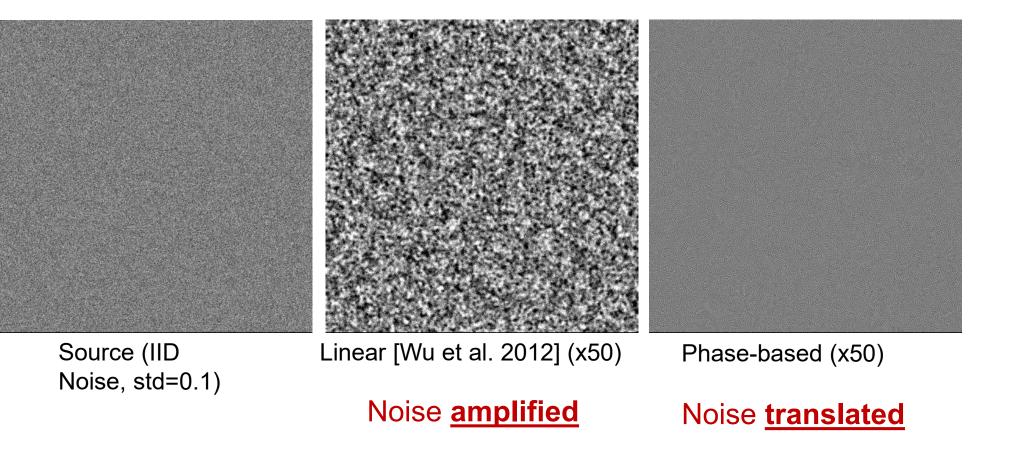
Phase over Time



New Phase-Based Pipeline

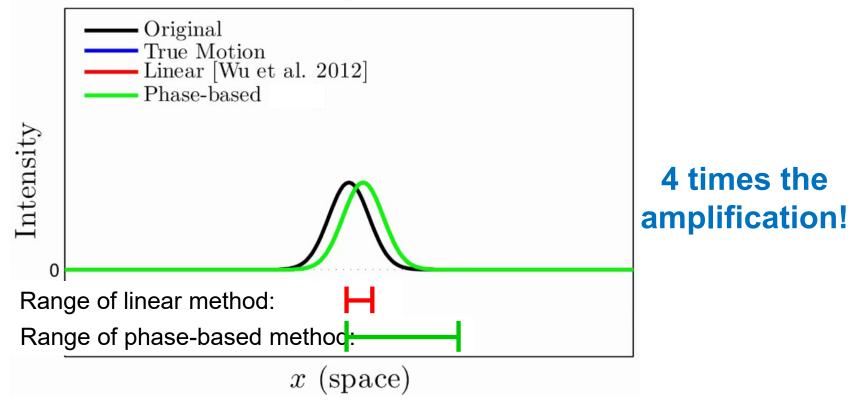


Improvement #1: Less Noise



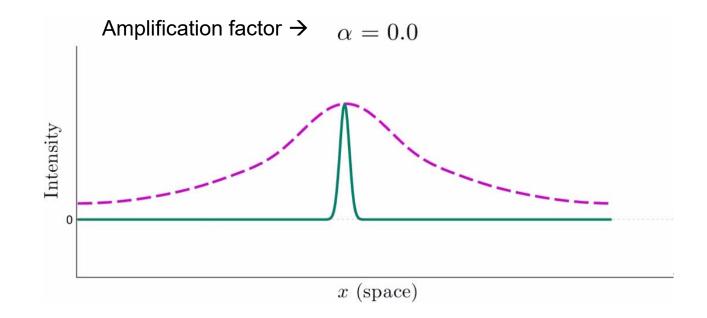
Improvement #2: More Amplification

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



Limits of Phase Based Magnification

• Local phase can move image features, but only within the filter window



Comparison with [Wu et al. 2012]



Wu et al. 2012

Eye Movements



Source (500FPS)

