Deep Video Deblurring for Hand-held Cameras

Supplementary Material

Outline

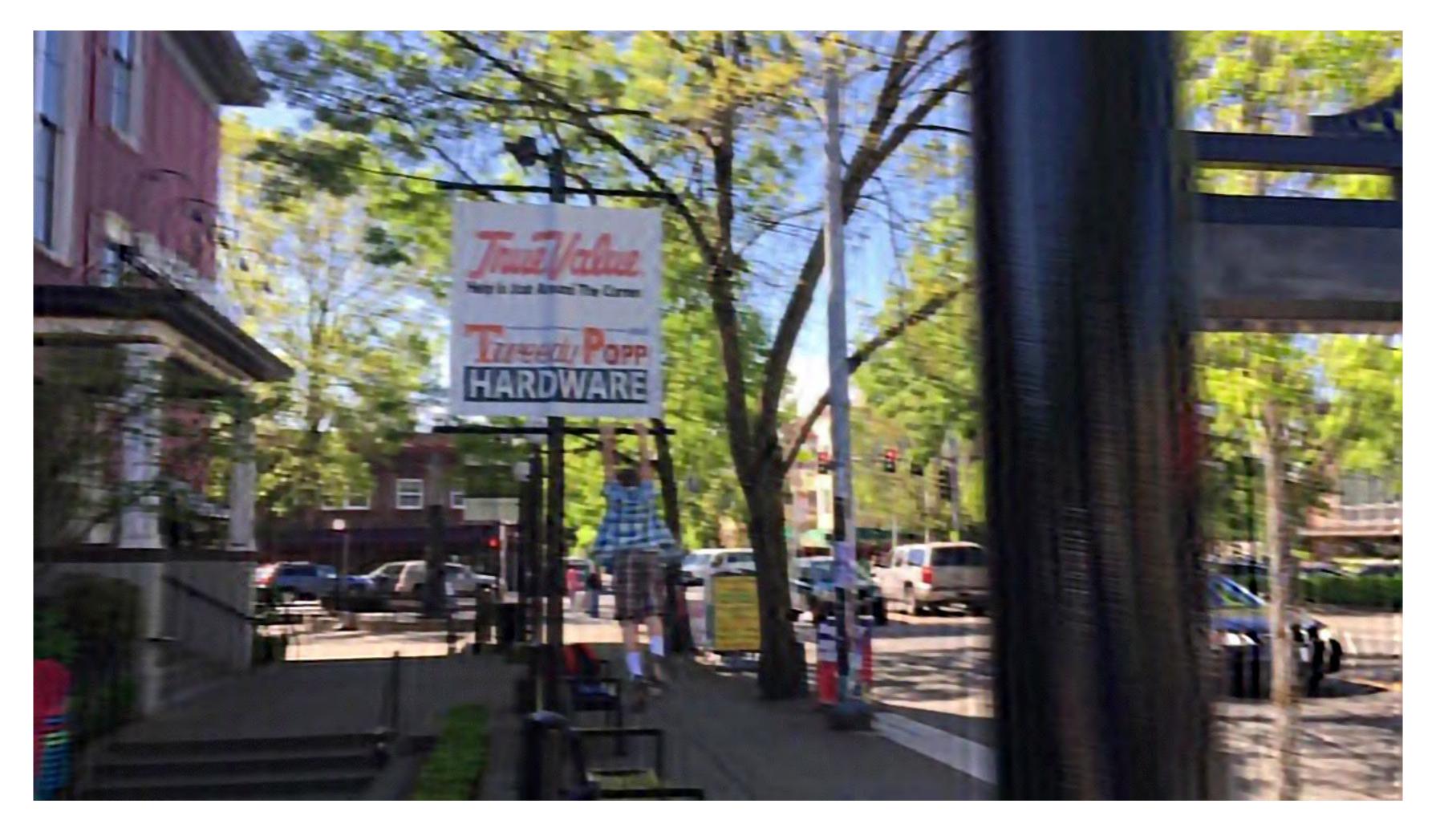
- Full-size images in fig. 4
- Full-size images in fig. 6
- Additional comparisons with Cho [3], Kim and Lee [15], and WFA [5]
- Additional results
- Analysis on the DeBlurNet

- "Cho [3]": S. Cho, J. Wang, and S. Lee. Video deblurring for hand-held cameras using patch-based synthesis. *ACM Transactions on Graphics (TOG)*, 31(4):64, 2012.
- "Kim and Lee [15]": T. H. Kim and K. M. Lee. Generalized video deblurring for dynamic scenes. In *Proc. IEEE Conf. Comput. Vis. Pattern Recognit. (CVPR)*, 2015.
- "WFA [5]": M. Delbracio and G. Sapiro. Hand-held video deblurring via efficient fourier aggregation. *IEEE Transactions on Computational Imaging*, 1(4):270–283, 2015.
- "LODeblur [48]": L. Xu, S. Zheng, and J. Jia. Unnatural IO sparse representation for natural image deblurring. In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition*, pages 1107–1114, 2013.
- "Neural [1]": A. Chakrabarti. A neural approach to blind motion deblurring. arXiv preprint arXiv:1603.04771, 2016.

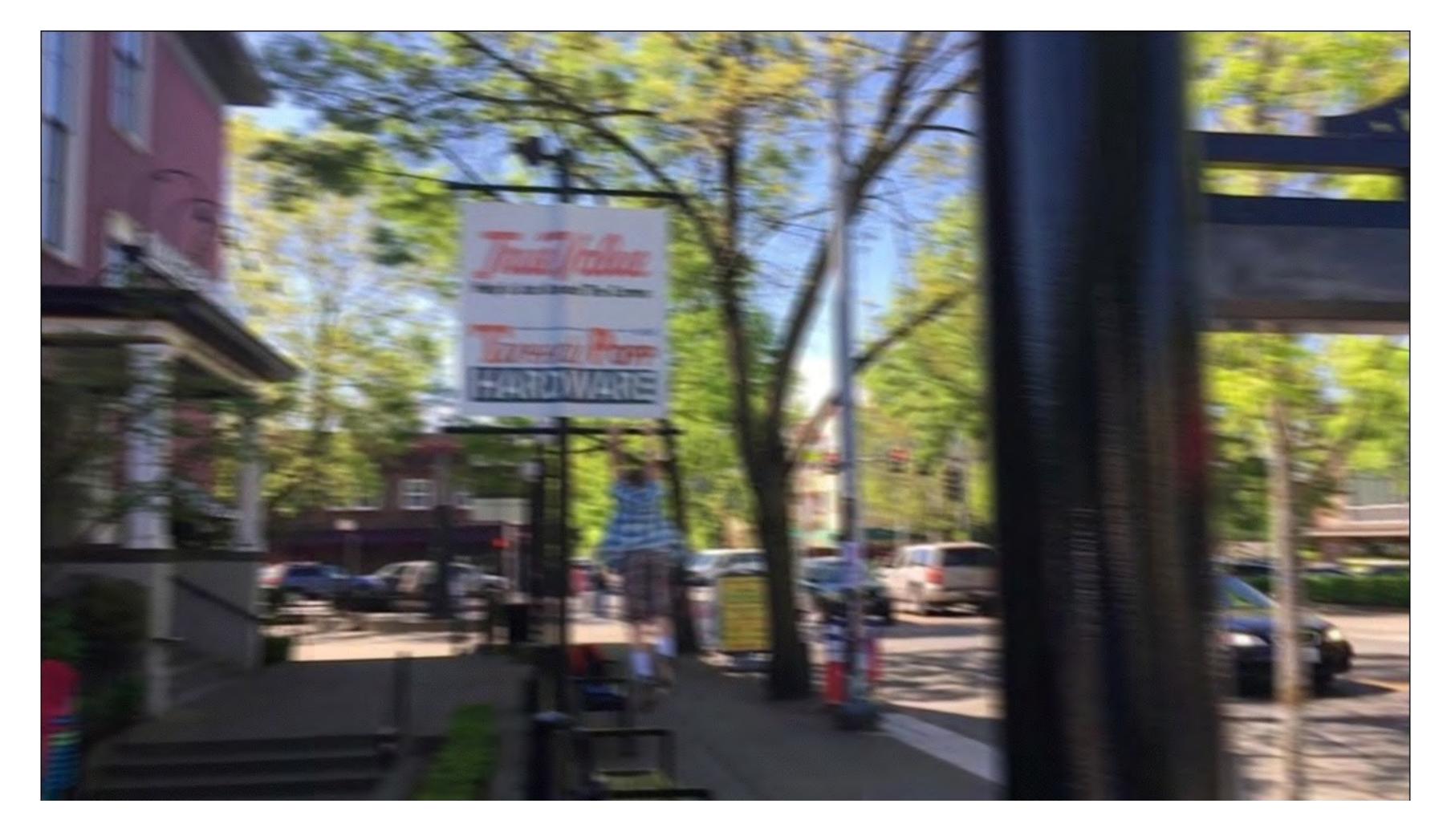
Full-size Images in Fig. 4



input



PSDeblur



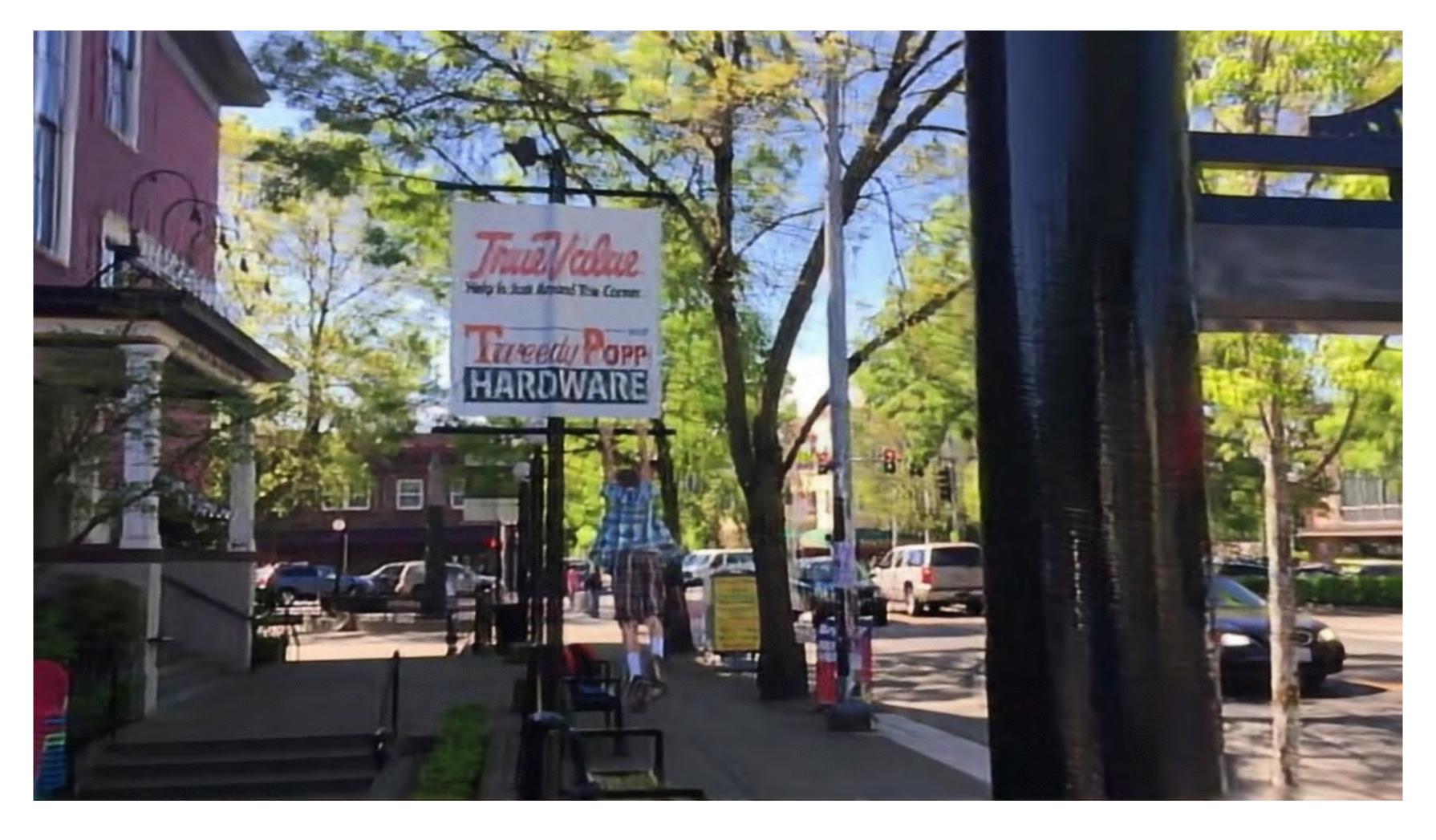
WFA [5]



DBN+Single



DBN+Noalign



DBN+Homog



DBN+Flow



ground-truth



input



PSDeblur



WFA [5]



DBN+Single



DBN+Noalign



DBN+Homog



DBN+Flow

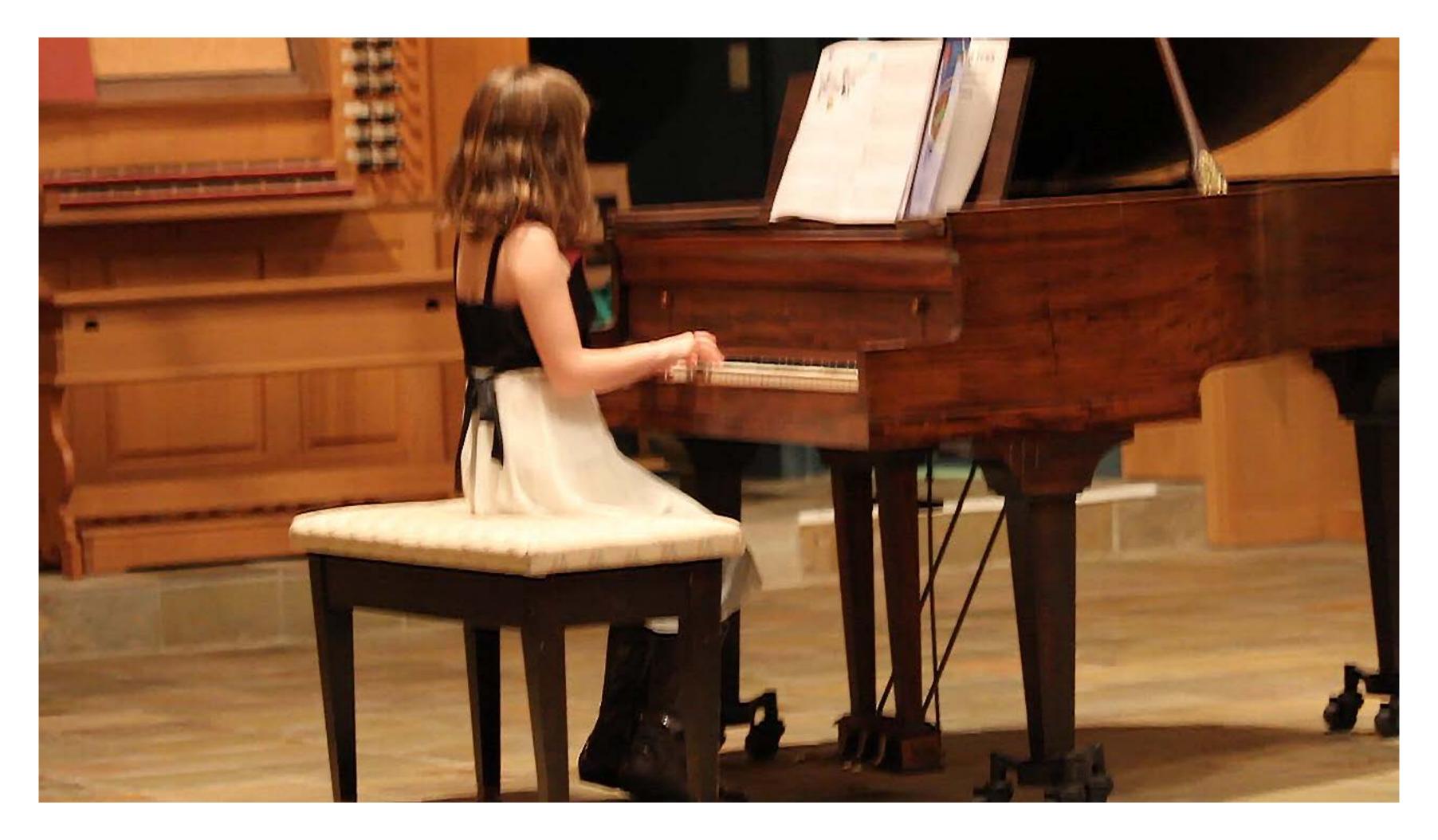


ground-truth

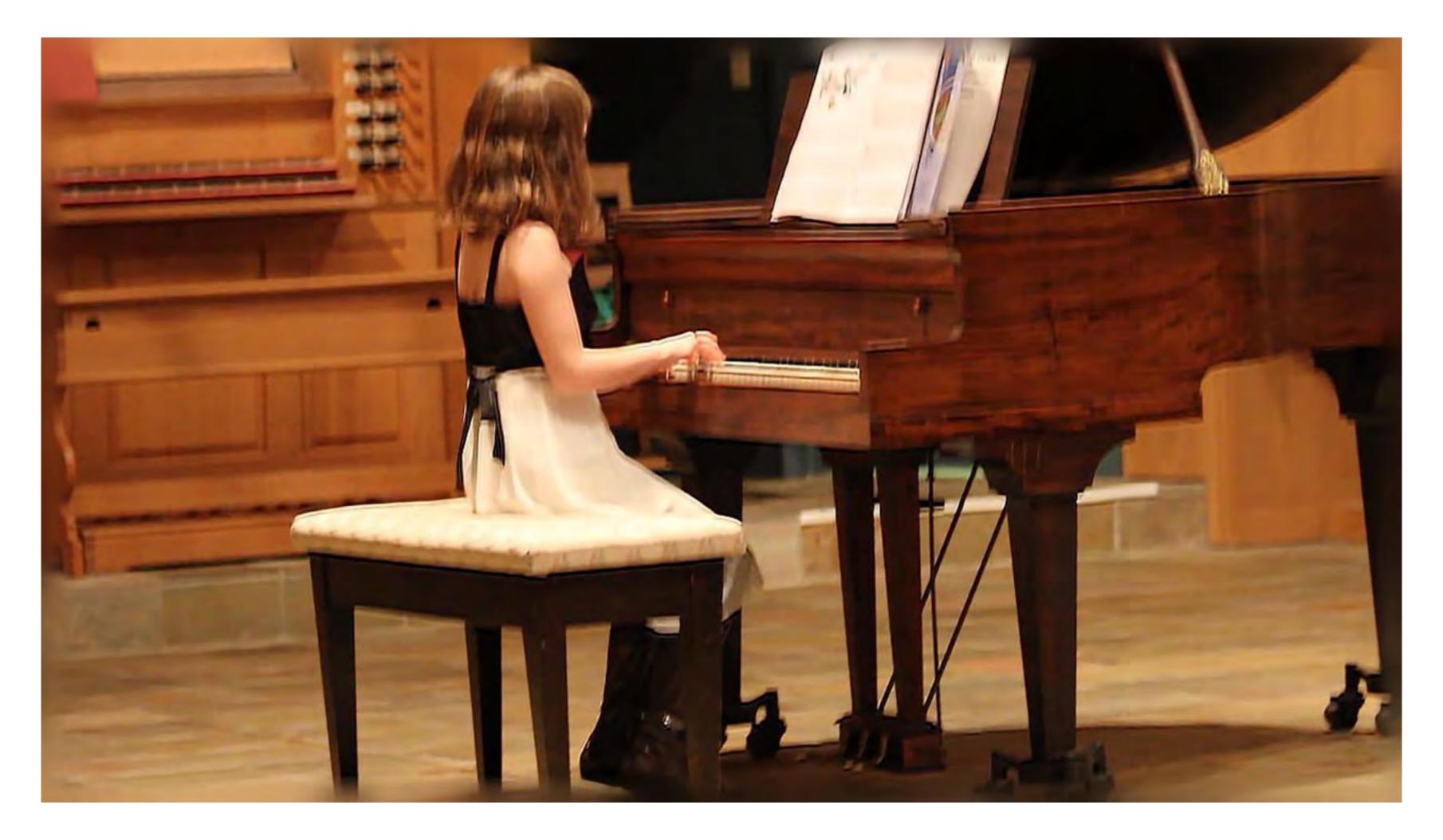
Full-size Images in Fig. 6



input



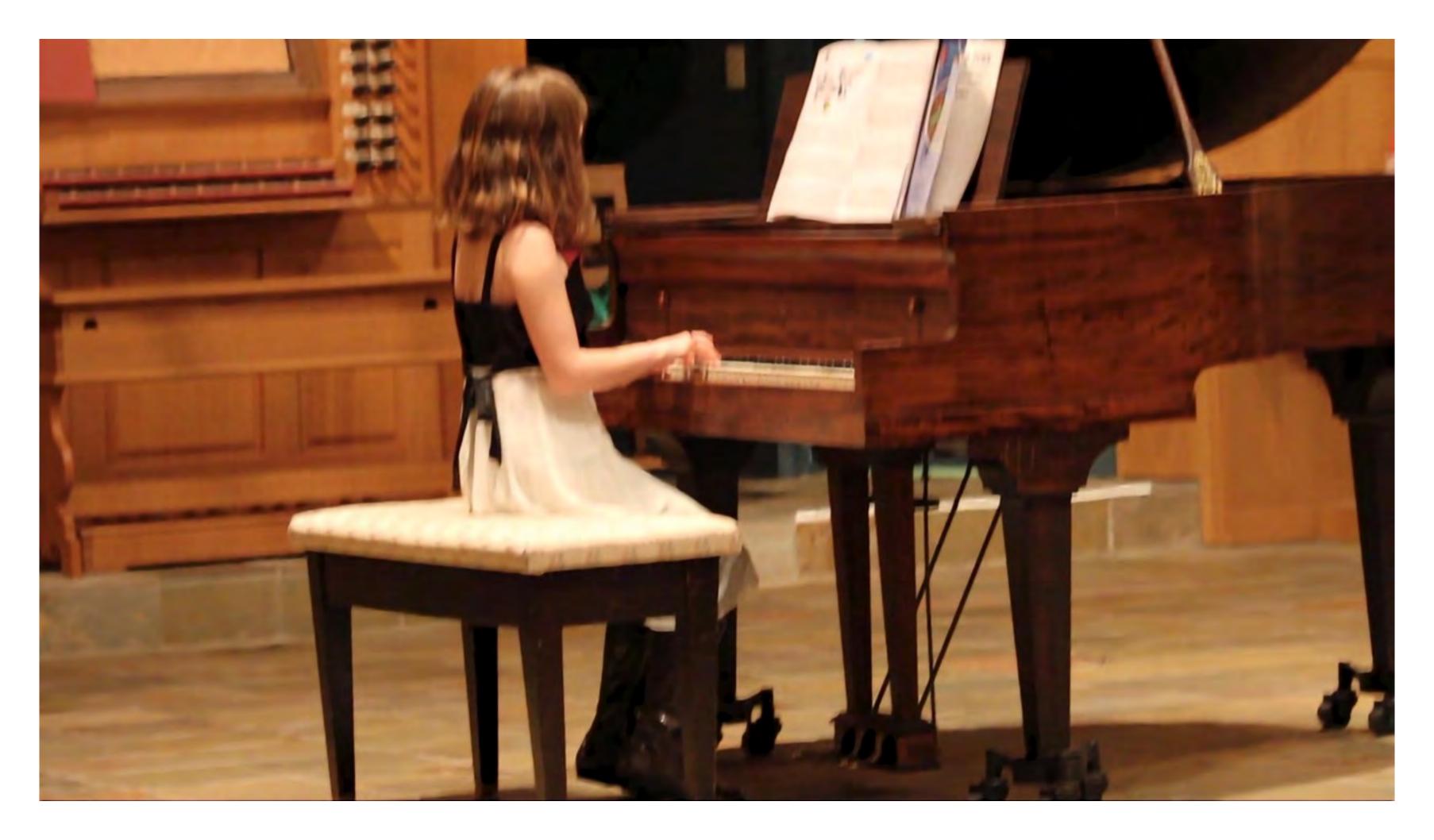
LODeblur [48]



Neural [1]



WFA [5]



