

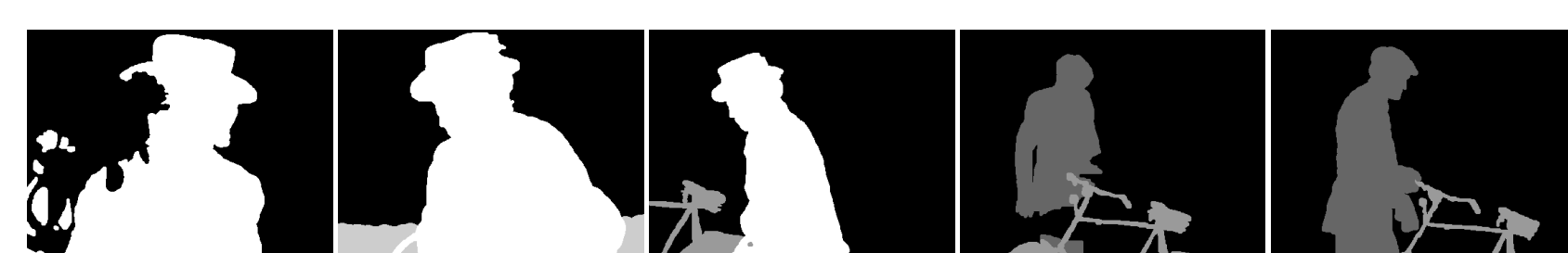
Motivation

- The objective of video segmentation is to separate out objects from a video sequence based on appearance and motion
- Segmentation is required for video editing and video understanding. It is a necessary step in many applications, but computationally intensive
- We are not just parallelizing existing algorithms, but are also developing new ones
- Parallelization has enabled us to move highly accurate computer vision algorithms from images to videos
- We are using a cluster today, but expect it to run on the desktop within 5-10 years