Expressions



Source



Low frequency motions

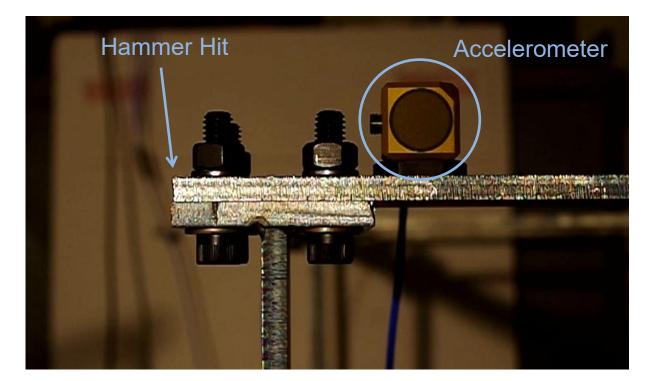


Mid-range frequency motions

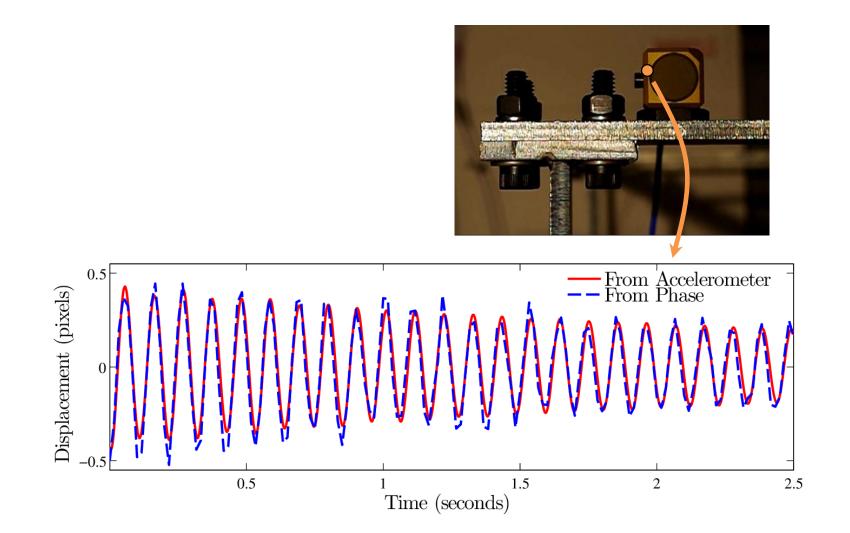
Ground Truth Validation

 Induce motion (with hammer)