DBN+Single



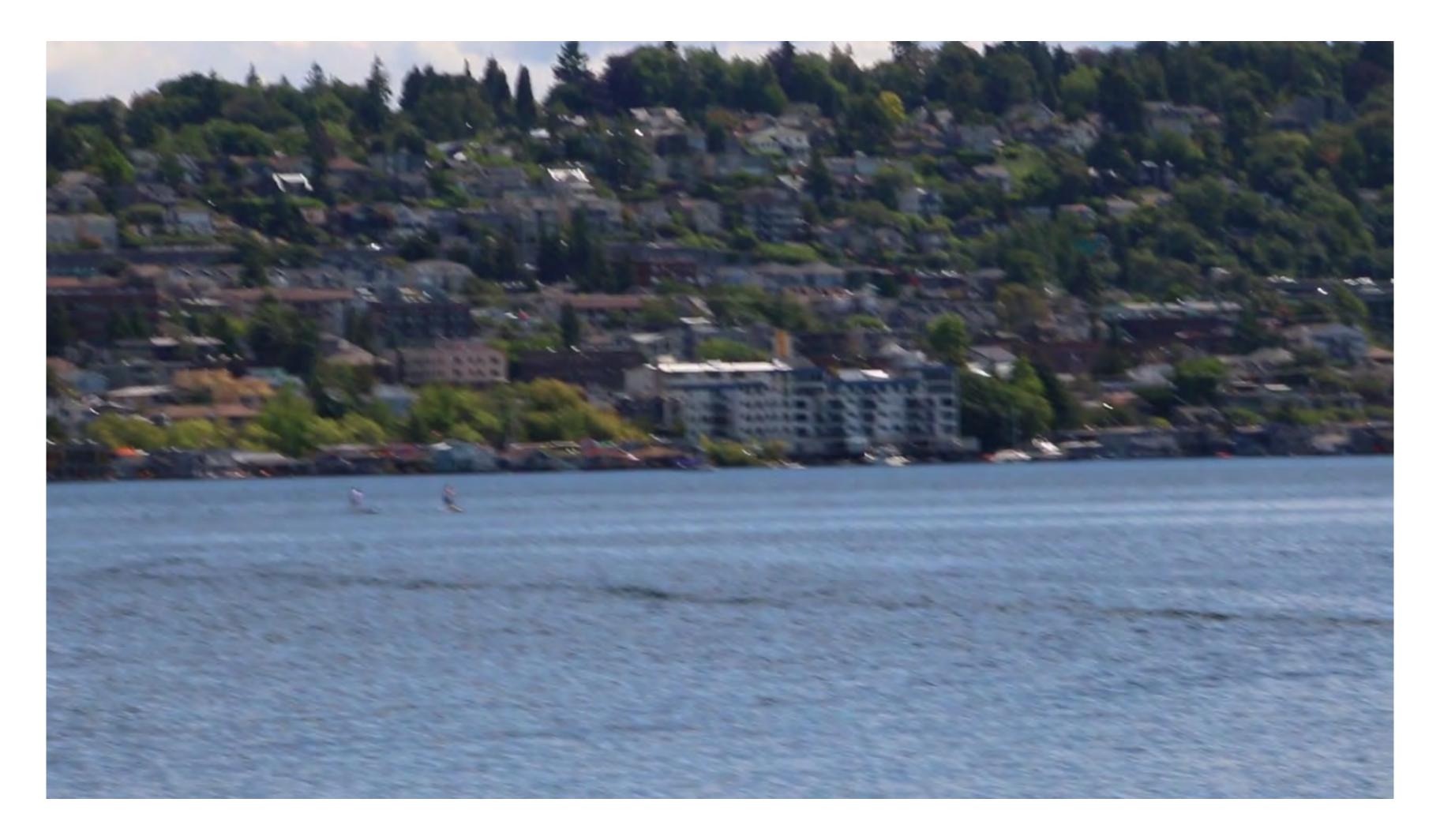
DBN+Noalign



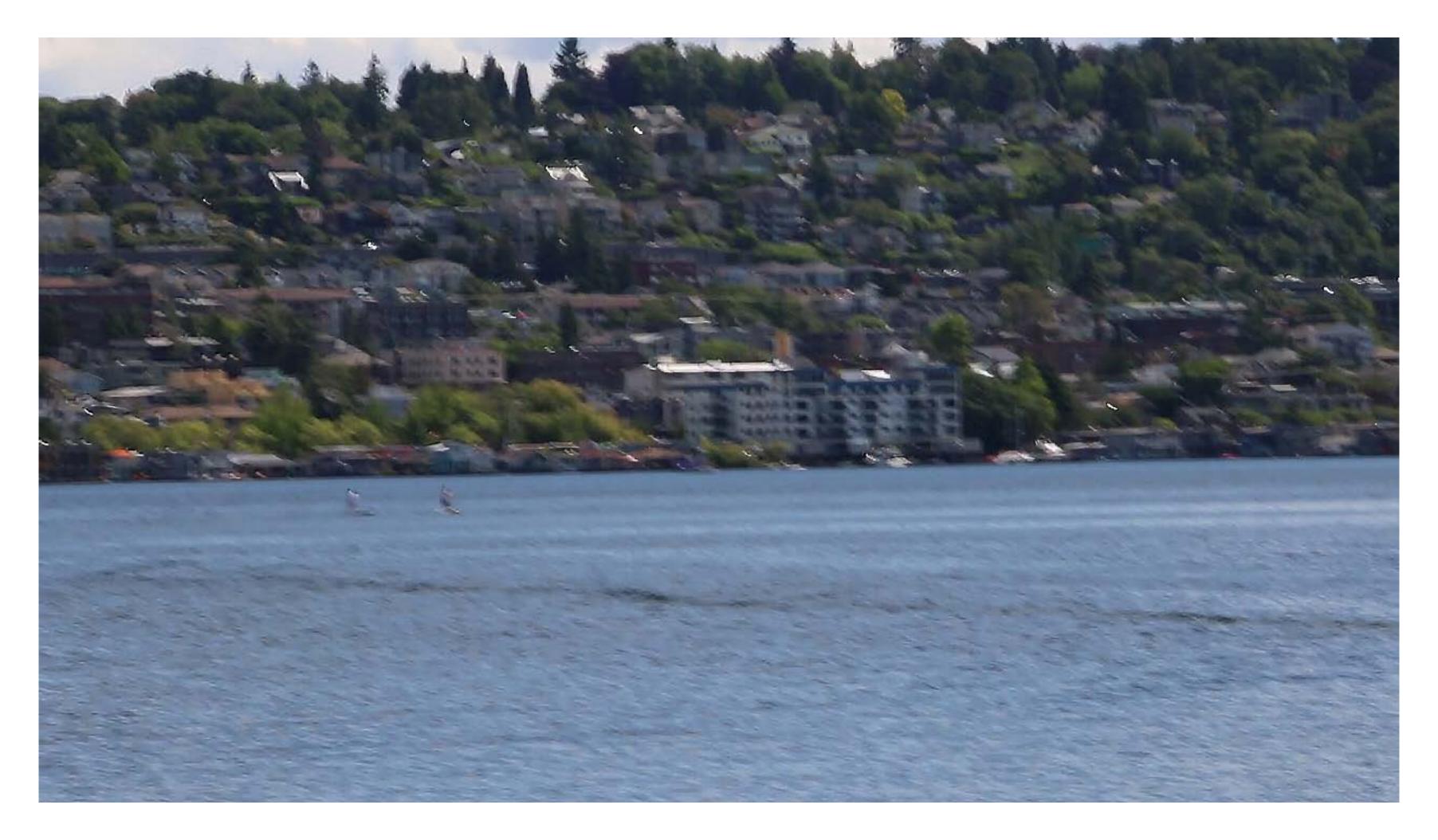
DBN+Homog



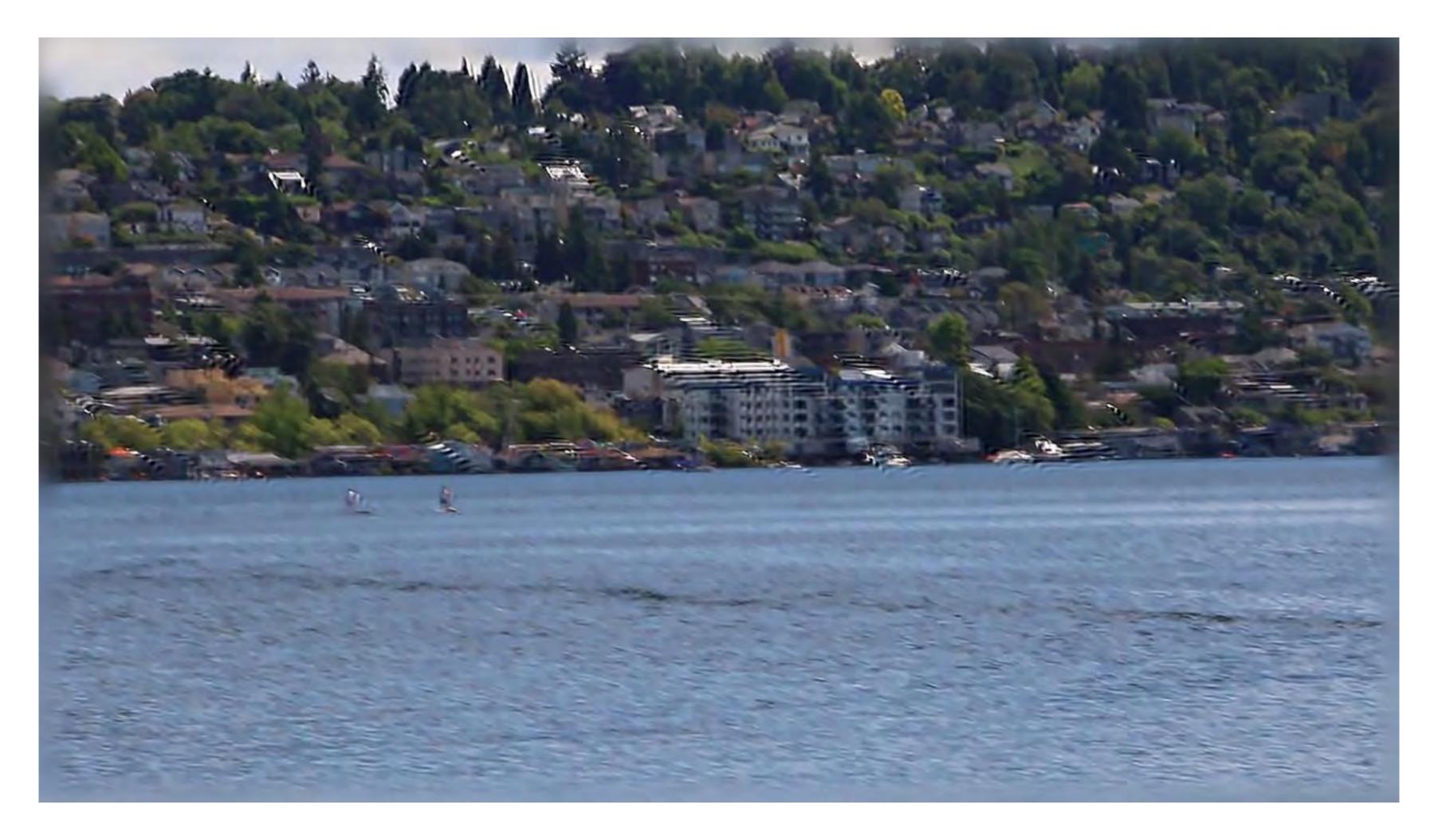
DBN+Flow



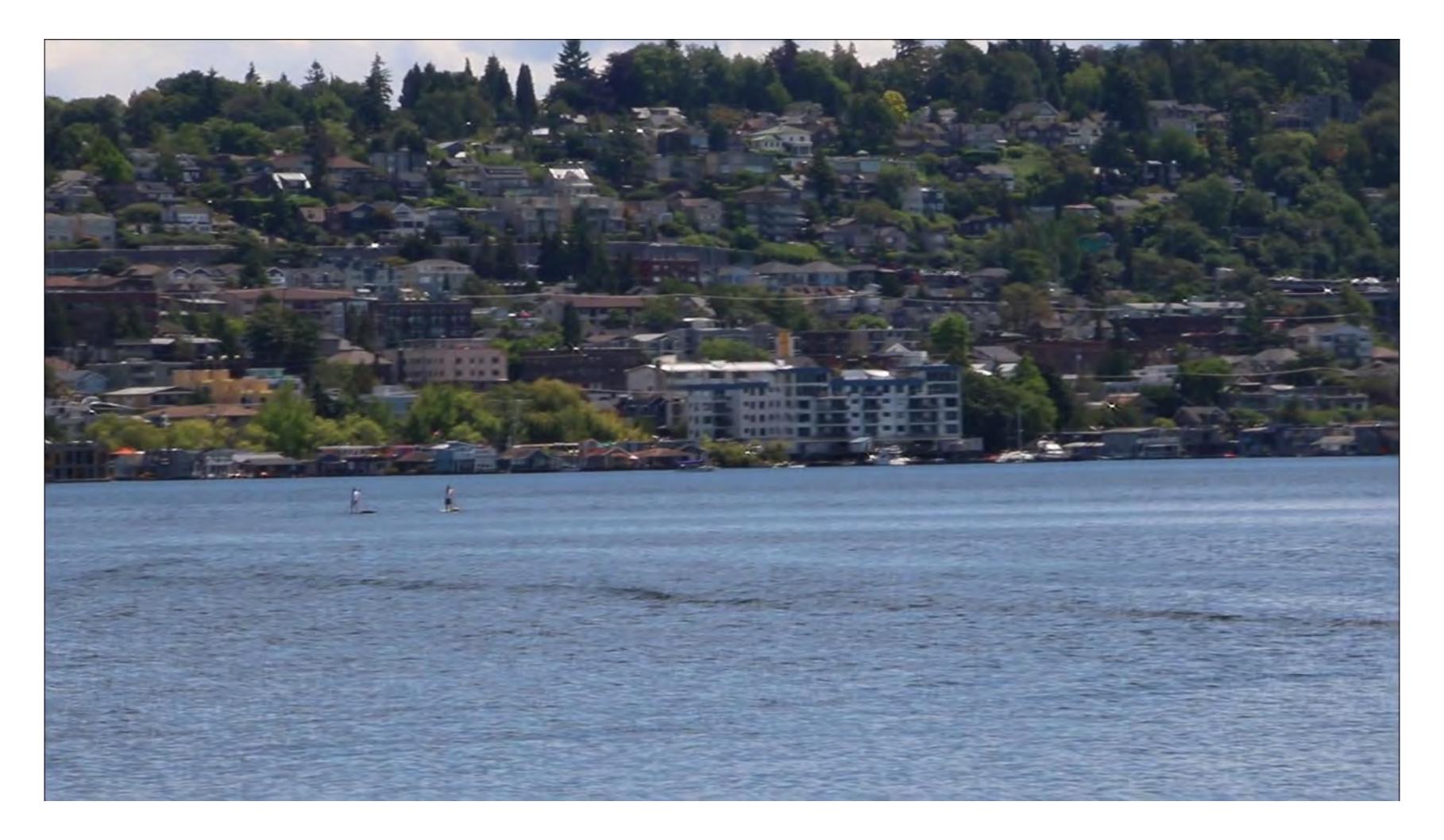
input



L0Deblur[48]



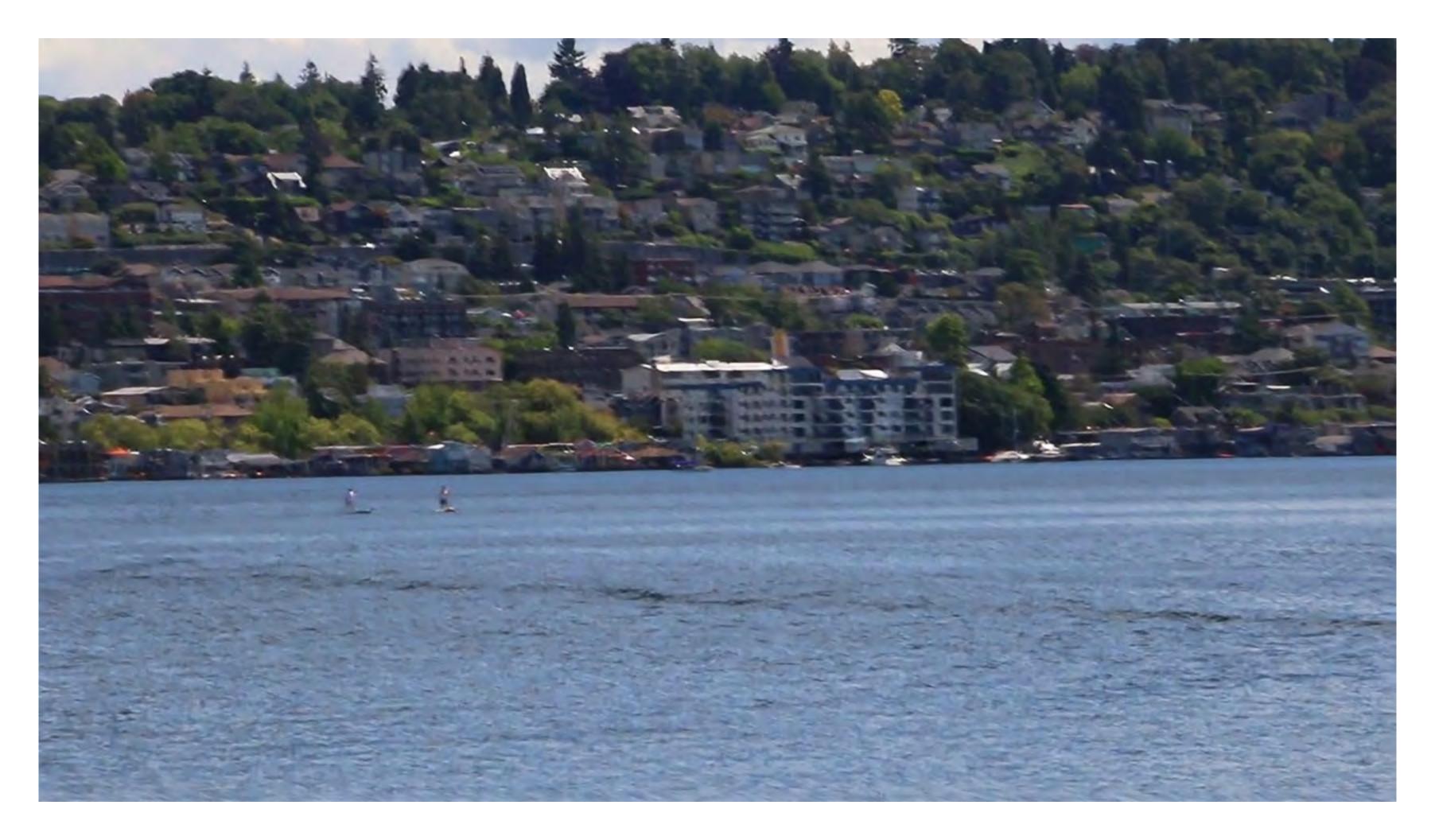
Neural [1]



WFA [5]



DBN+Single



DBN+Noalign



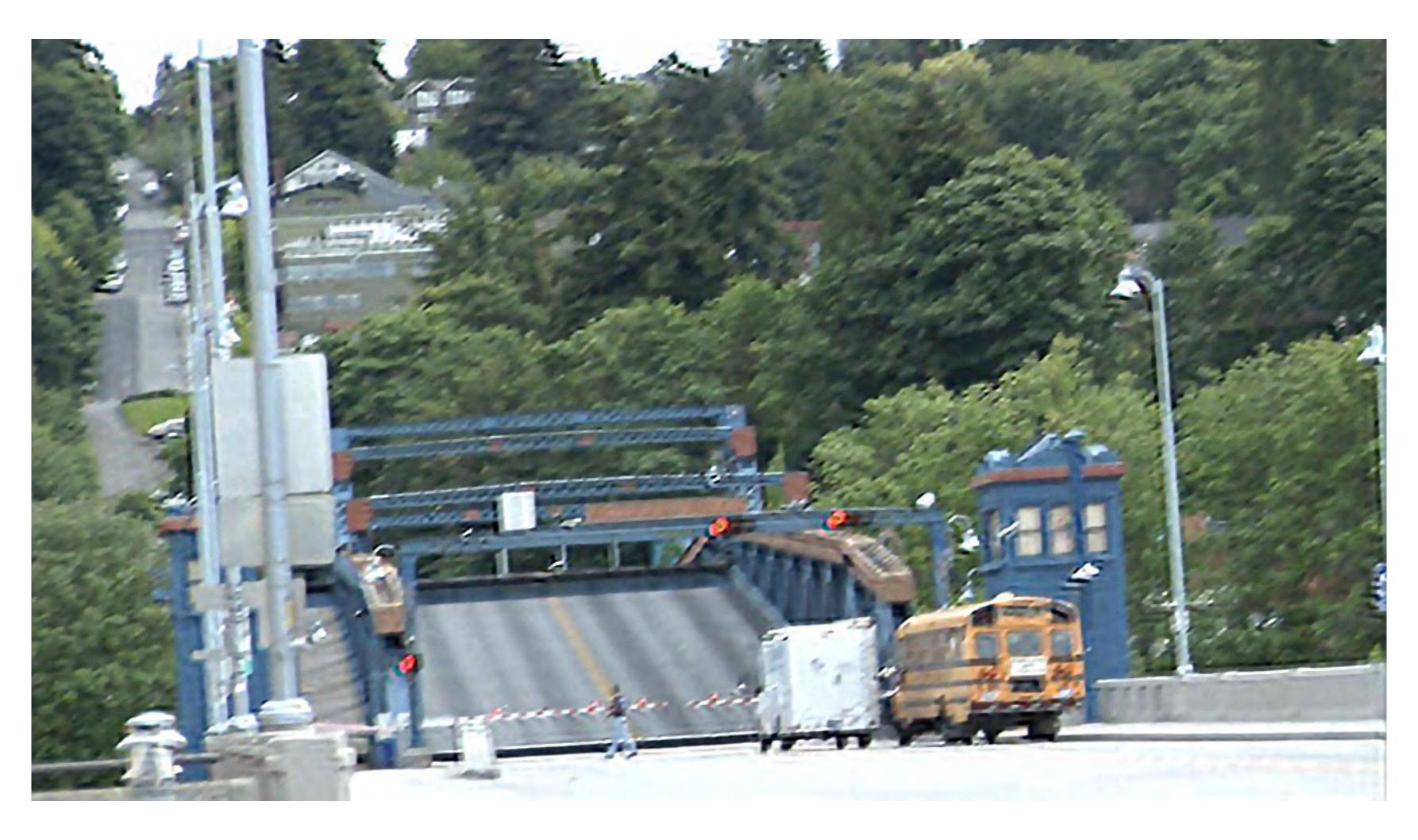
DBN+Homog



DBN+Flow



input



PSDeblur



Cho [3]



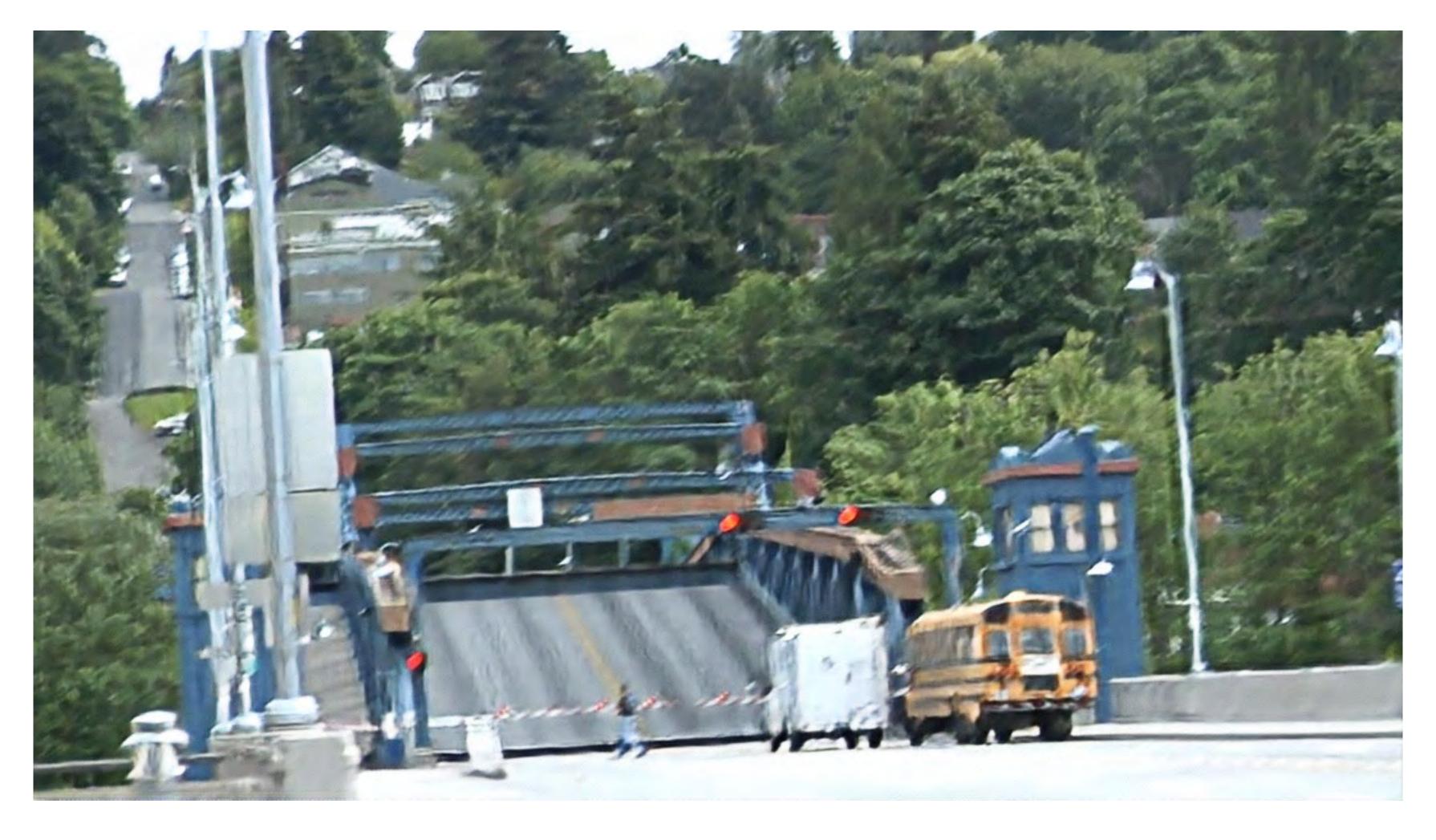
Kim and Lee [15]