Example

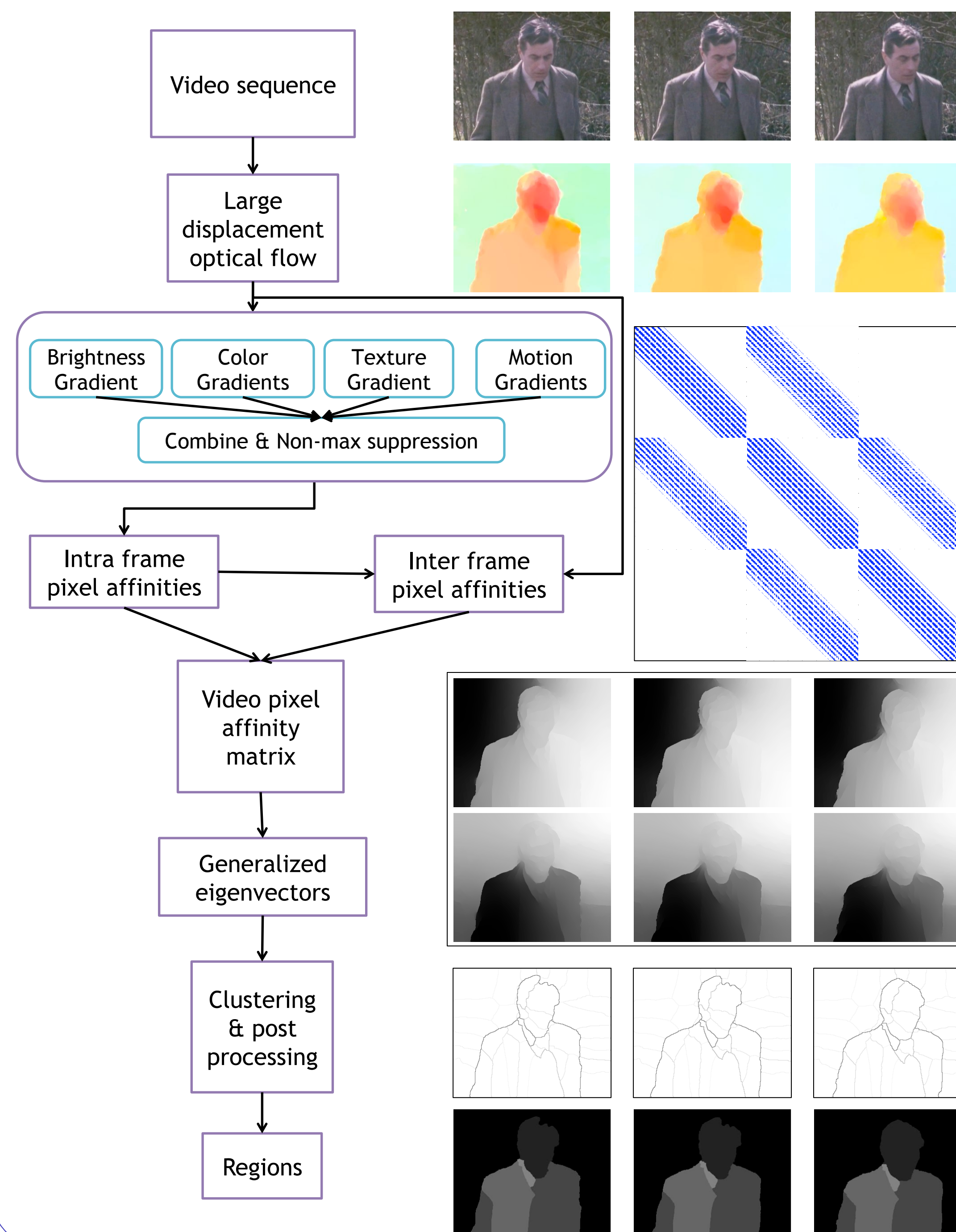


Example video sequence



Ground truth segmentation

Application Overview



Computational Challenges

- Most of the computational challenges are in the eigensolver
- Large memory requirements
 - The video affinity matrix takes up ~20 GB for 100 frames of size 640x480
 - Single node solution is impractical
 - Solution: Distributed computing – Dirac cluster at NERSC
- Computing $A^T x$ is much slower than Ax for matrix blocks
 - $A^T x$ runs about 3x slower than Ax
- Bugs caused due to floating point non-determinism
 - MPI_Reduce, AtomicAdd result in errors due to non-determinism

Algorithmic Challenges

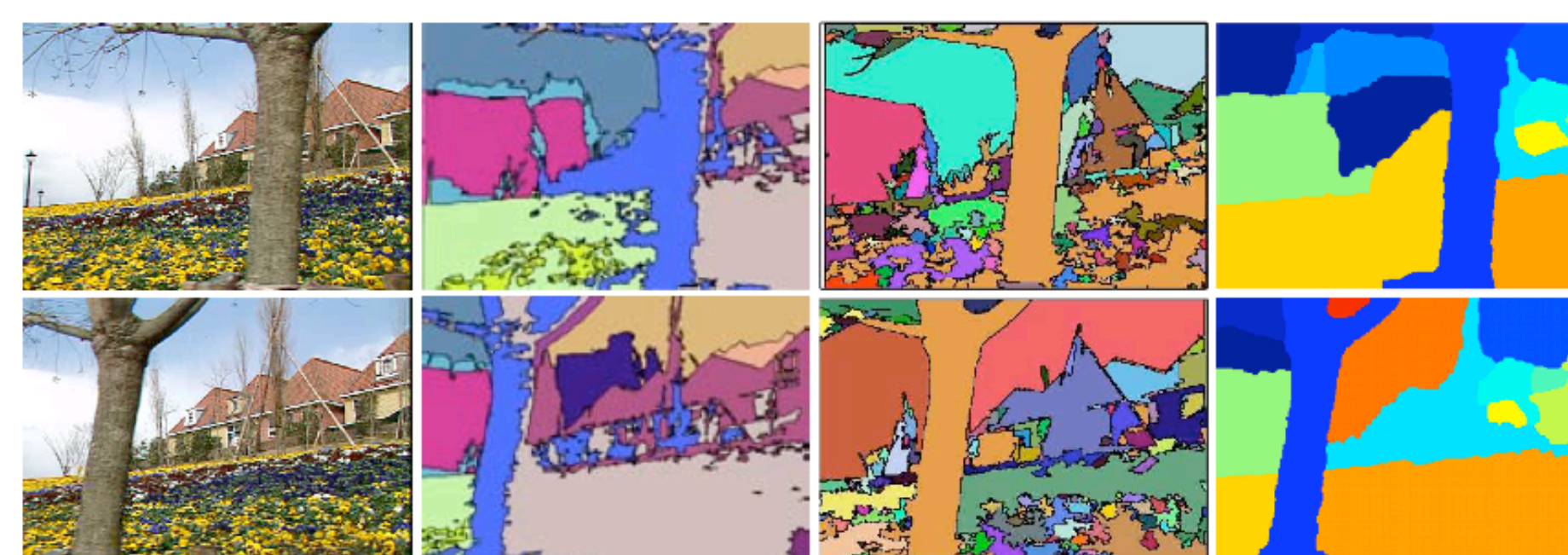
- Defining inter-frame pixel affinities
 - Combine intra-frame affinities and optical flow
- “Leakage” in eigenvectors
 - Solved through k-means clustering and Ultrametric contour maps

Results

Algorithm	Label density	Overall accuracy (% pixels correctly labeled)	Objects extracted with <10% error	Over-segmentation
Brox & Malik - trajectory segmentation [1]	3.43%	3.17%	7	3.14
ALC with incomplete tracks [2]	3.43%	2.77%	0	54.57
Hierarchical Graph Segmentation [3]	100%	79.23%	0	10.42
Our technique - Pixel level spectral clustering	100%	84.54%	3	5.86

[1] T. Brox and J. Malik. Object segmentation by long term analysis of point trajectories. In ECCV, 2010.
 [2] S. Rao, R. Tron, R. Vidal, and Y. Ma. Motion segmentation via robust subspace separation in the presence of outlying, incomplete or corrupted trajectories. In CVPR, 2008.
 [3] M. Grundmann, V. Kwatra, M. Han, and I. Essa. Efficient hierarchical graph-based video segmentation. In CVPR, 2010.
 [4] W. Brendel and S. Todorovic. Video object segmentation by tracking regions. In ICCV, 2009.

- ❑ Spectral clustering at the pixel level is practical and better than existing algorithms
 - ❑ 30x more dense than sparse clustering
 - ❑ 2x less over-segmentation than dense techniques
- ❑ Runtime is ~5 minutes for a 200 frame sequence (1.5 sec/frame) on a 34-node GPU+multicore CPU cluster at NERSC



Original sequence Results from [3] Results from [4] Our results

Results on marple1 sequence