• Record with accelerometer



Ground Truth Validation

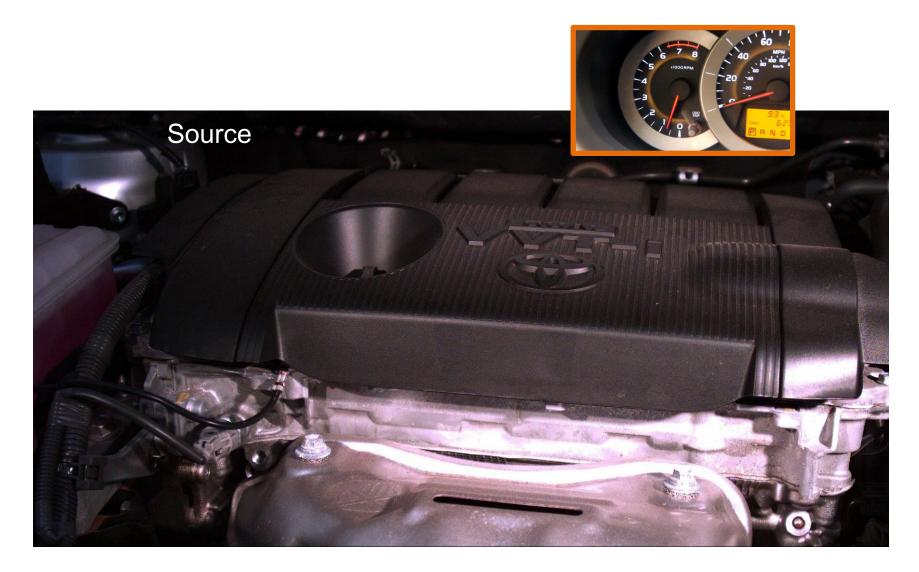


Motion Attenuation

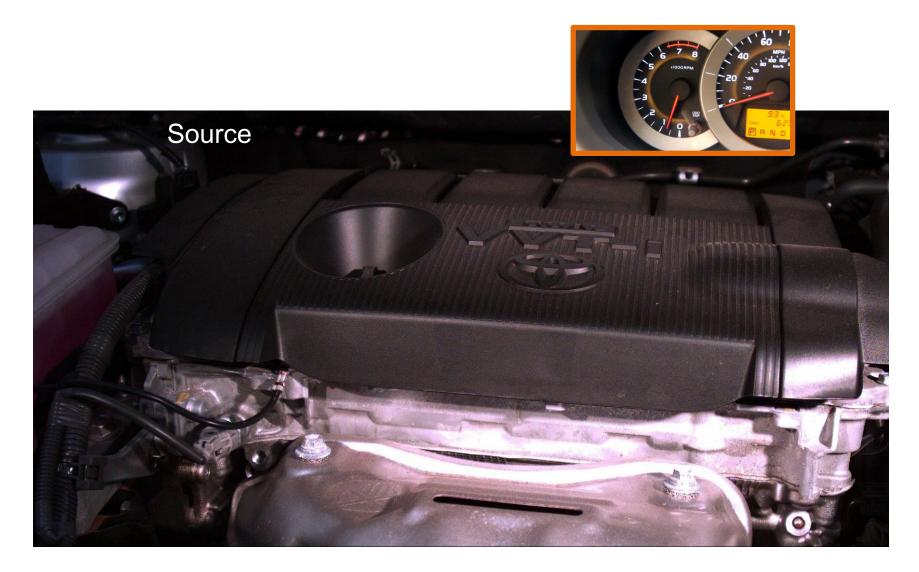


Source

Sequence courtesy Vimeo user Vincent Laforet

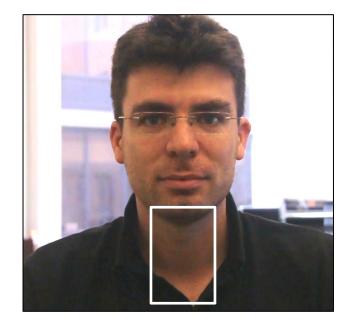


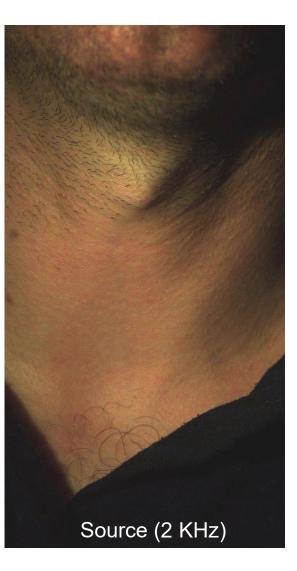


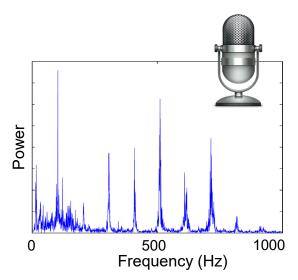


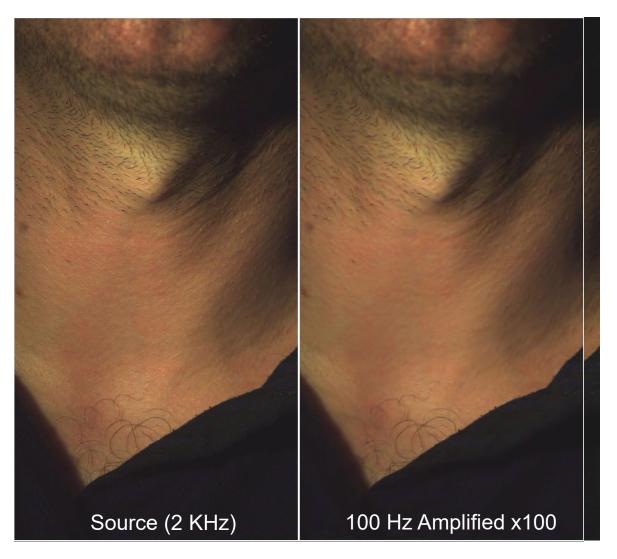


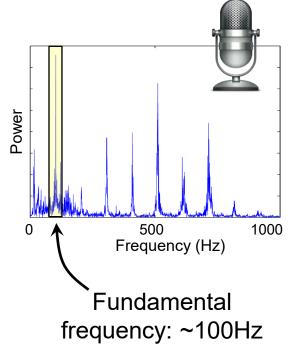
Neck Skin Vibrations

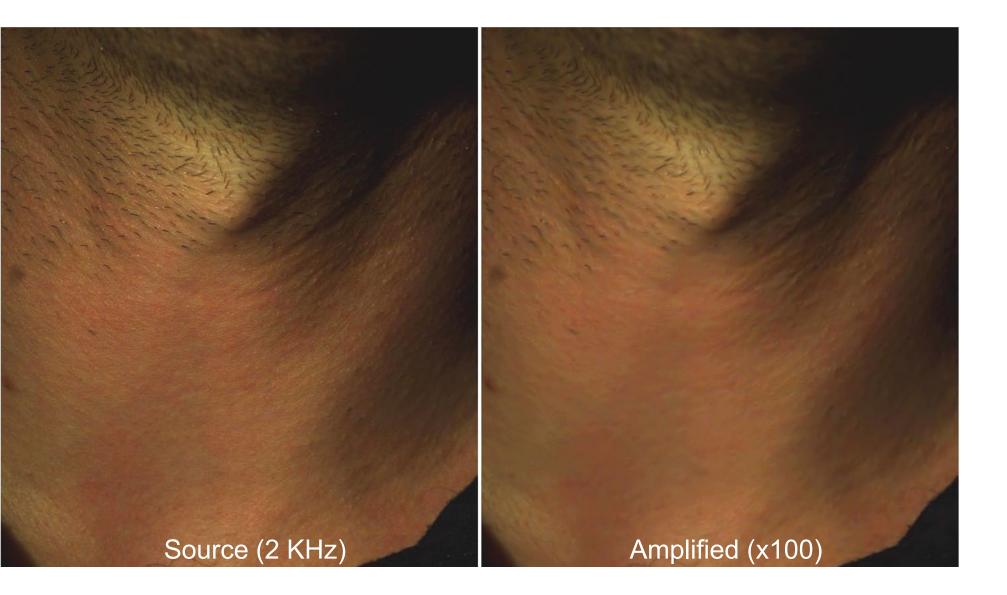








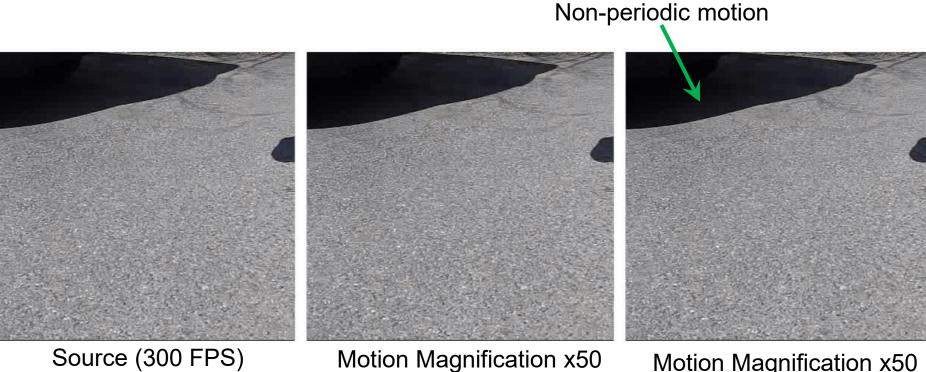




Discussion of pixelwise phase magnification approach