WFA [5]



DBN+Single



DBN+Noalign



DBN+Homog



DBN+Flow



input



PSDeblur



Cho [3]



Kim and Lee [15]



WFA [5]



DBN+Single



DBN+Noalign

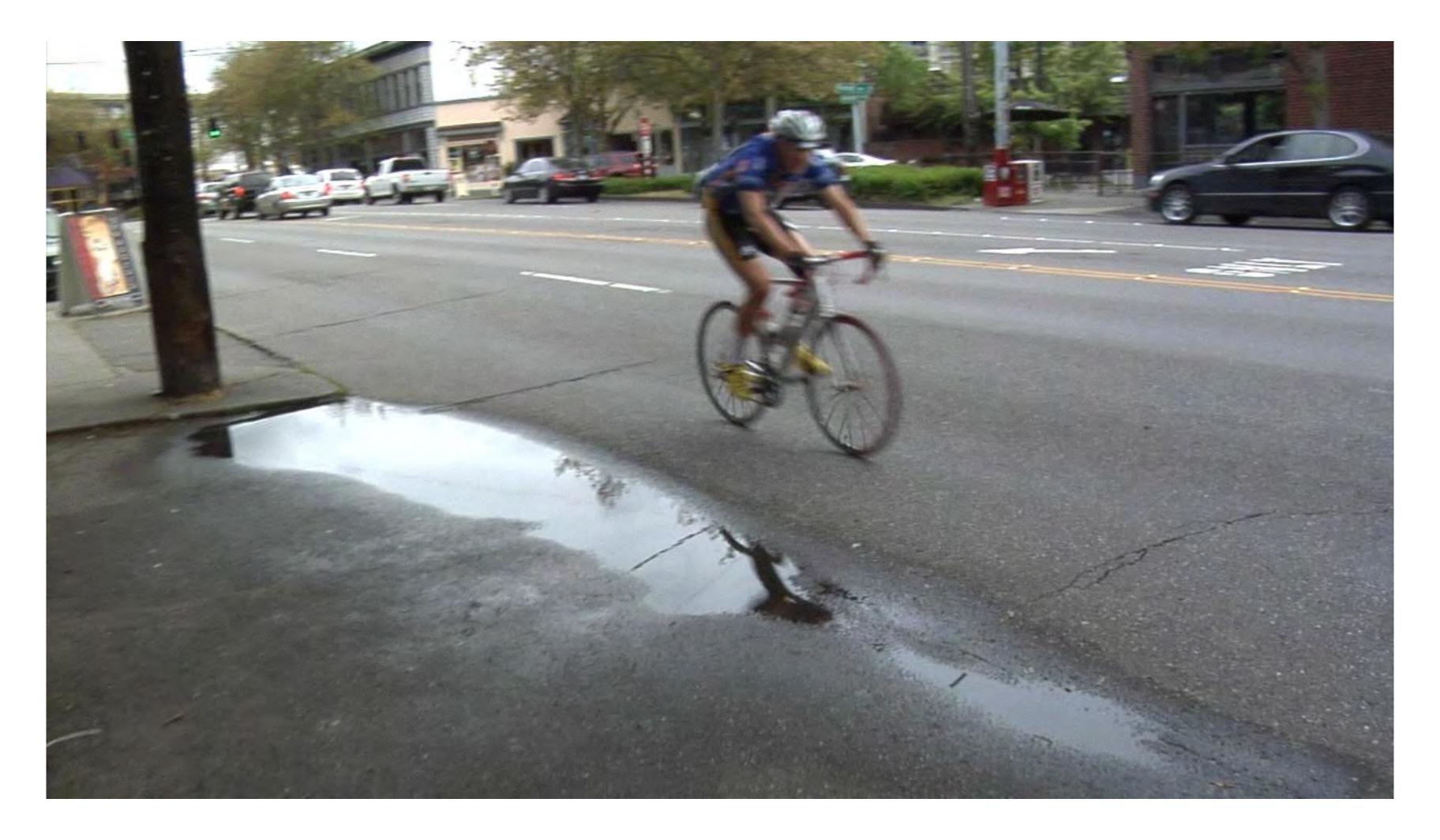


DBN+Homog

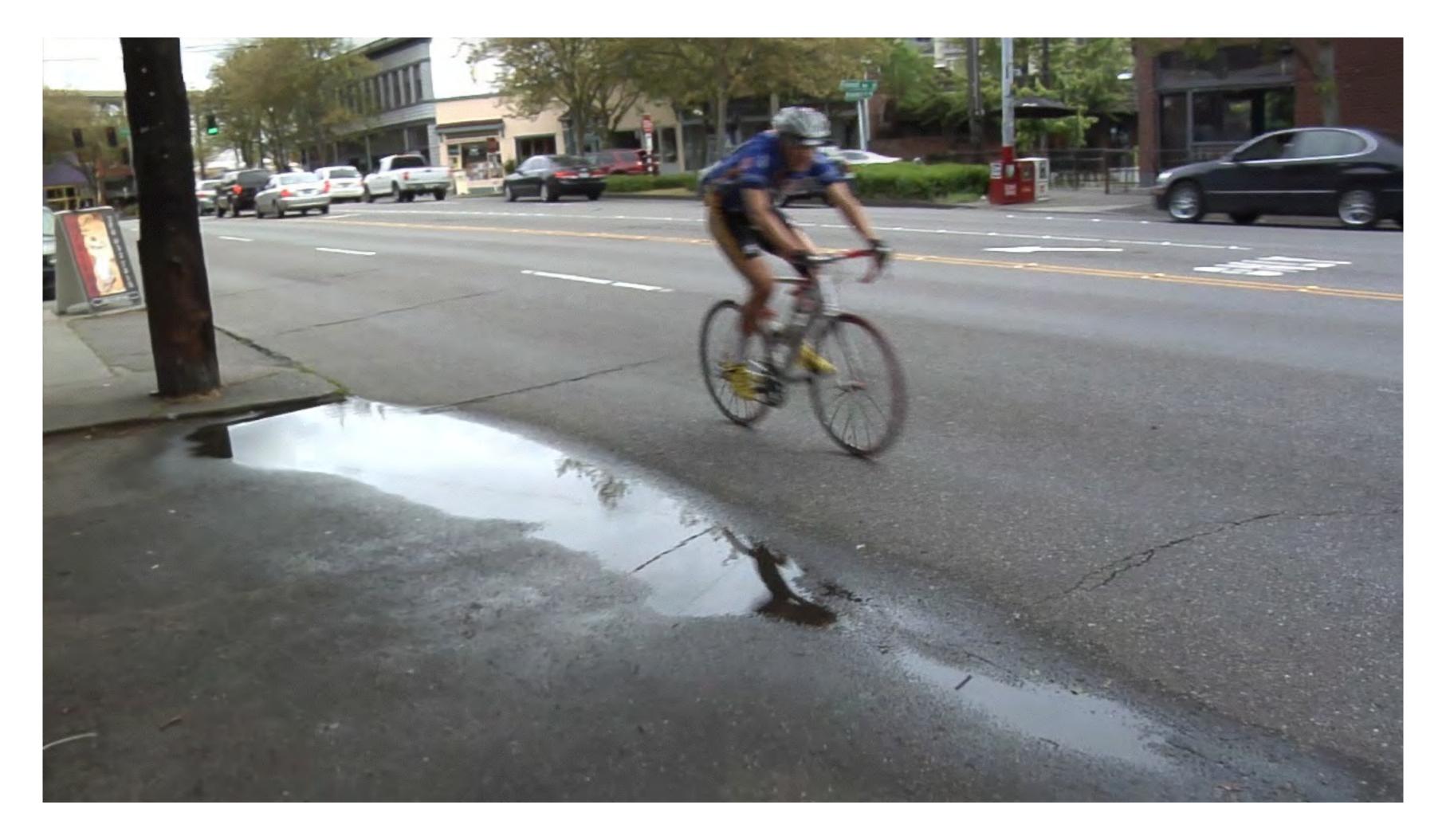


DBN+Flow

Additional Comparisons with Cho [3], Kim and Lee [15], and WFA [5]



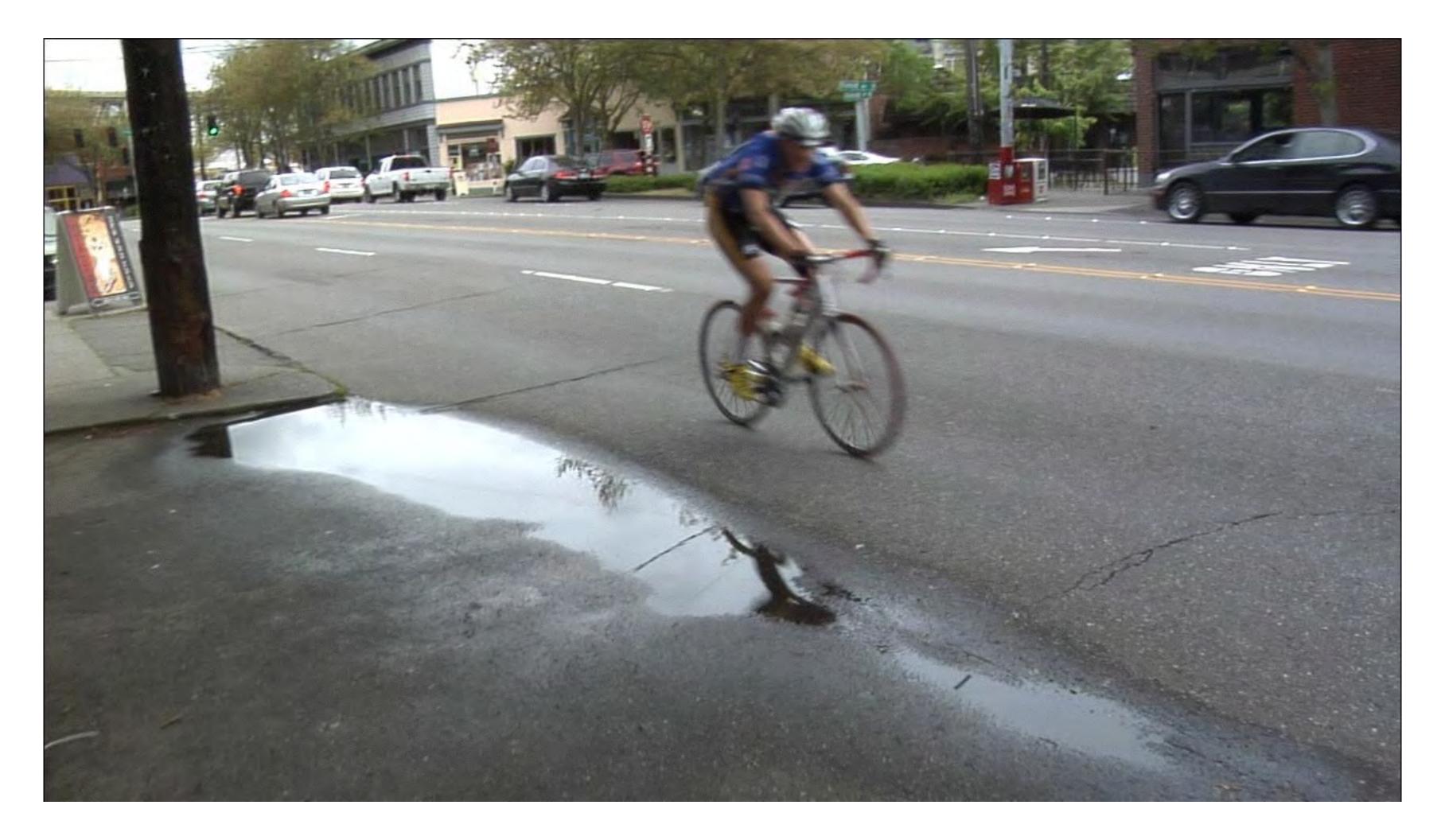
input



Cho [3]



Kim and Lee [15]



WFA [5]



DBN+Single



DBN+Noalign



DBN+Homog



DBN+Flow



input



Cho [3]



Kim and Lee [15]



WFA [5]