- Good:
 - Does not require explicit motion estimation
 - Produces more direct translations (instead of perceived motion)
 - Does not amplify noise
- Bad:
 - Limited in range of amplication (compared to pointwise approach)
 - May have difficulty with non-periodic motion and large motions

Non-periodic Motions and Large Motions



Motion Magnification x50 Large Motions Unmagnified



Abe Davis Michael Rubinstein Neal Wadhwa Gautham Mysore Fredo Durand William T. Freeman



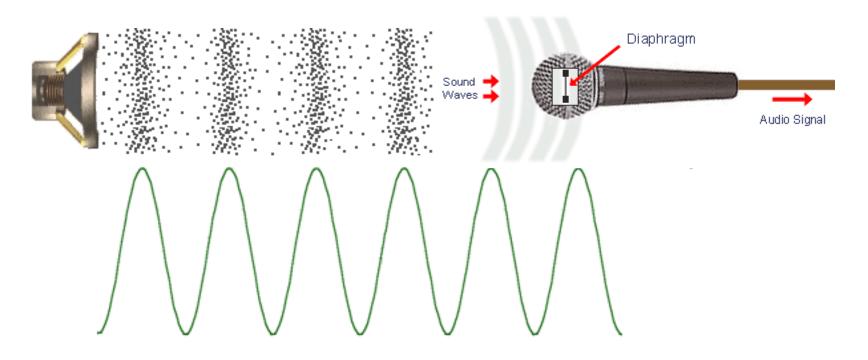


(slides adopted from Siggraph presentation)

Remote Sound Recovery

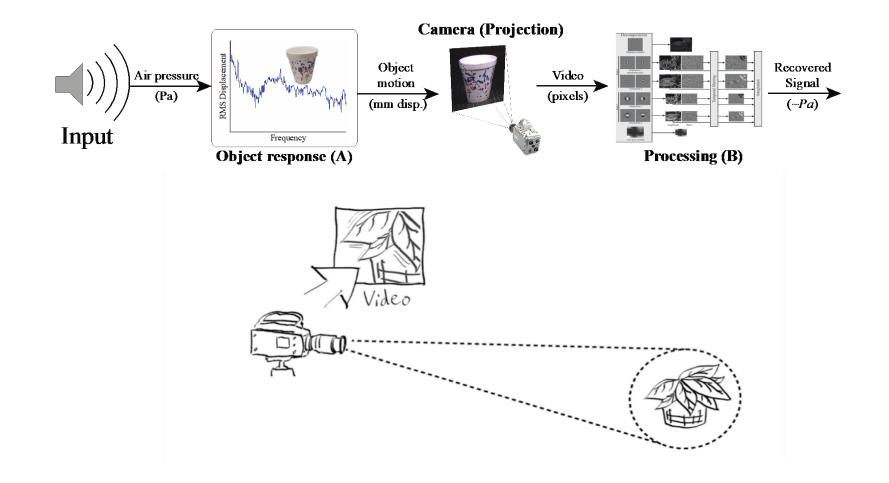


Sound and Motion



Source: mediacollege.com

The Visual Microphone

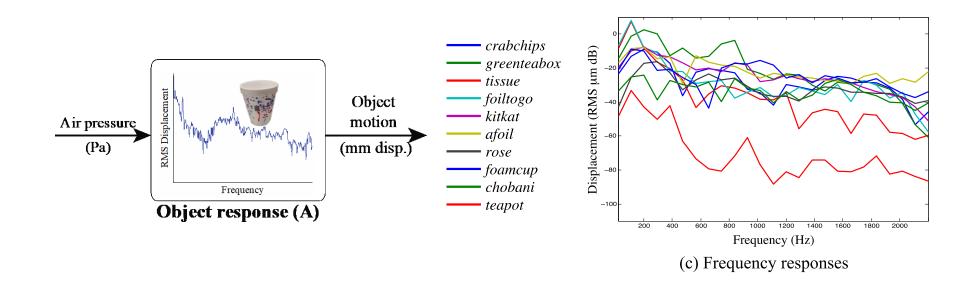


Processing

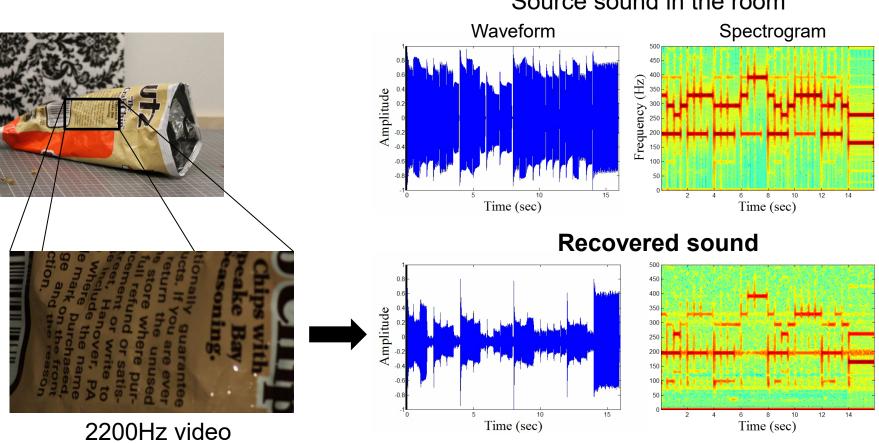
- Extract local motion sigr
- Average and Align
- Post-process