DBN+Single



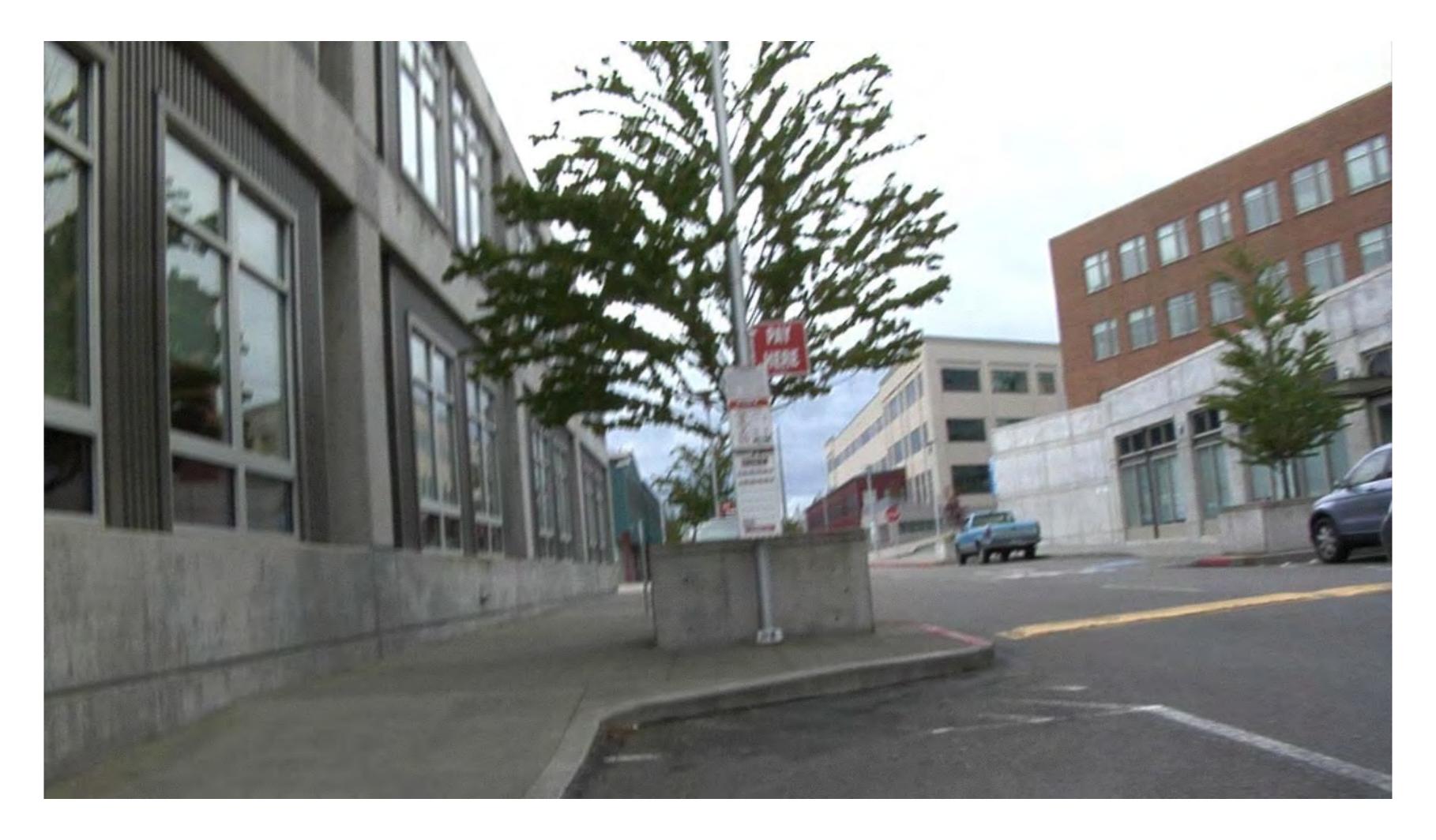
DBN+Noalign



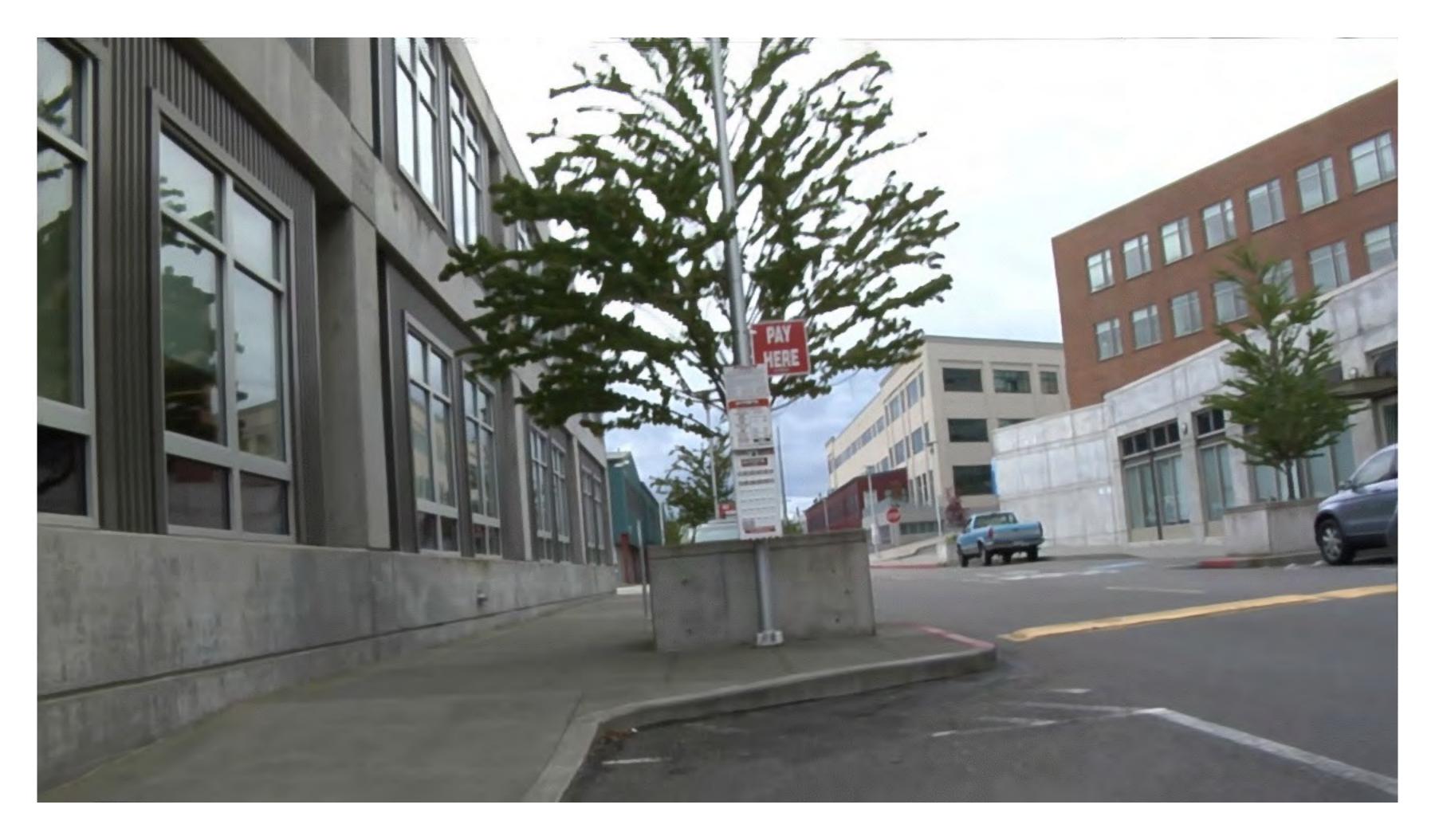
DBN+Homog



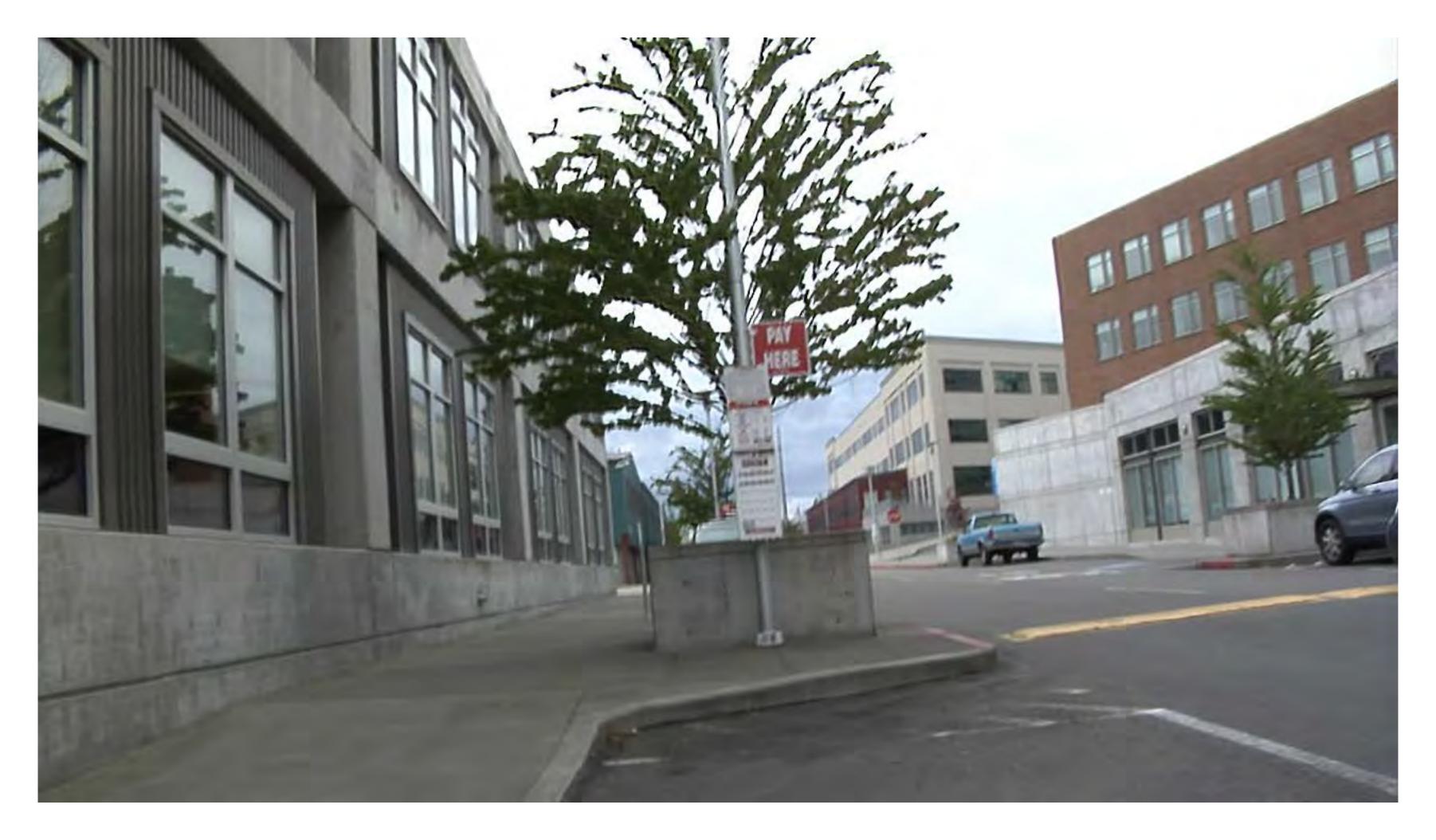
DBN+Flow



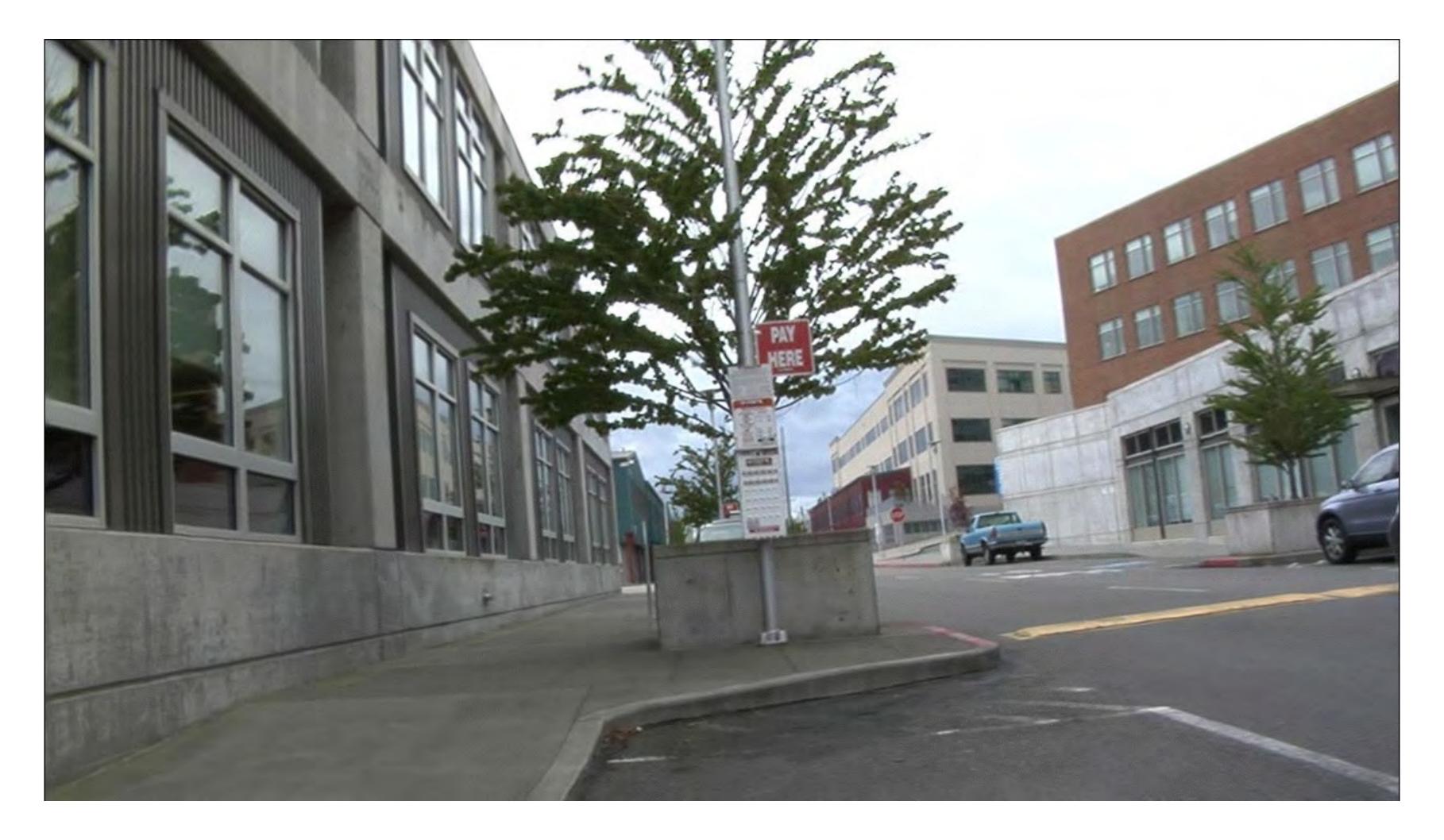
input



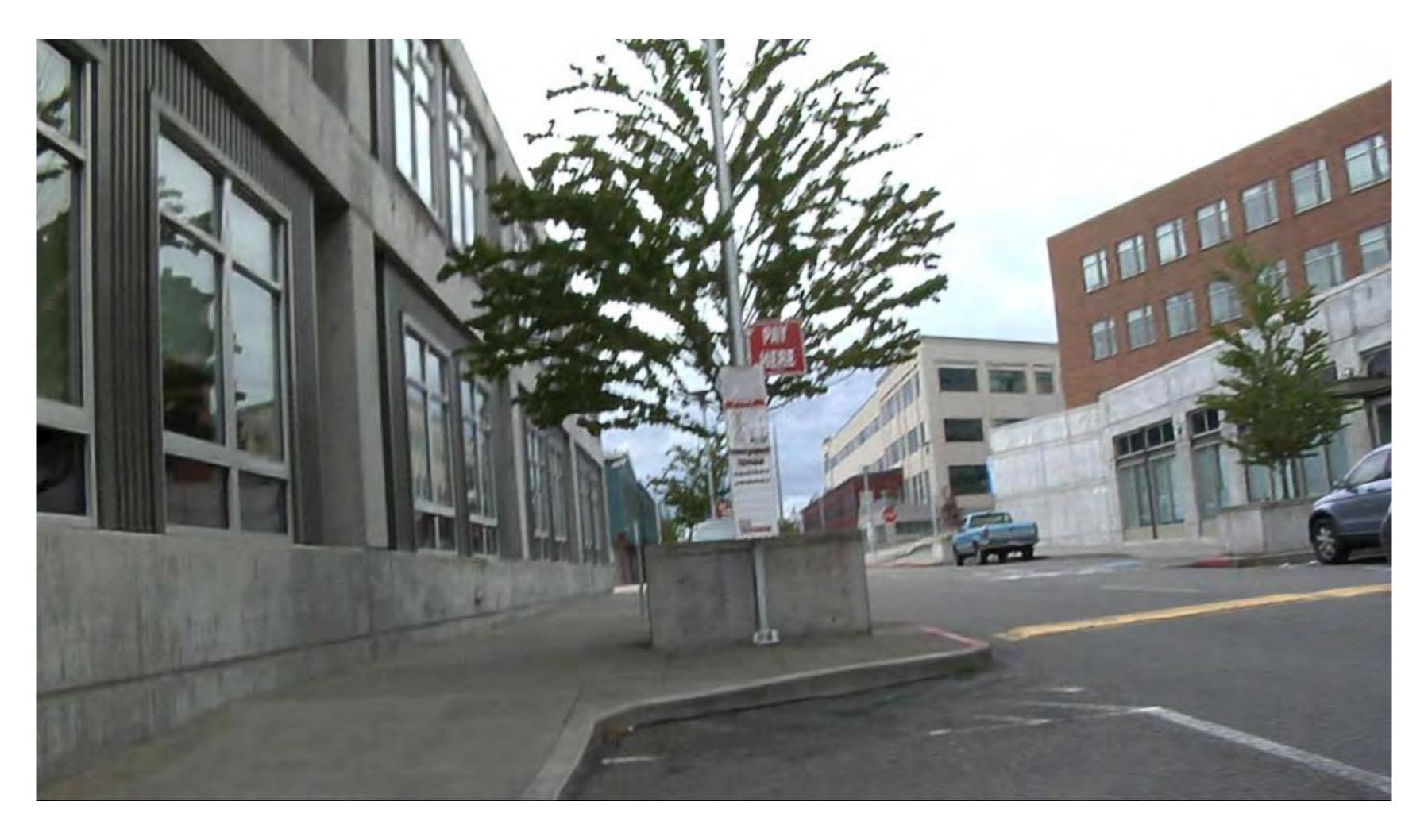
Cho [3]



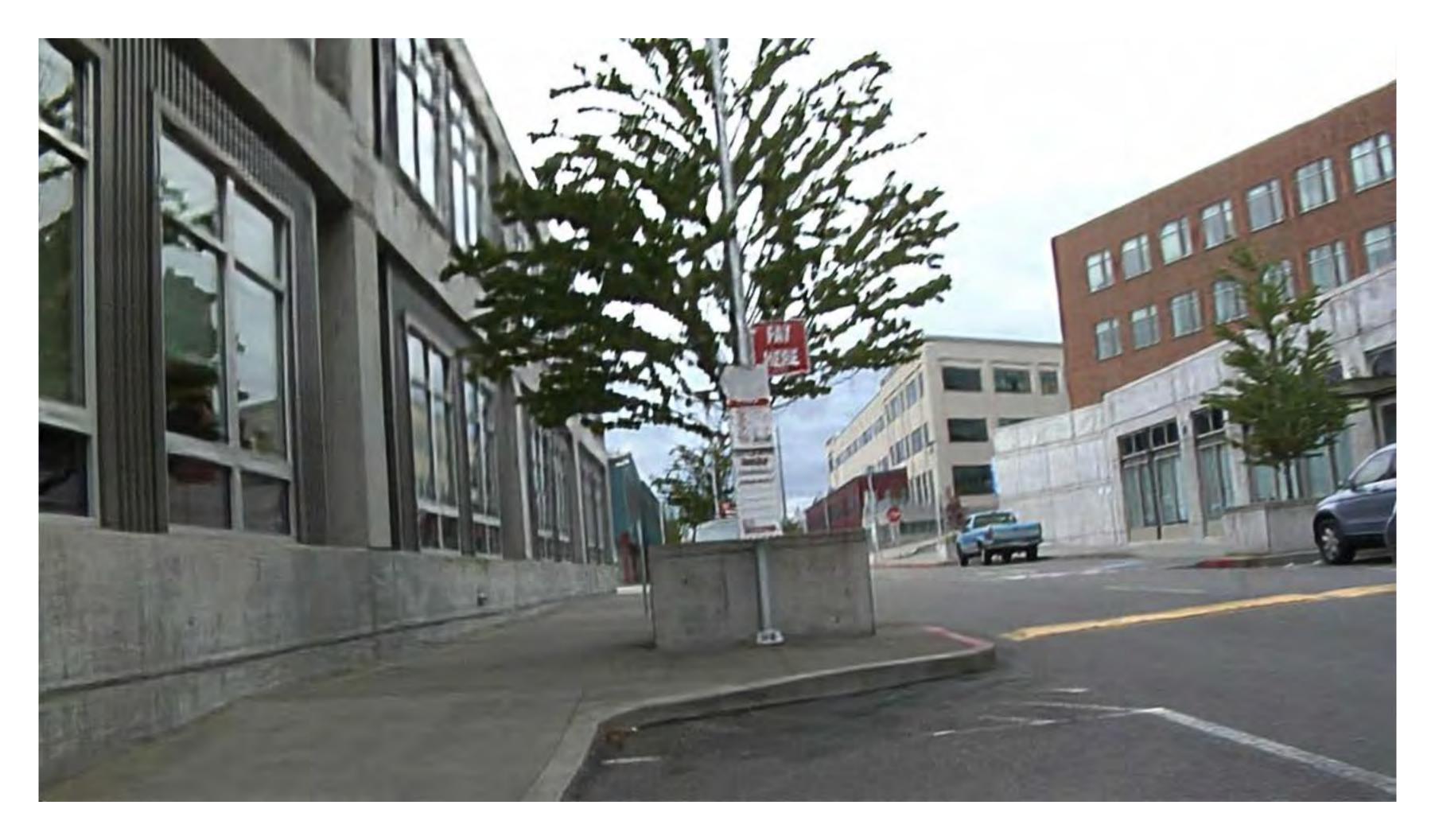
Kim and Lee [15]



WFA [5]