Some materials are better microphones than others

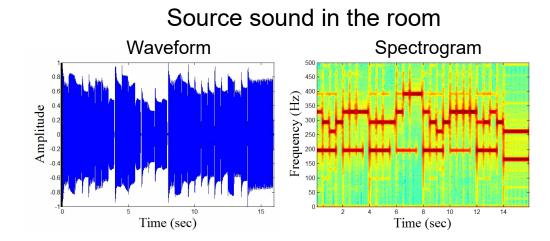


Sound Recovered from Video



Source sound in the room

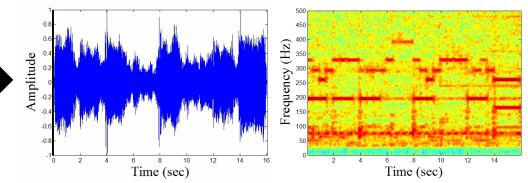
Sound Recovered from Video



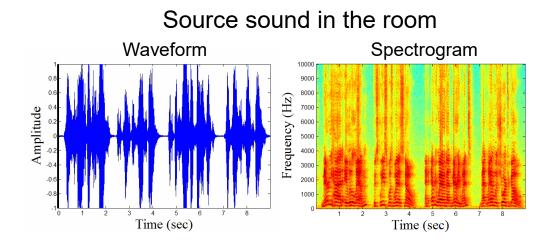
Recovered sound



2200Hz video



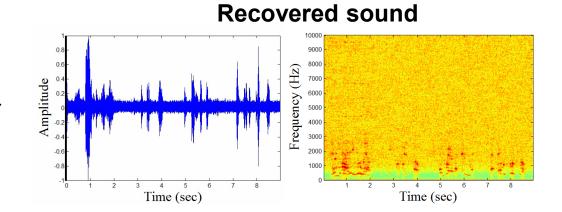
Sound Recovered from Video



(small patch on the chip bag)



20 kHz video











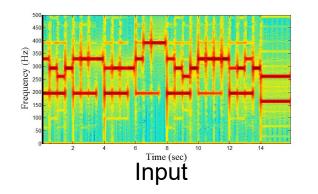
https://www.flickr.com/photos/sorenragsdale/3904937619/ http://www.flickr.com/photos/boo66/5730668979/

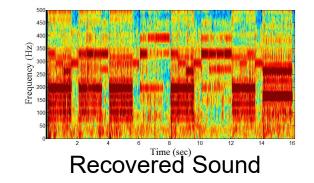




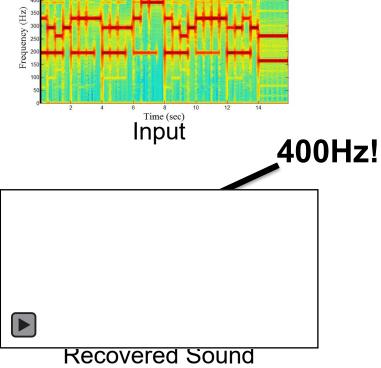


Input video (60 fps)





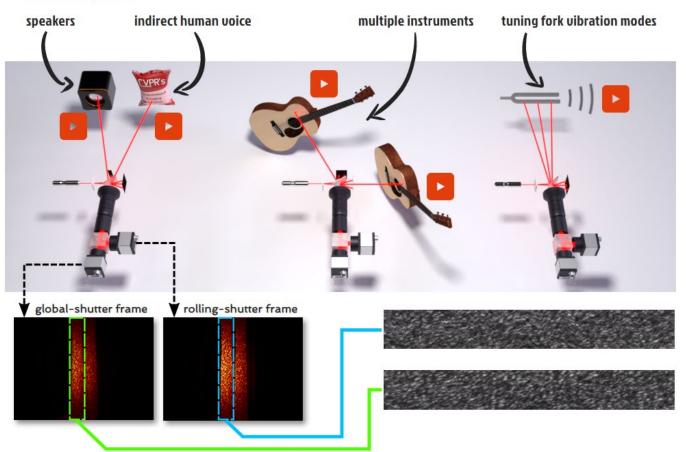




See the latest version (great video): https://imaging.cs.cmu.edu/vibration/

"Seeing" sound in a novel way

We propose a novel dual-shutter approach for sensing surface vibrations at high speeds (up to 63kHz), for multiple scene sources at once, in a bandwidth-efficient way (using "slow" 130FPS cameras). We demonstrate our method by capturing vibration caused by audio sources, such as:



Use laser to create speckle and combo of global and rolling shutter to accurately measure changes

Summary

- Several ways to magnify motion
 - Directly measure and exaggerate point motions
 - Amplify intensity changes after temporal filtering (creating apparent motion)
 - Amplify local phase variations after temporal filtering
- Micro-motion estimates can be used to measure sound

Next week

- Final class
 - A few examples of cutting edge applications
 - Where to learn more
 - Course feedback is online please complete!