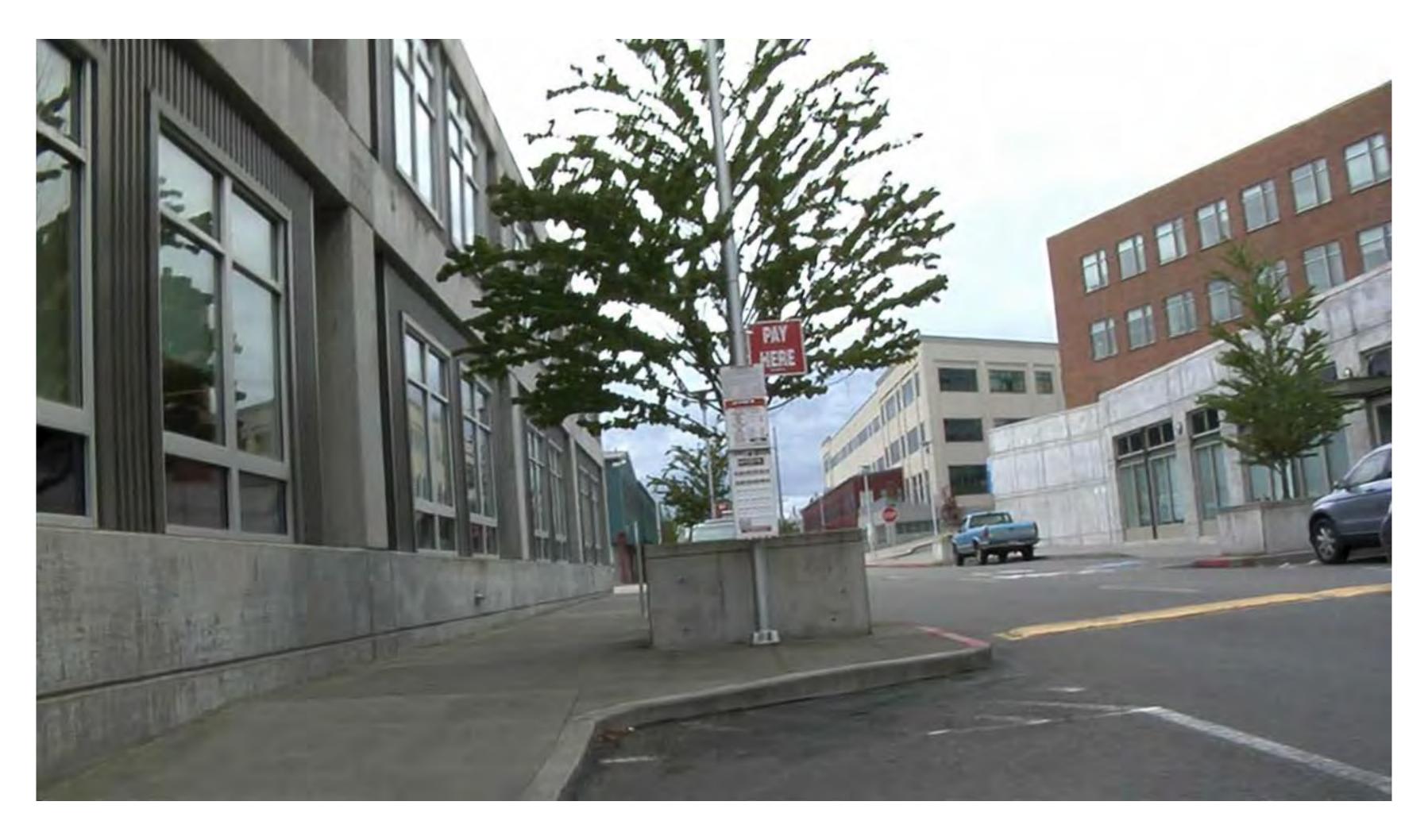
DBN+Single



DBN+Noalign

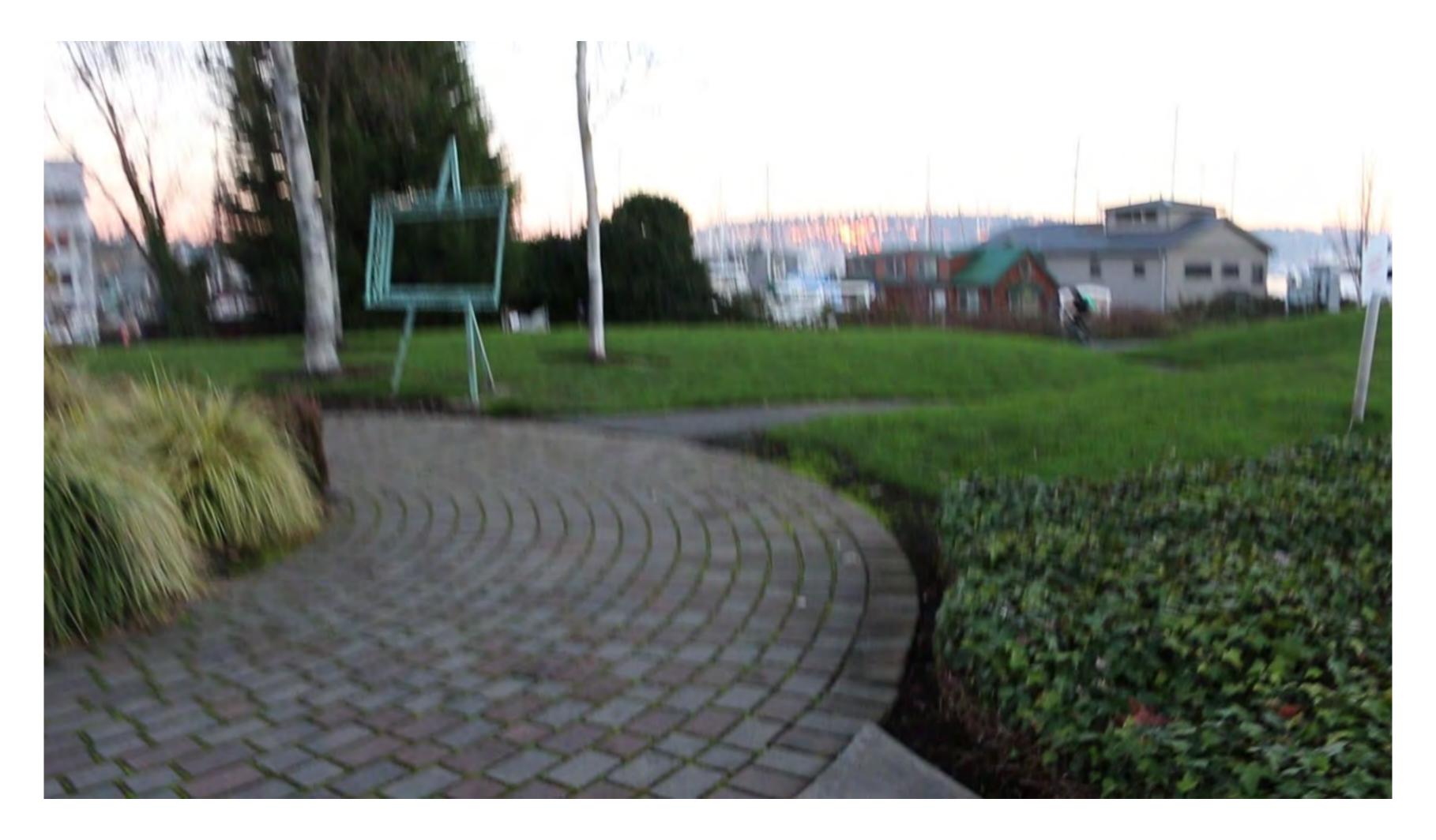


DBN+Homog

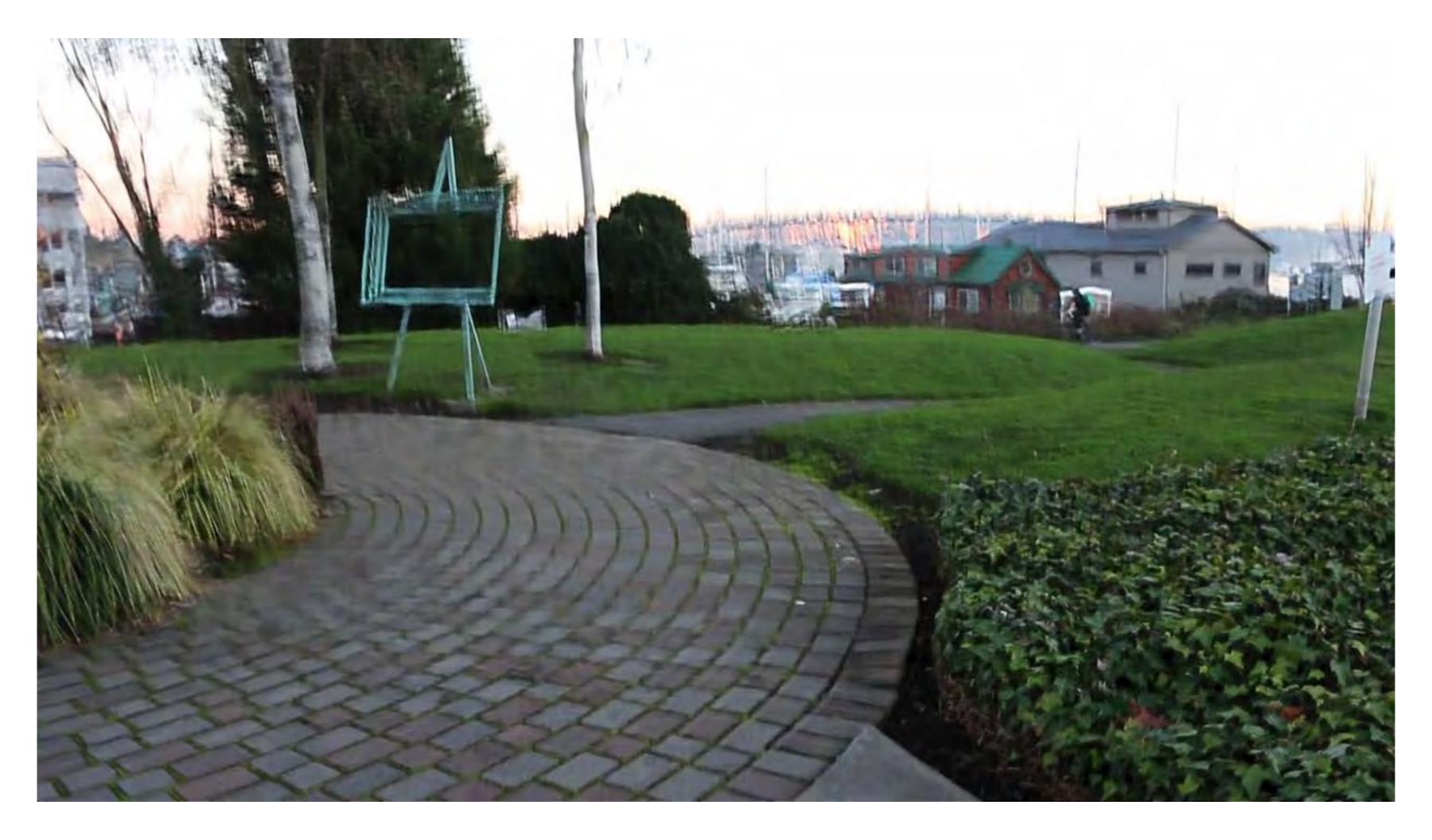


DBN+Flow

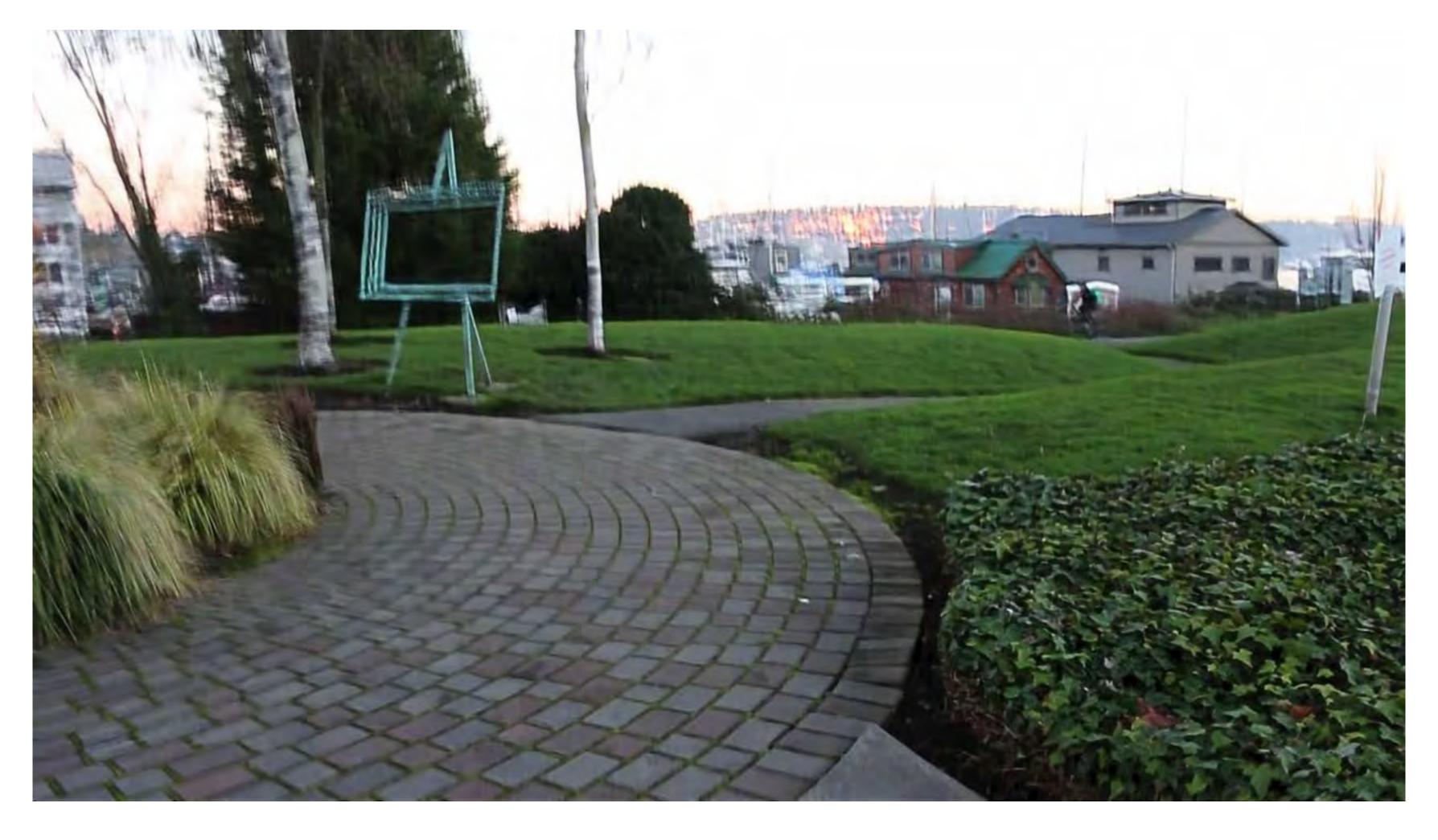
Additional Results



input



DBN+Noalign



DBN+Homog



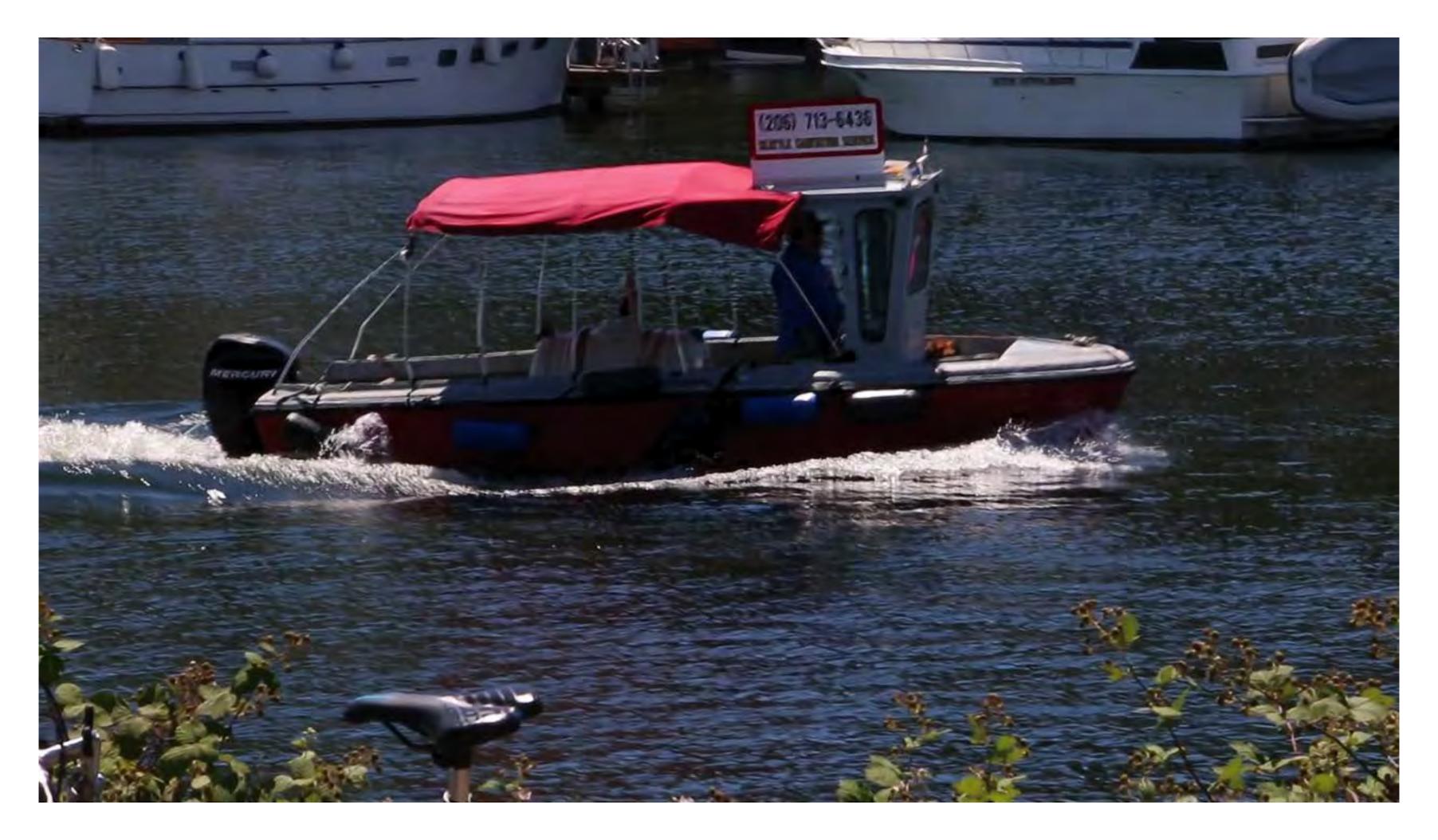
DBN+Flow



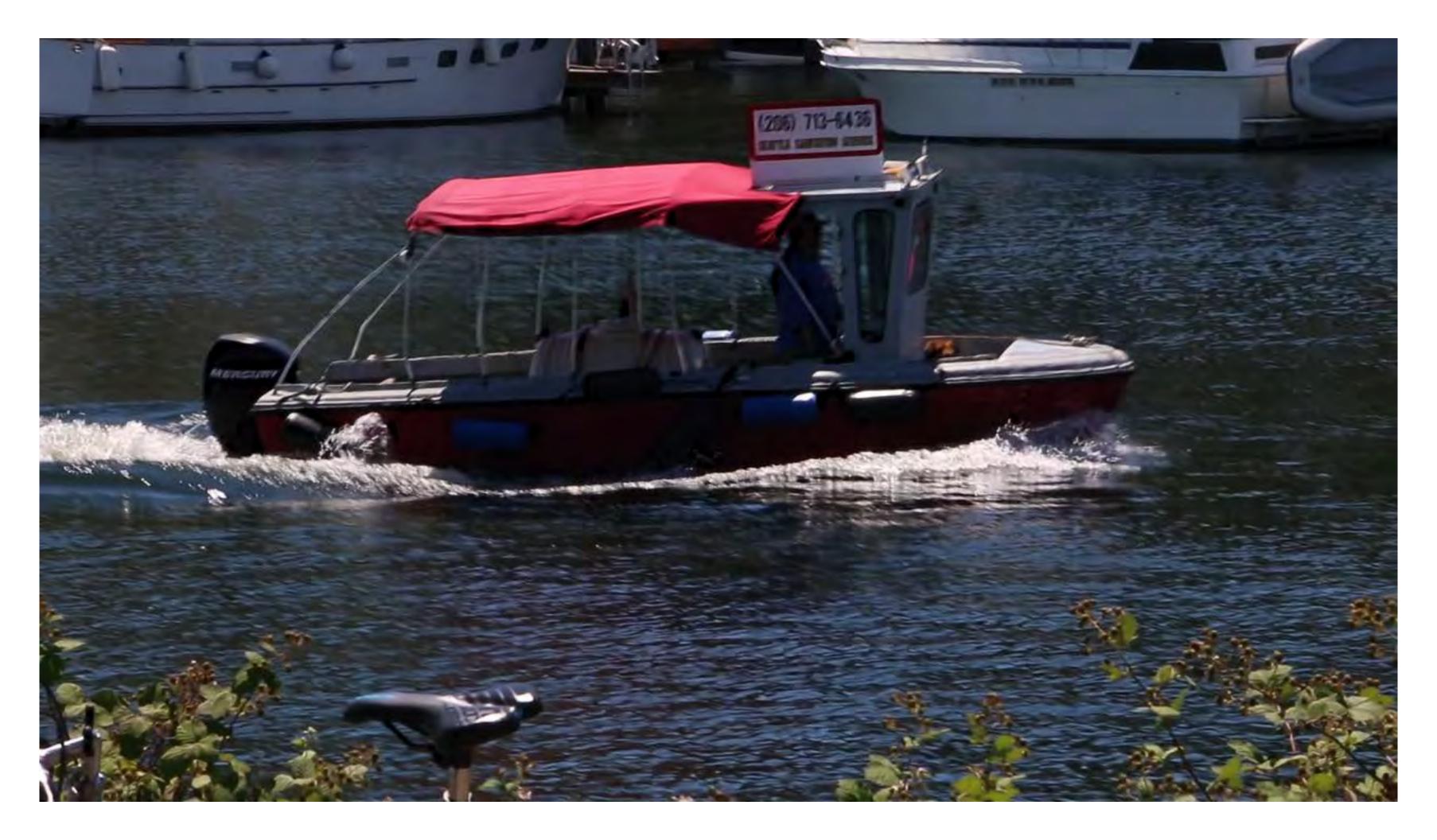
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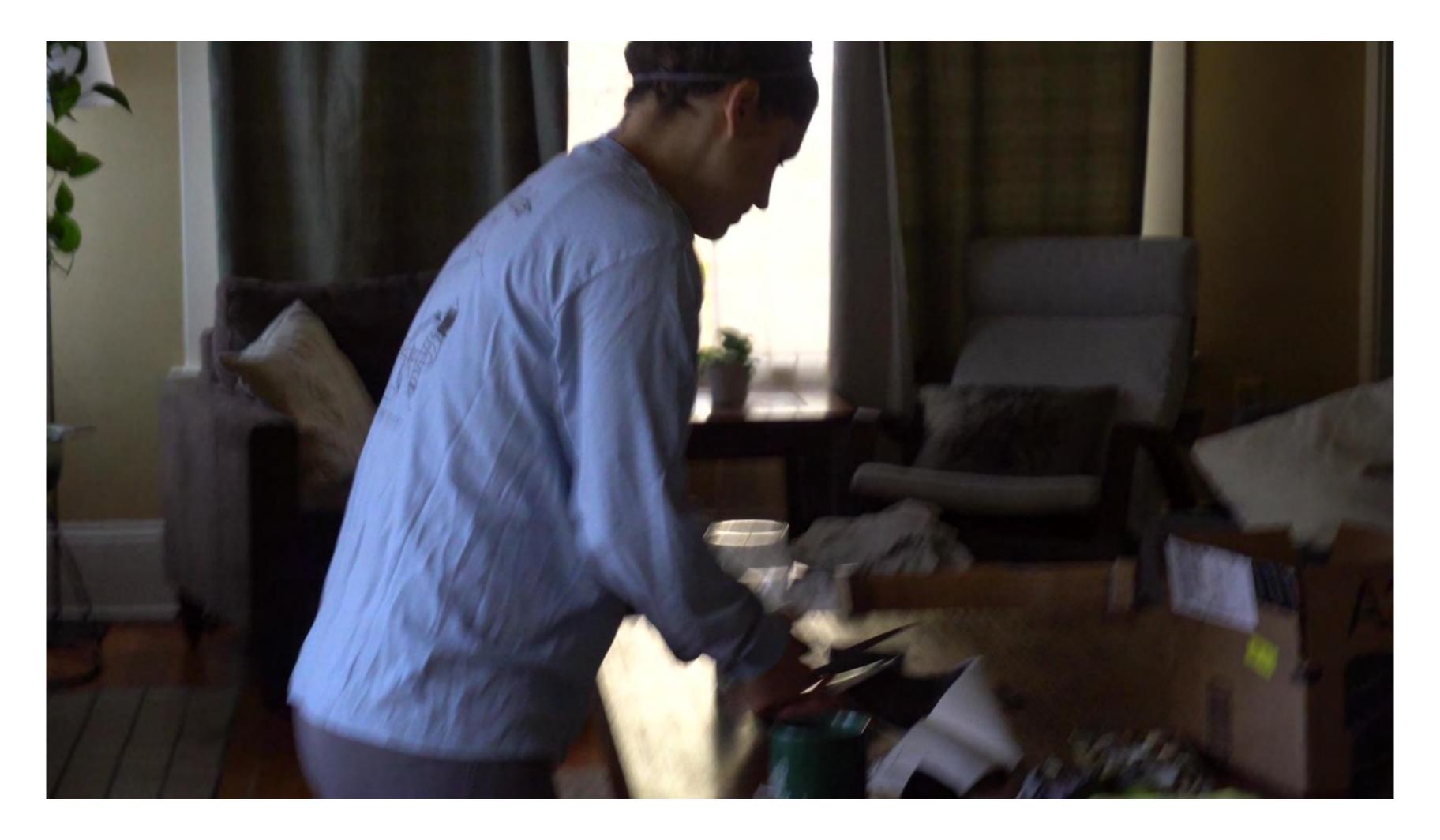
DBN+Noalign



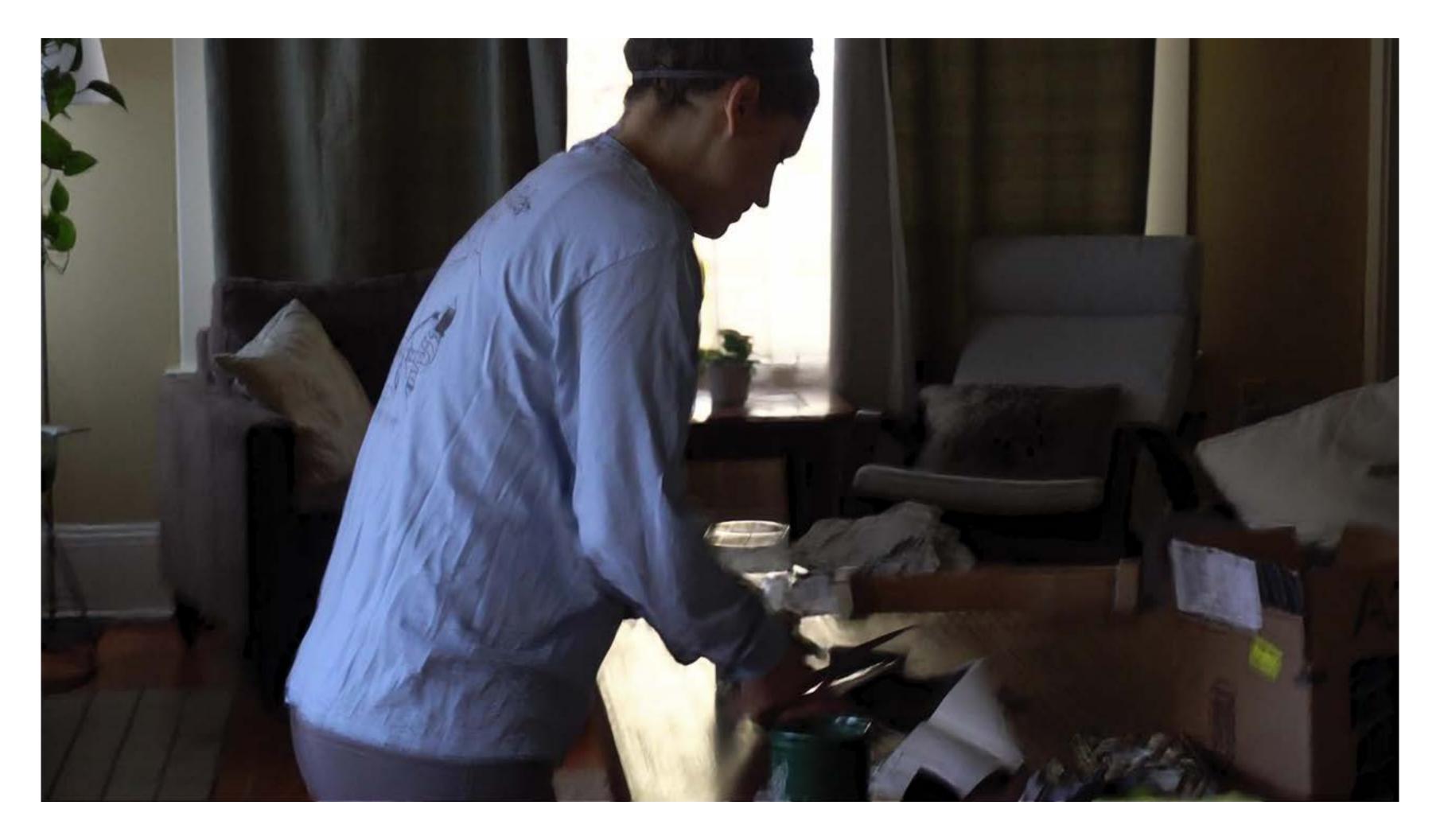
DBN+Homog



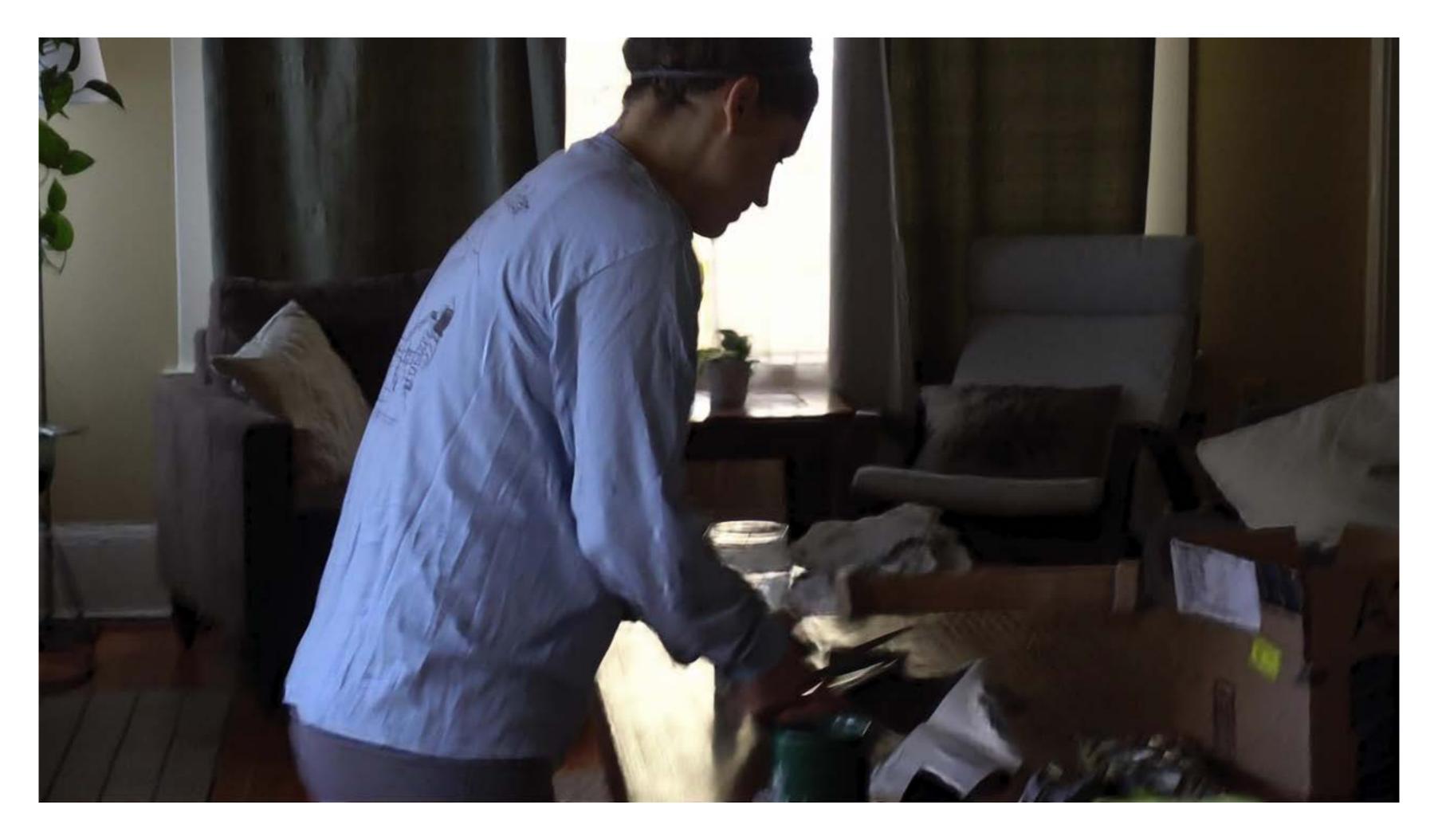
DBN+Flow



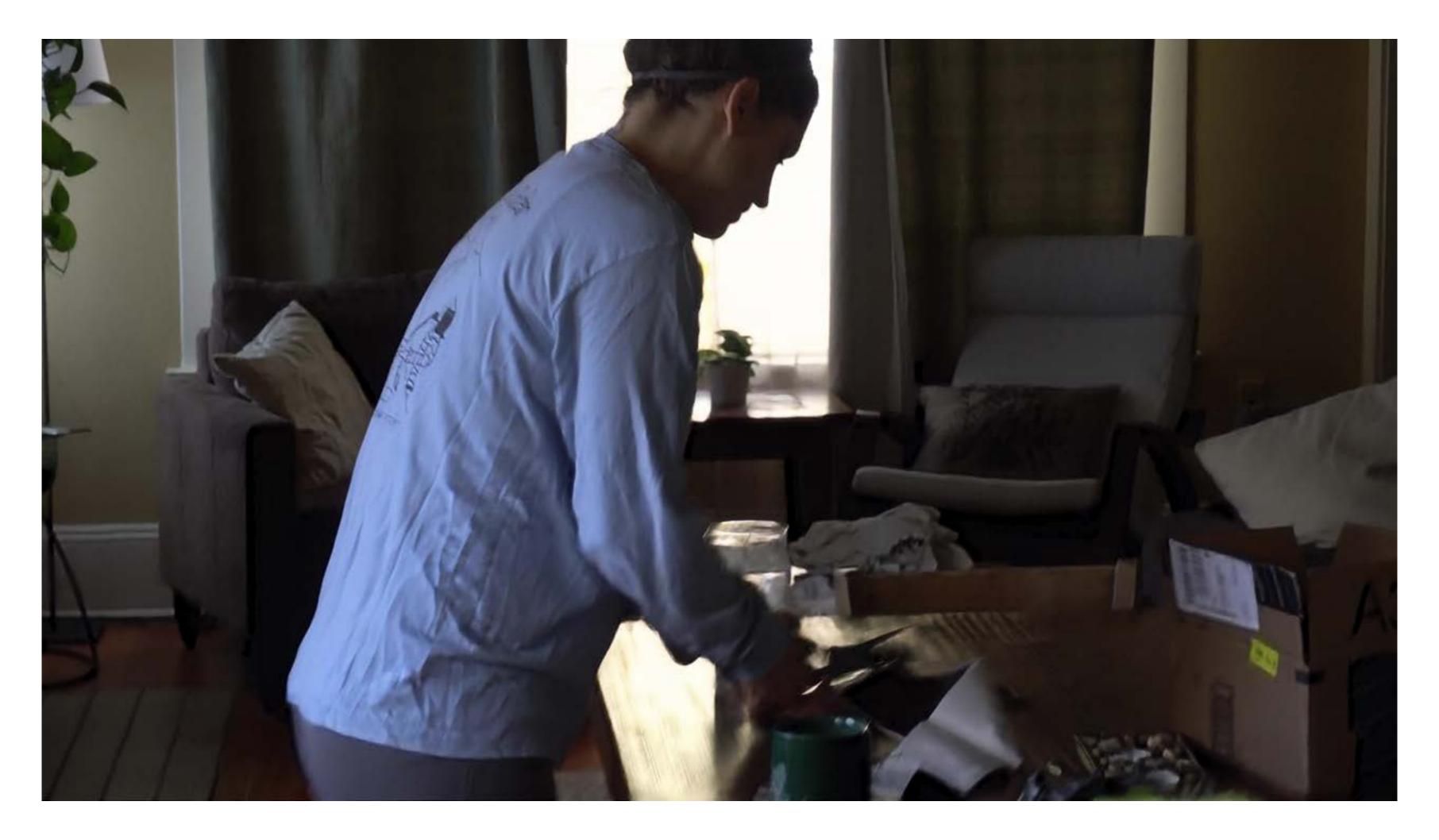
input



DBN+Noalign



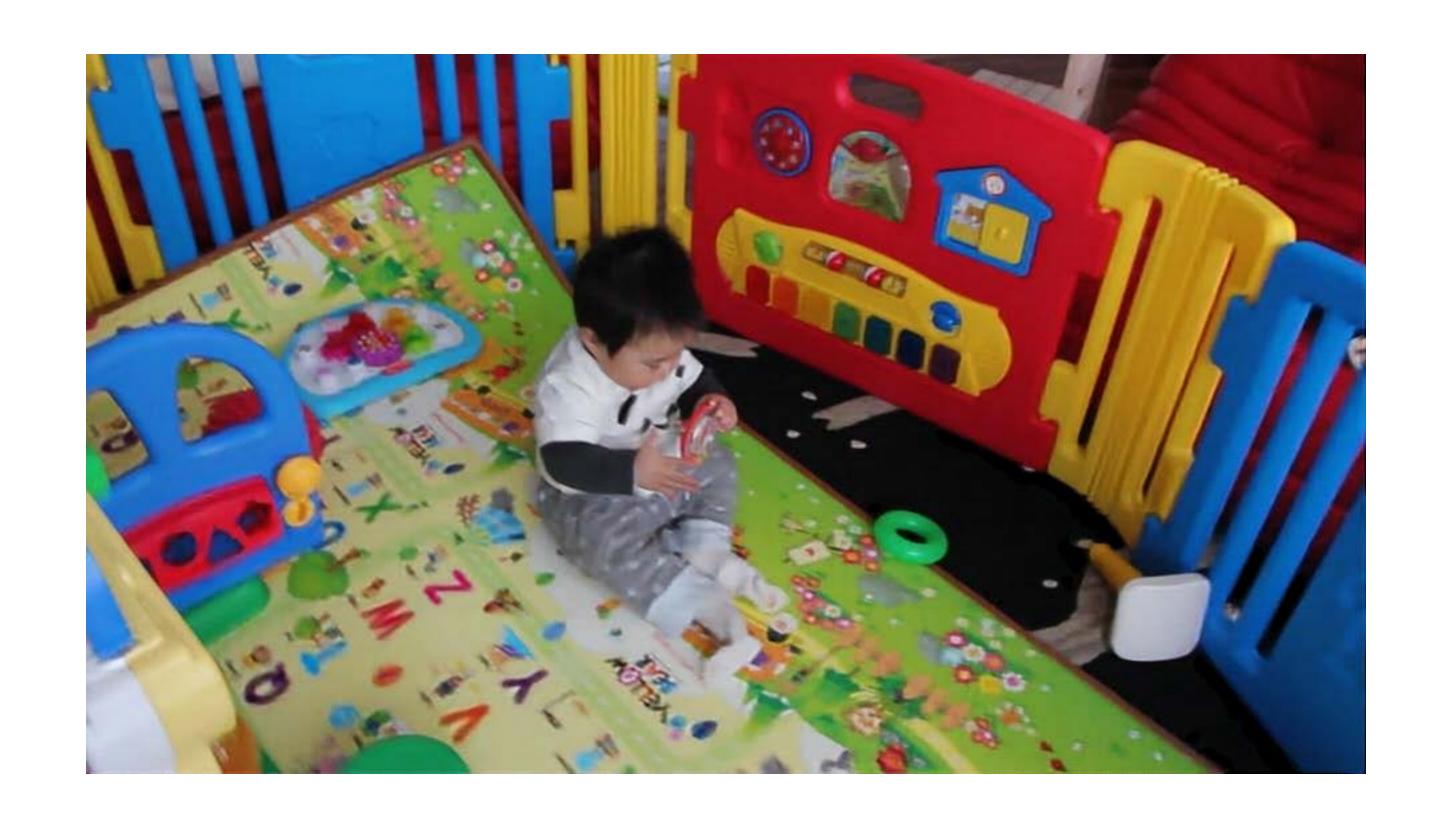
DBN+Homog



DBN+Flow



input

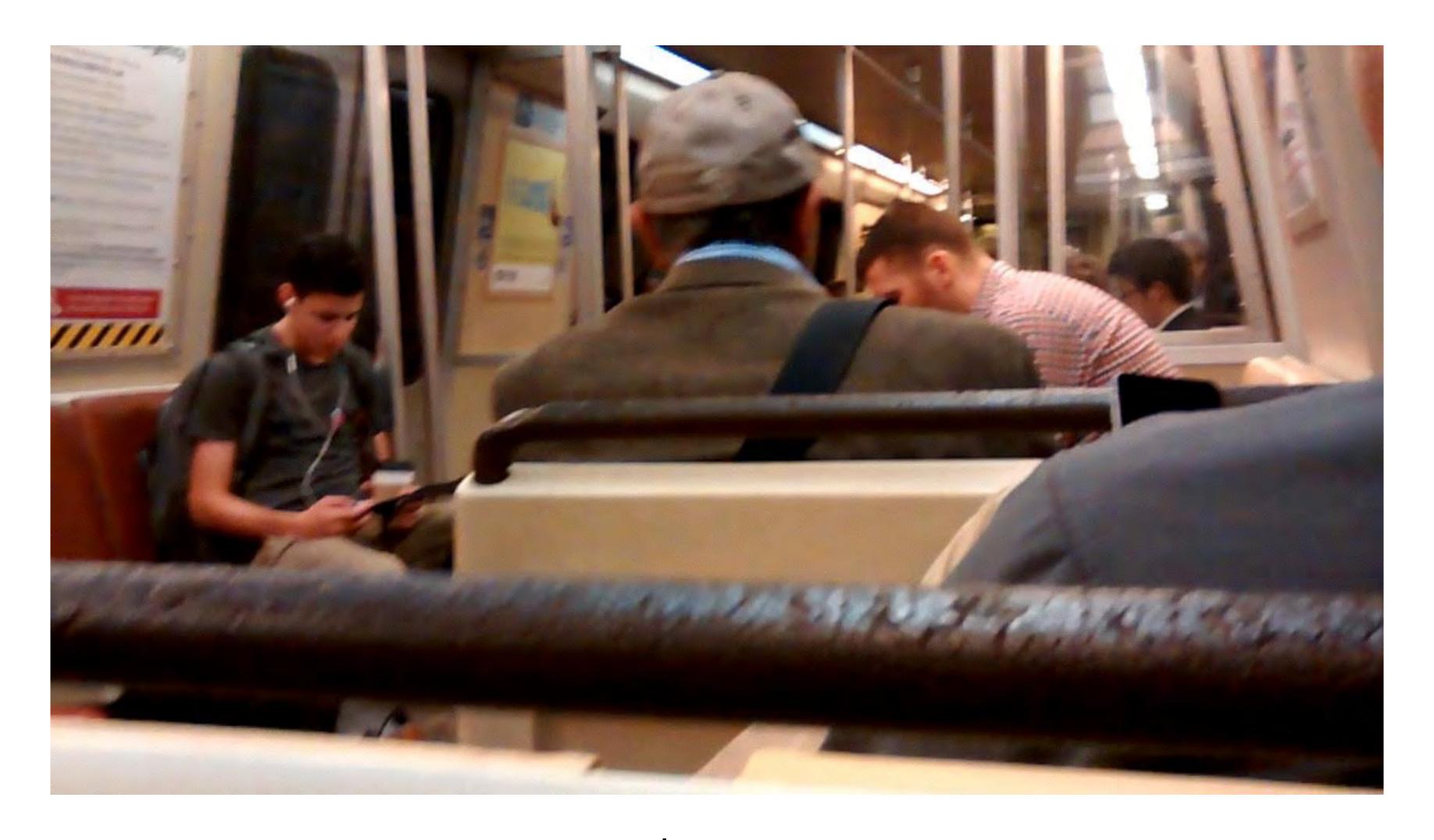


DBN+Noalign

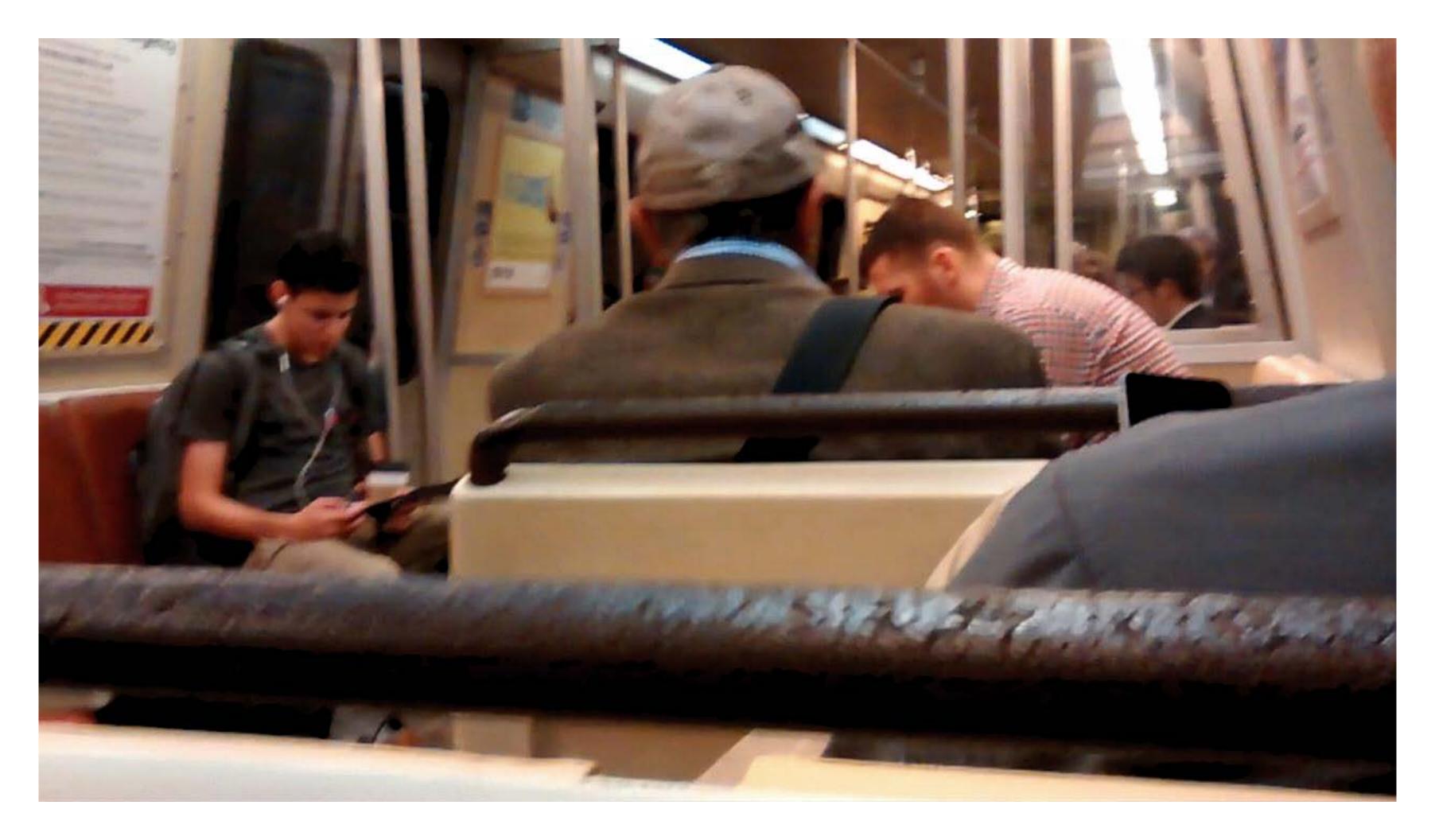


DBN+Homog

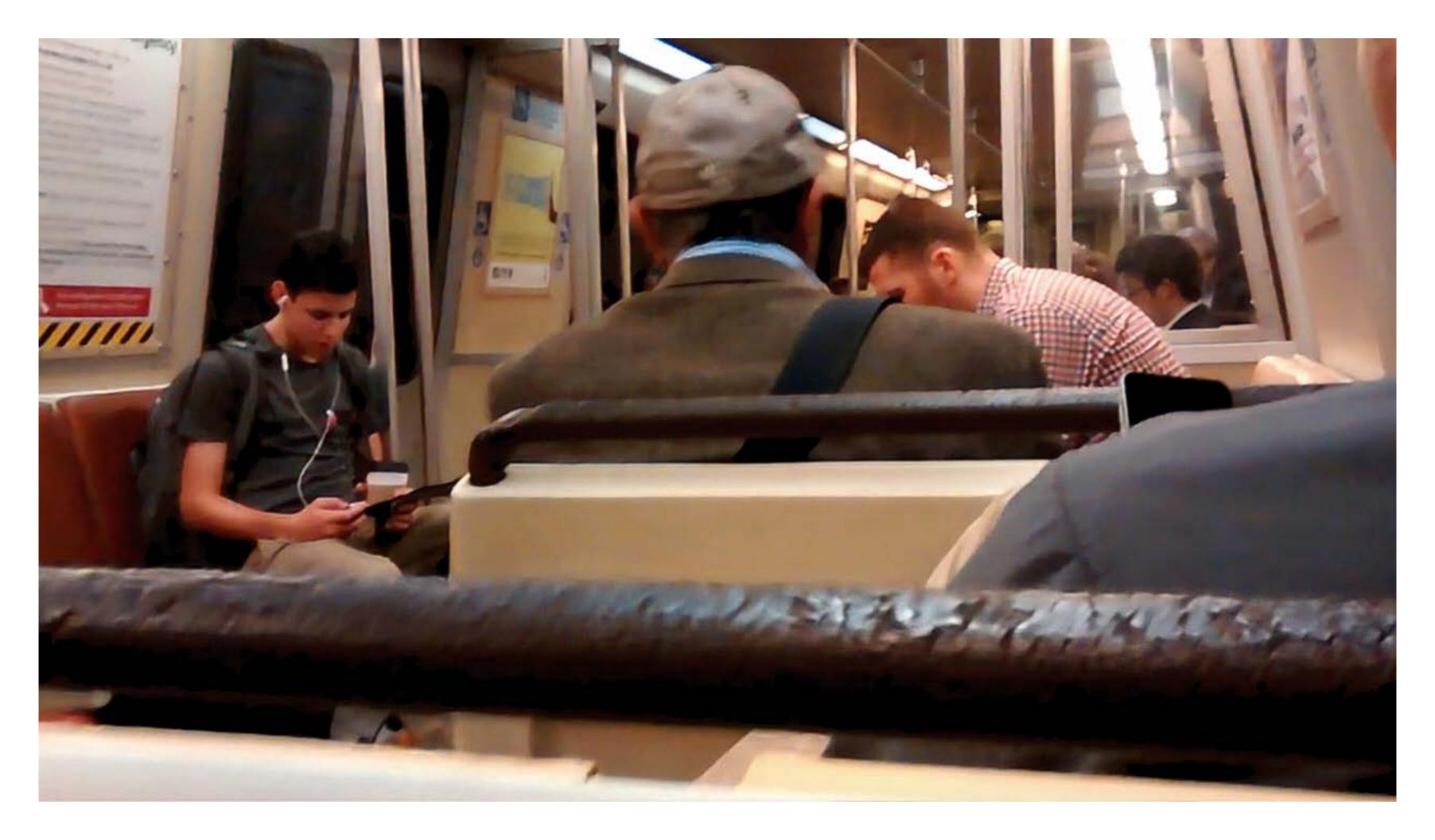




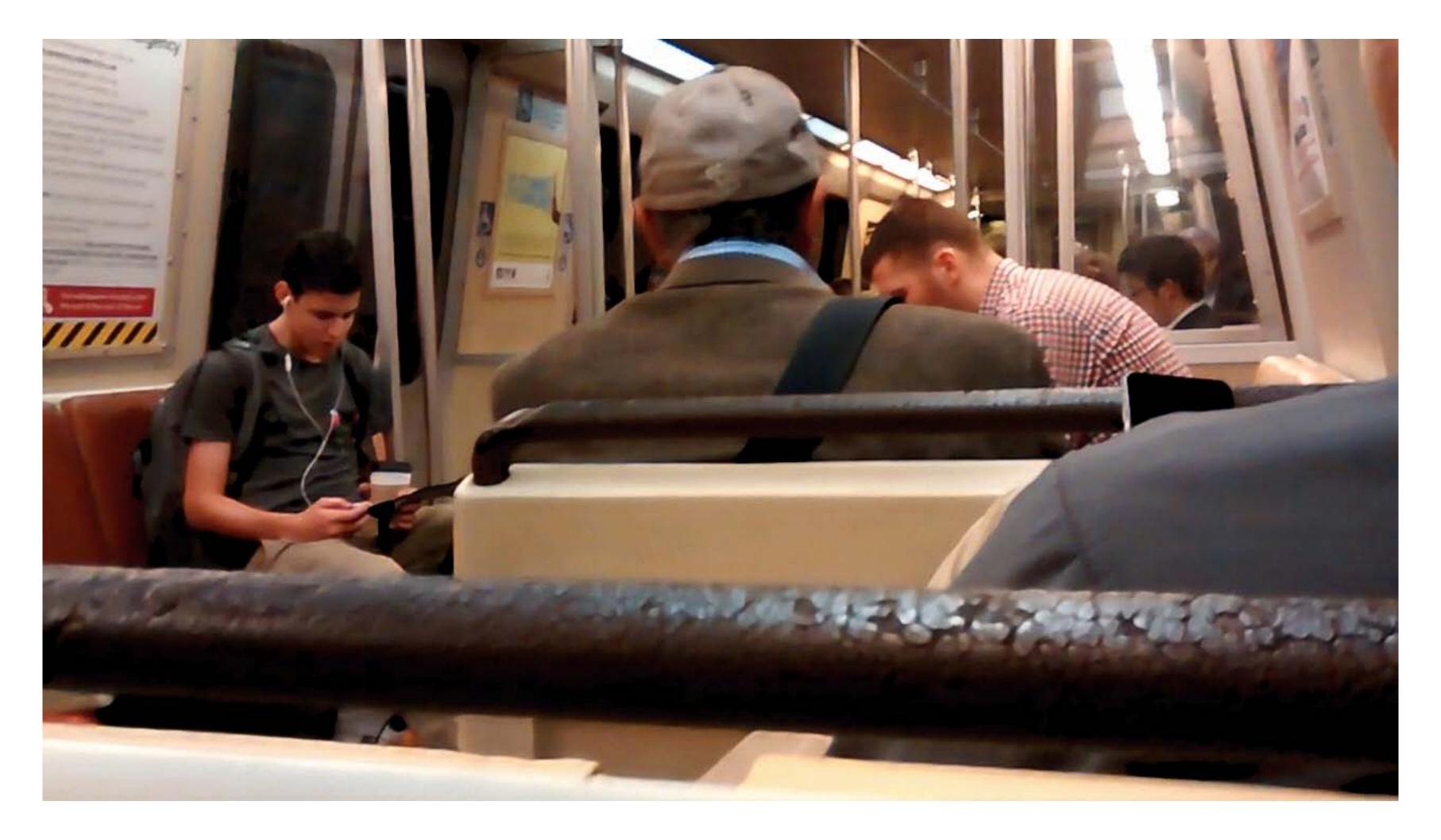
input



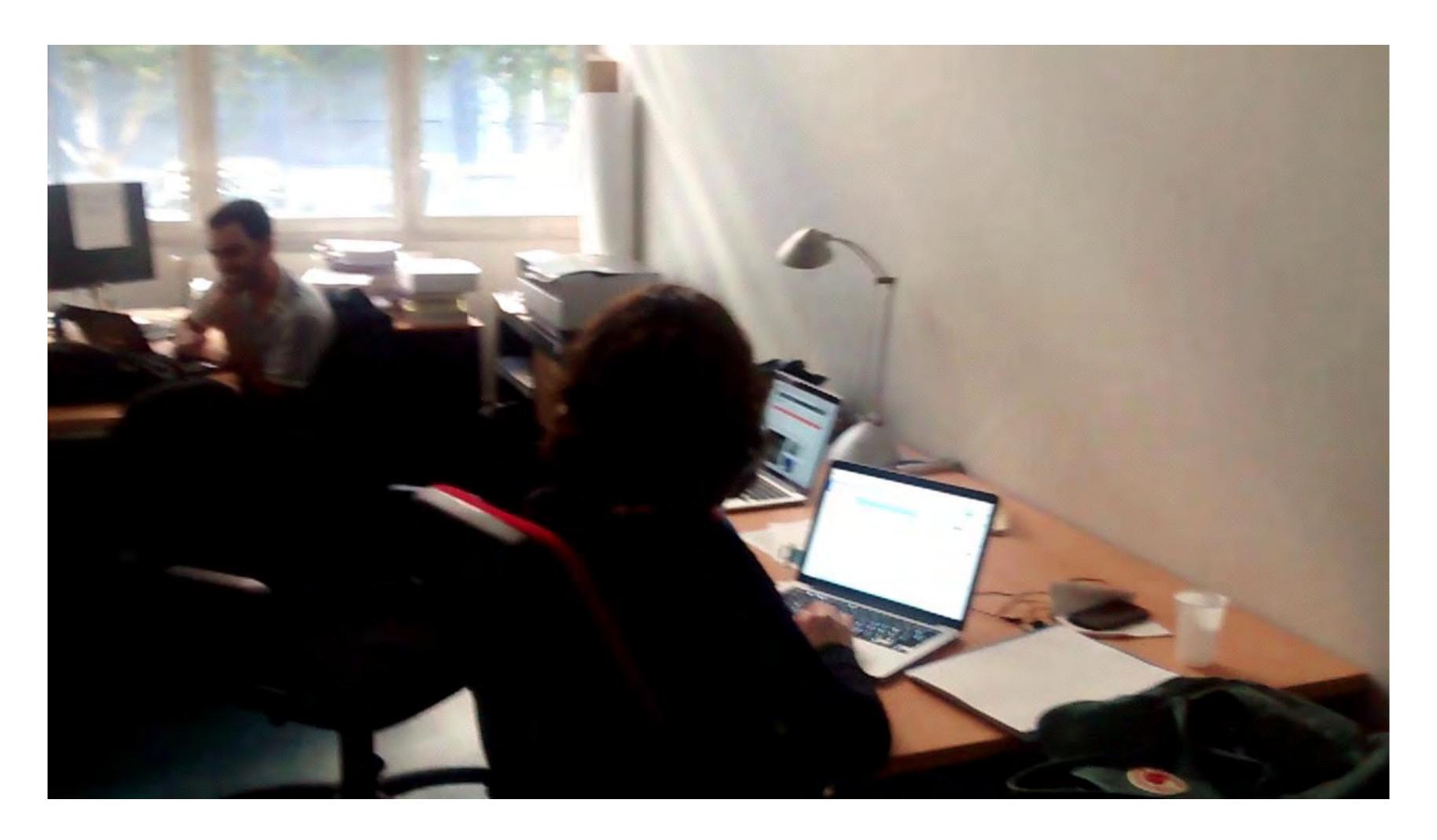
DBN+Noalign



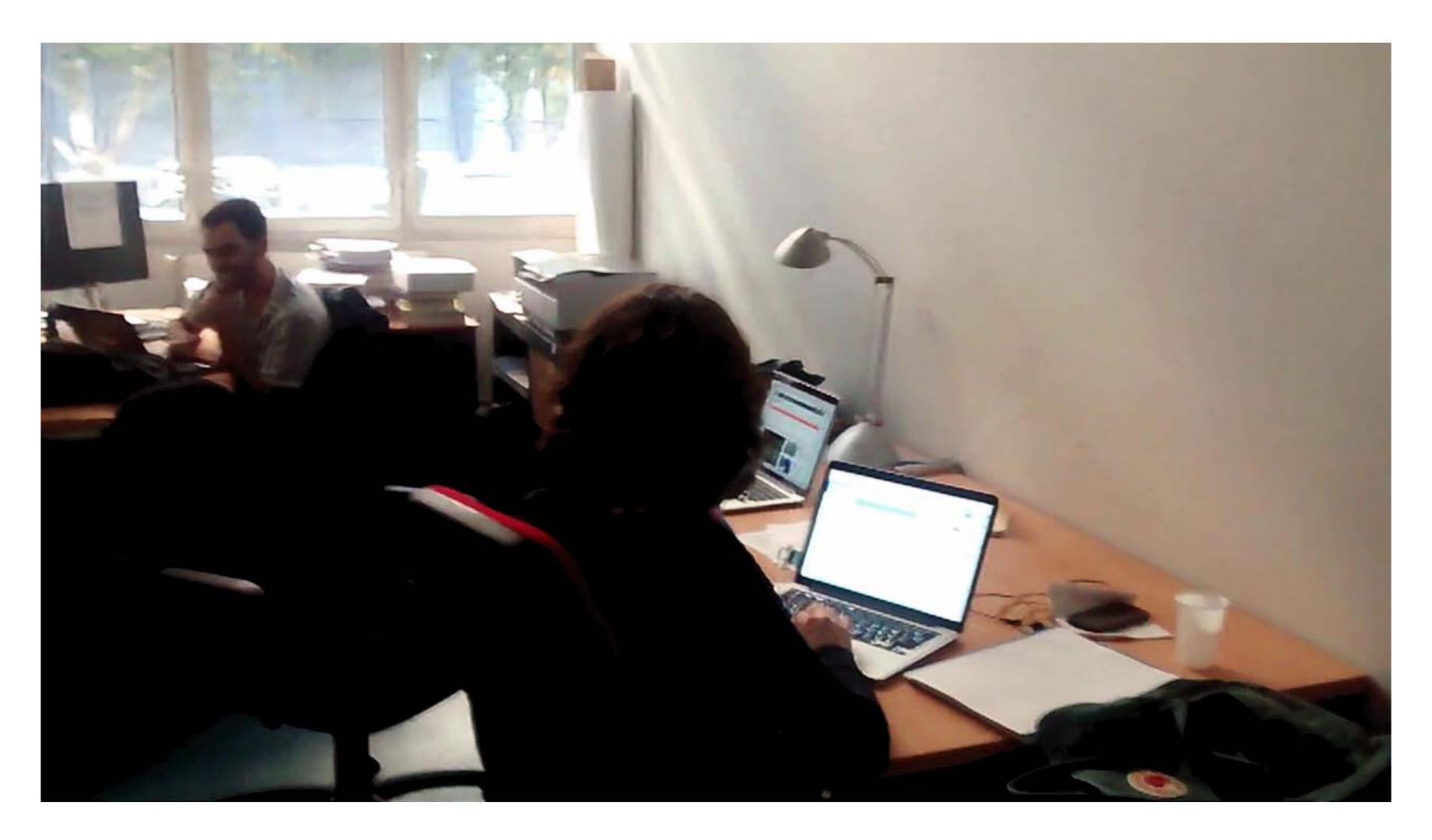
DBN+Homog



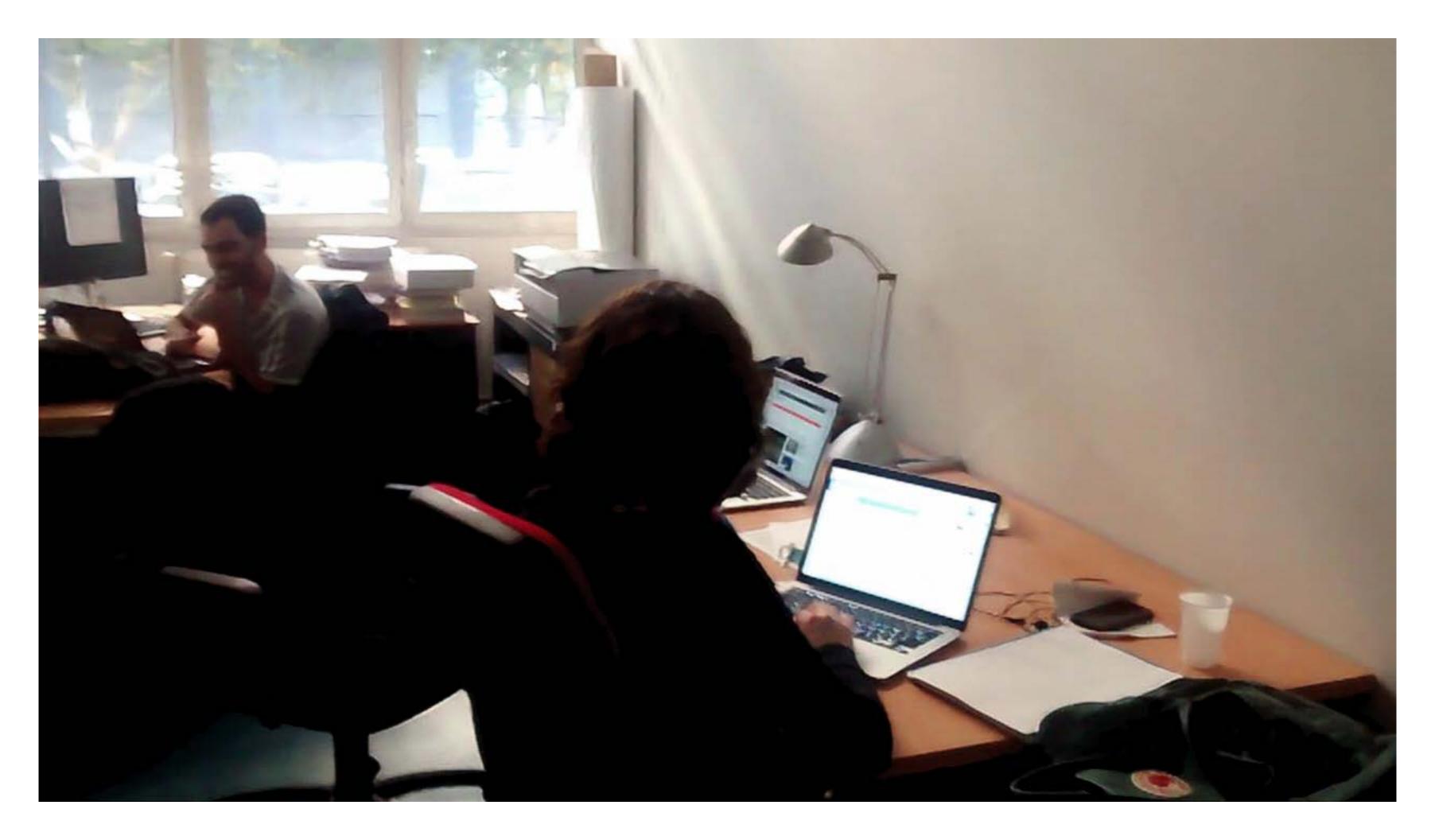
DBN+Flow



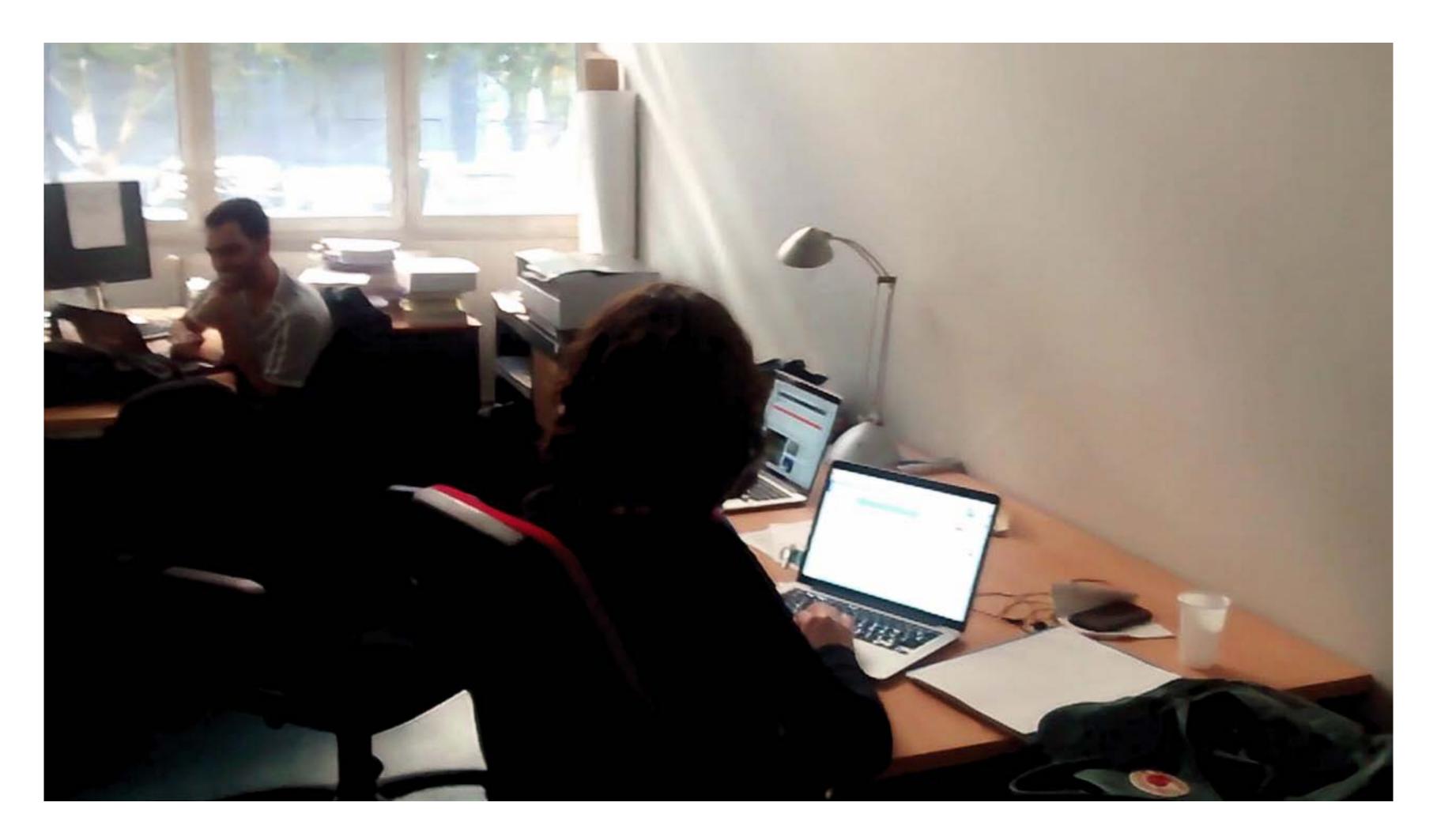
input



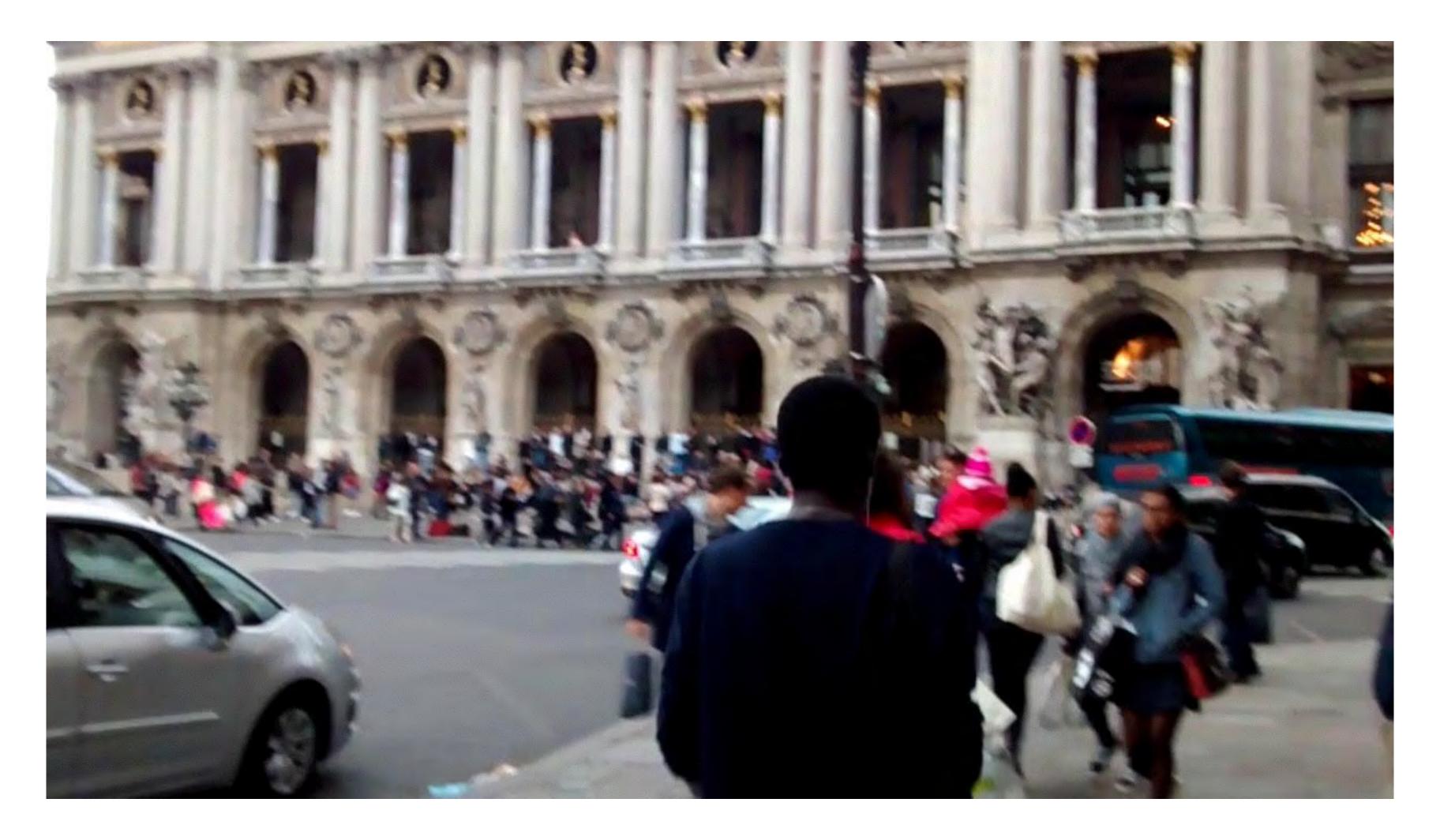
DBN+Noalign



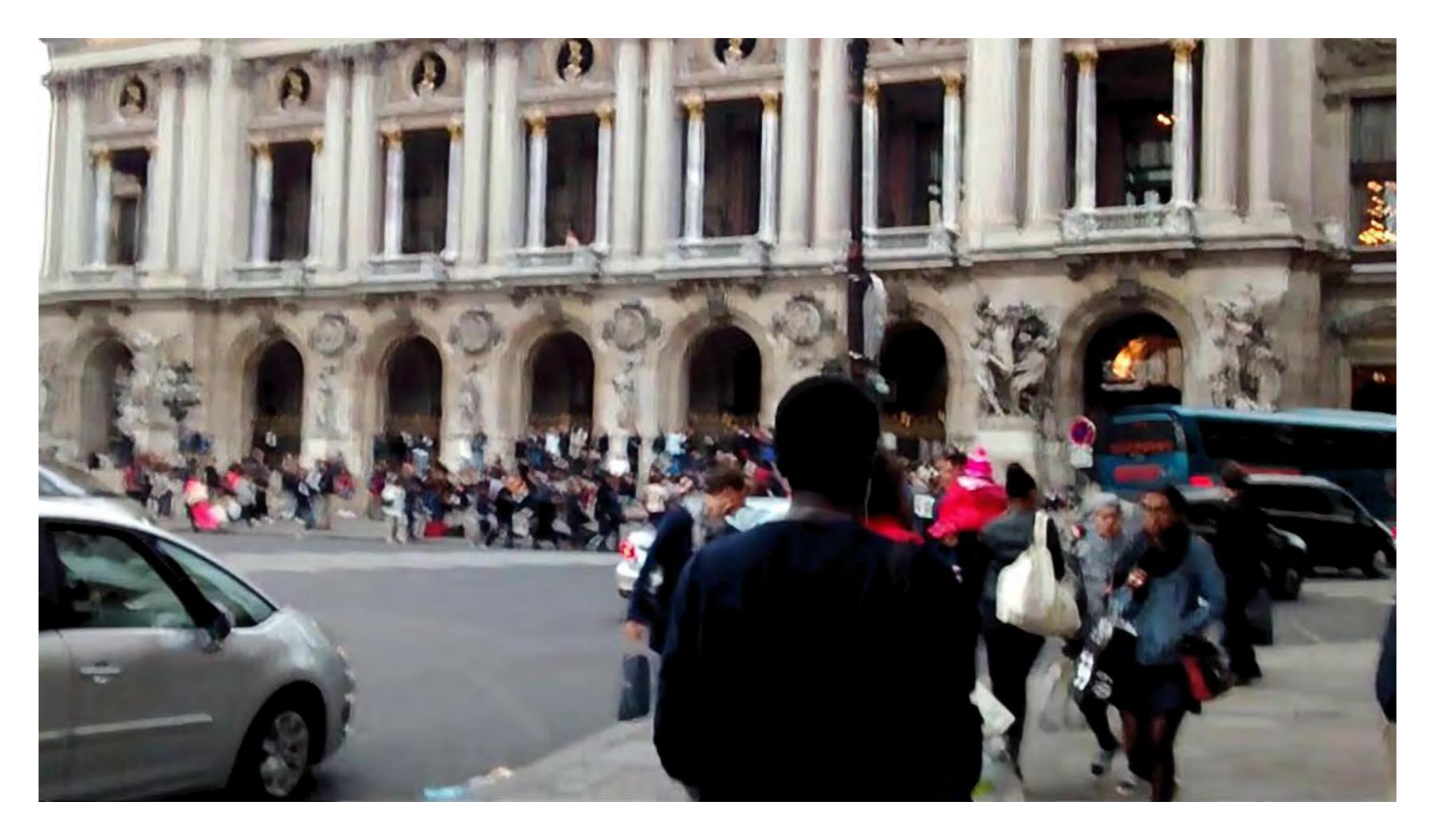
DBN+Homog



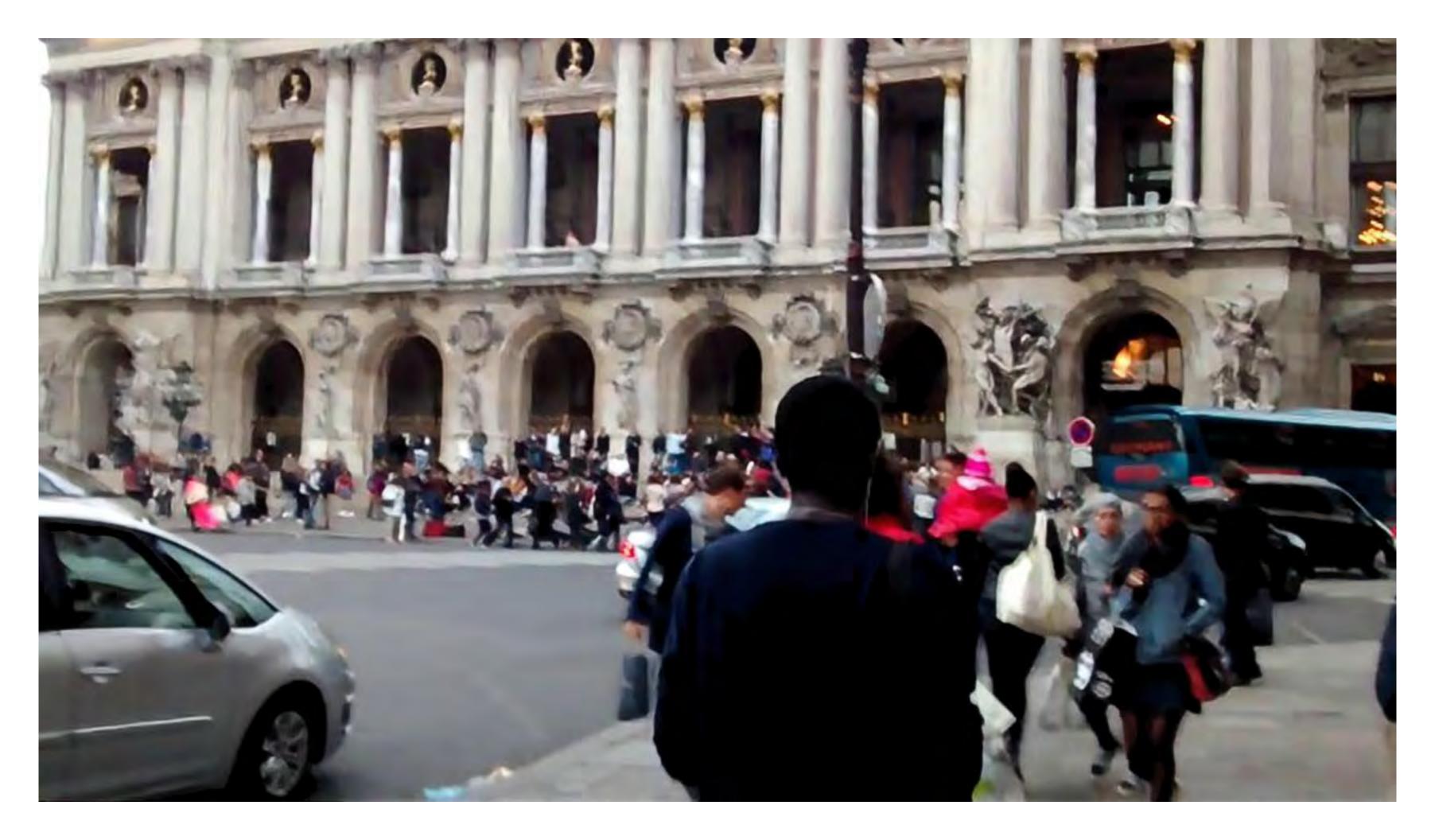
DBN+Flow



input



DBN+Noalign



DBN+Homog



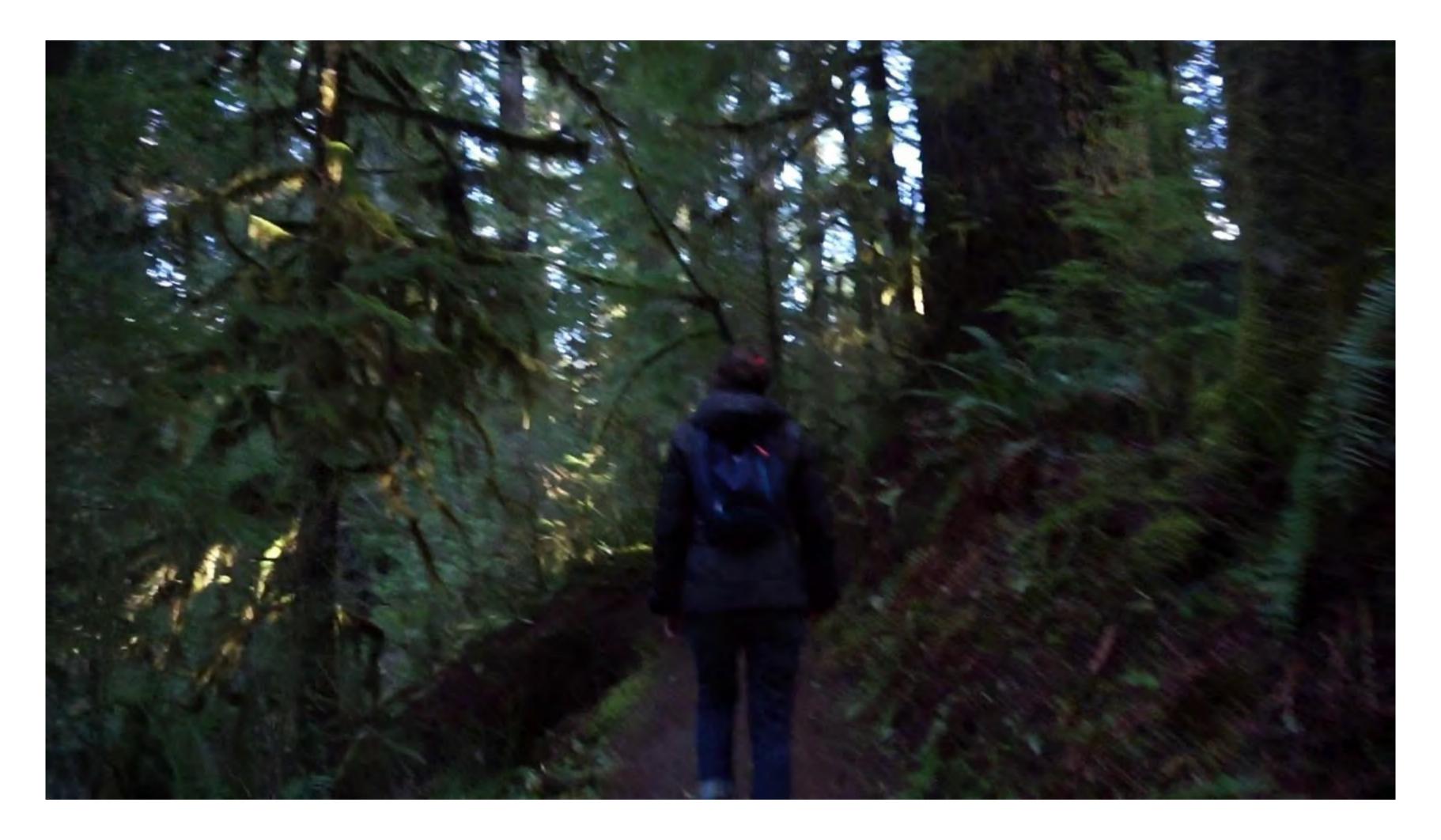
DBN+Flow



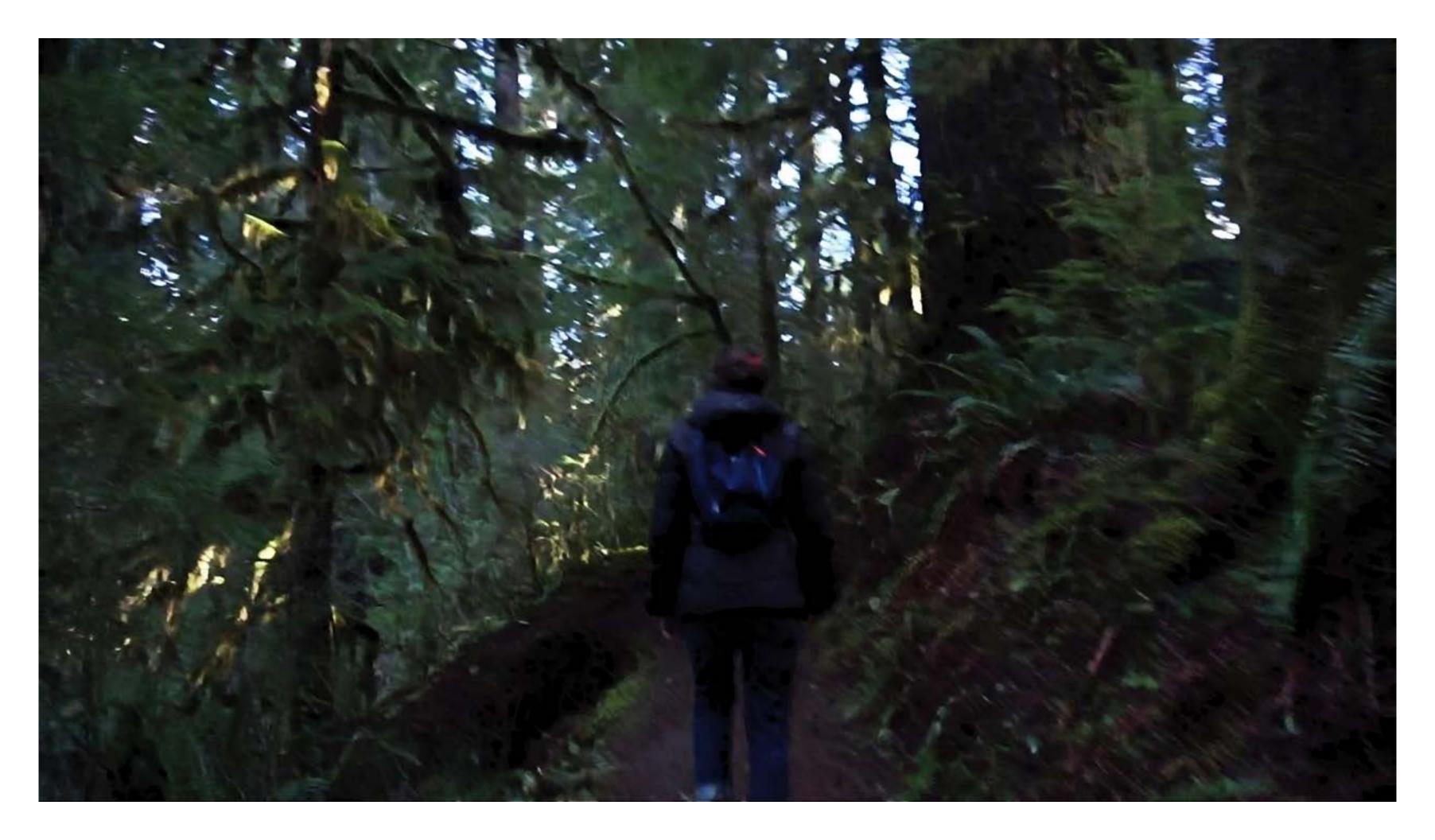




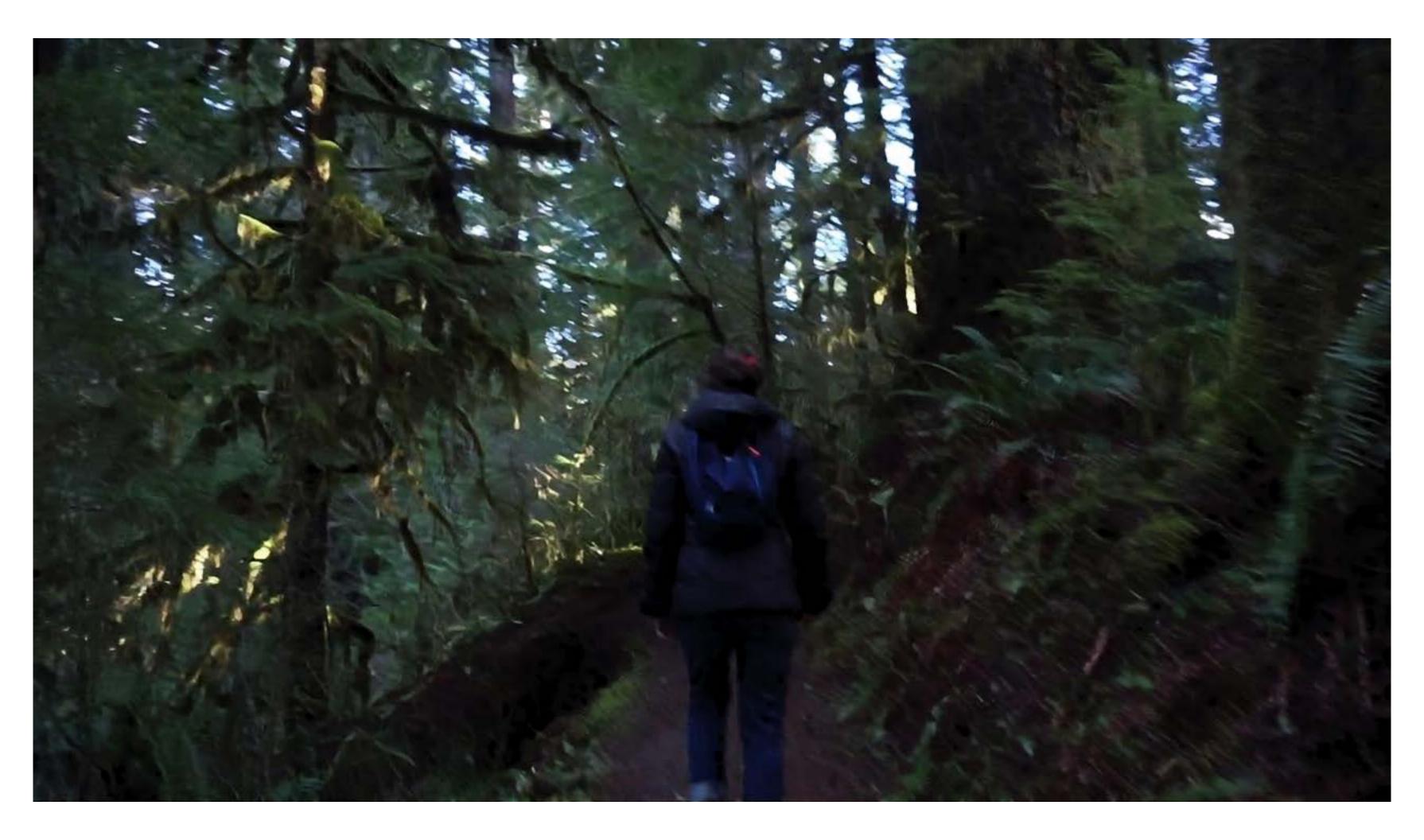




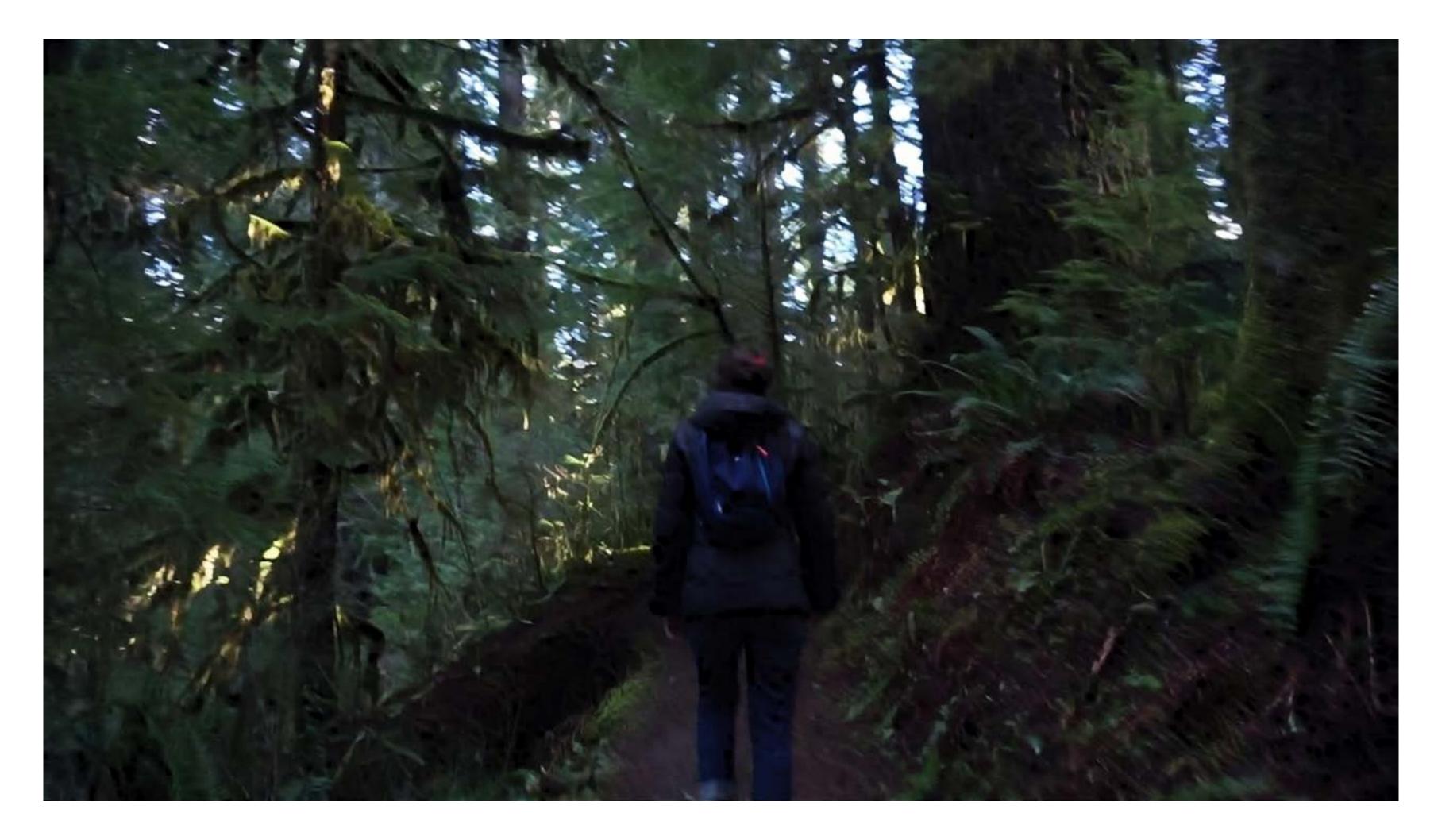
input



DBN+Noalign



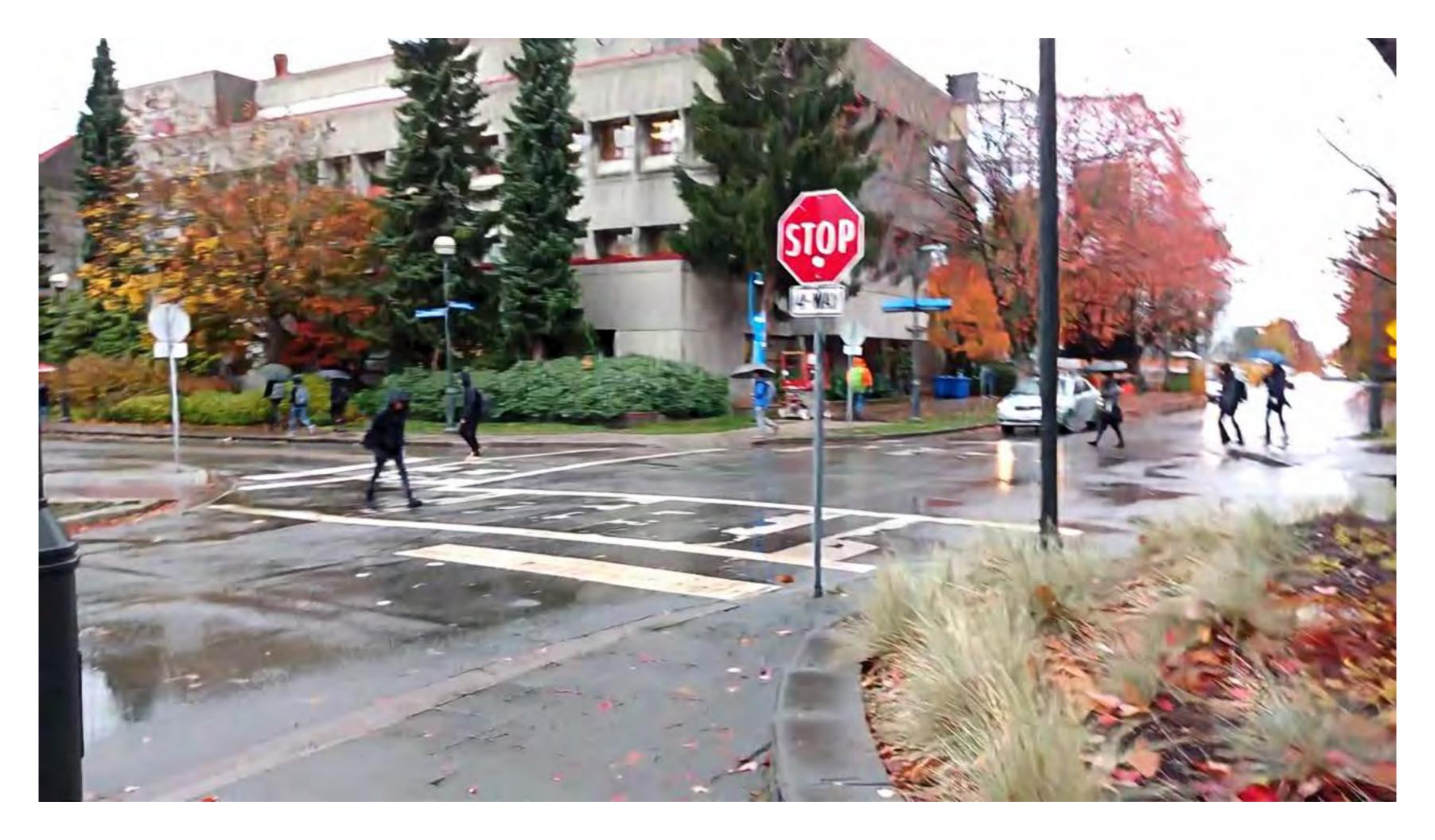
DBN+Homog



DBN+Flow



input



DBN+Noalign



DBN+Homog



DBN+Flow

Analysis on DeBlurNet

Depth of DBN

 Here we compare DBN with a slightly shallower version, where D3 to F4_3 are removed. The bottleneck layers in DBN greatly help with handling misalignment and aggregation.



DBN-Shallower+Flow



DBN+Flow

Late v.s. Early Fusion

 Here we compare different fusion strategies. While late fusion occasionally helps with challenging cases where DBN+Noalign fails, this improvement is not consistent. This might be related to the failure of artifacts/misalignments detections without the presence of reference frame.



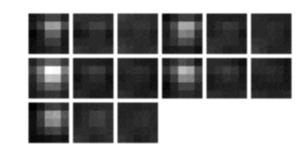
DBN-LateFusion+Noalign



DBN+Noalign

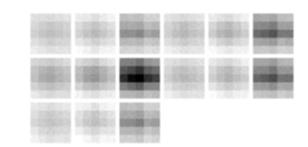
Visualization of Filters Learned

Here we visualize some filters learned from DBN, specifically at F0.
It can be observed that DBN not only learns to locate the
corresponding color channels in the input stack, but is also able to
learn to extract edges of different orientations, and to locate the
warping artifacts.



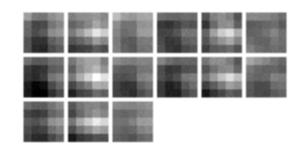


Filters and responses from F0 at DBN+Flow, that appear to find color channel correspondences.



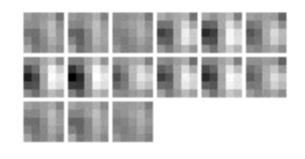


Filters and responses from F0 at DBN+Flow, that appear to find color channel correspondences.

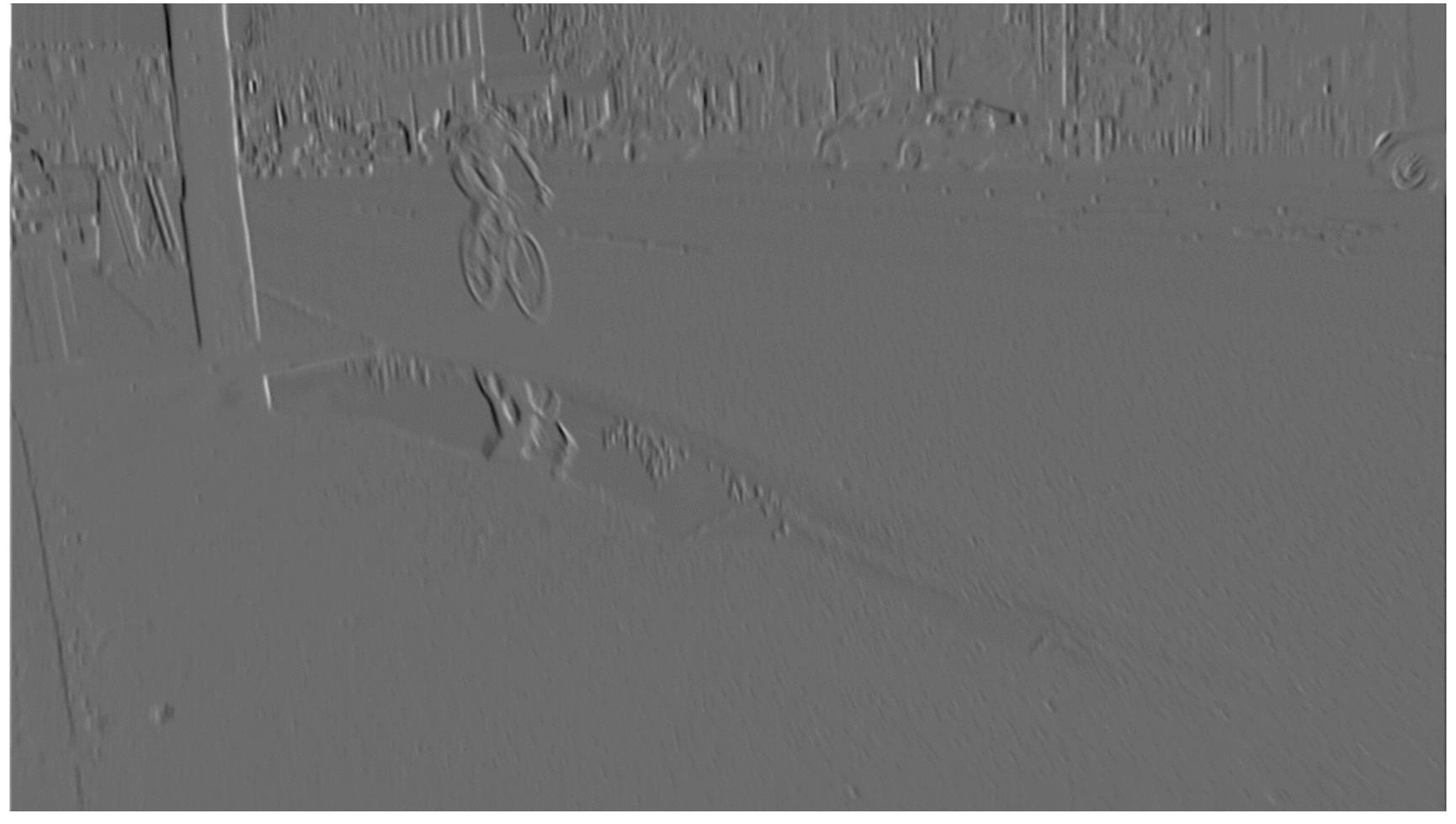




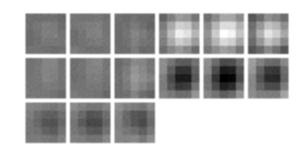
Filters and responses from F0 at DBN+Flow, that appear to find edges.

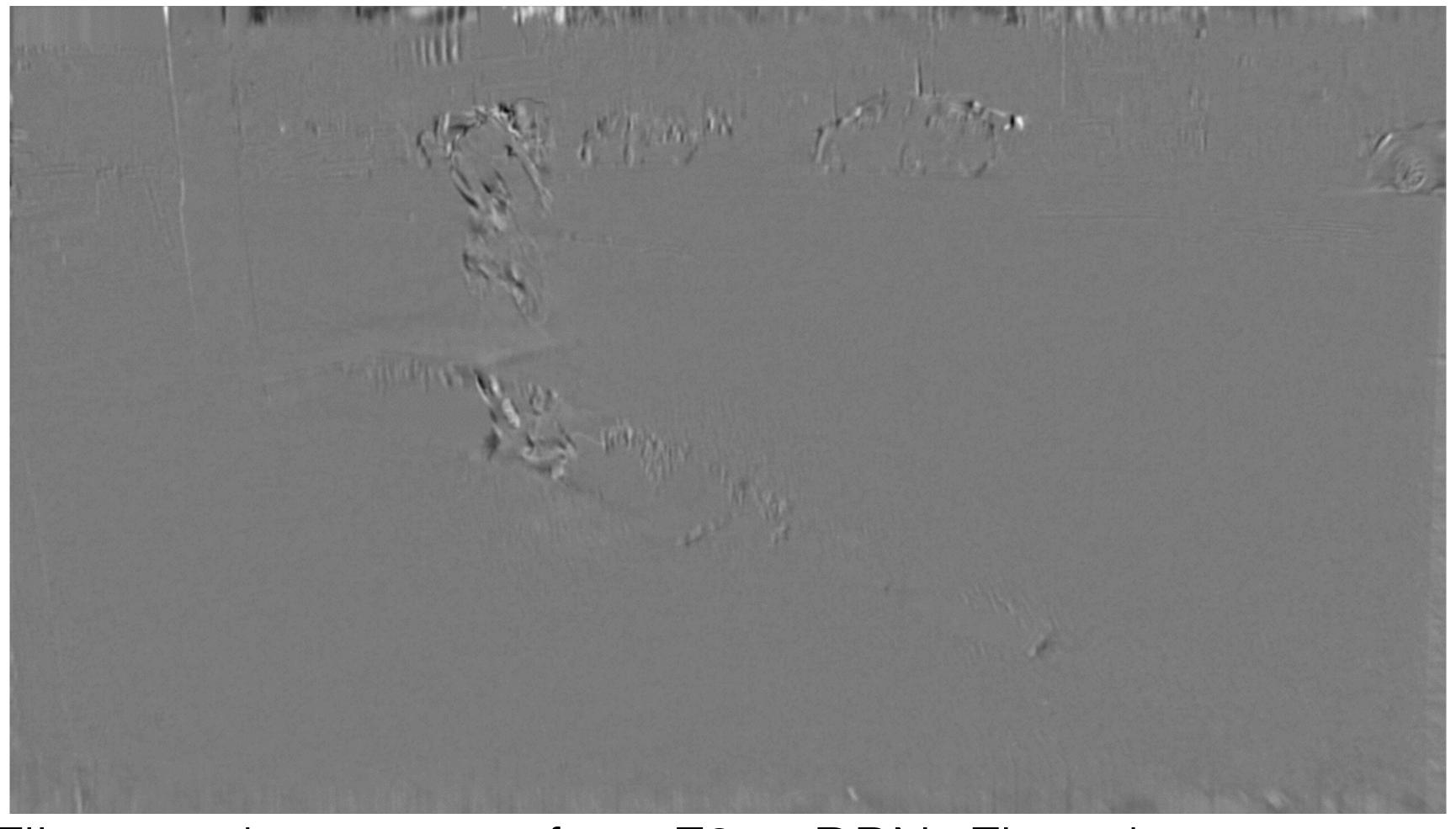


Edges

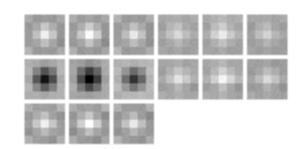


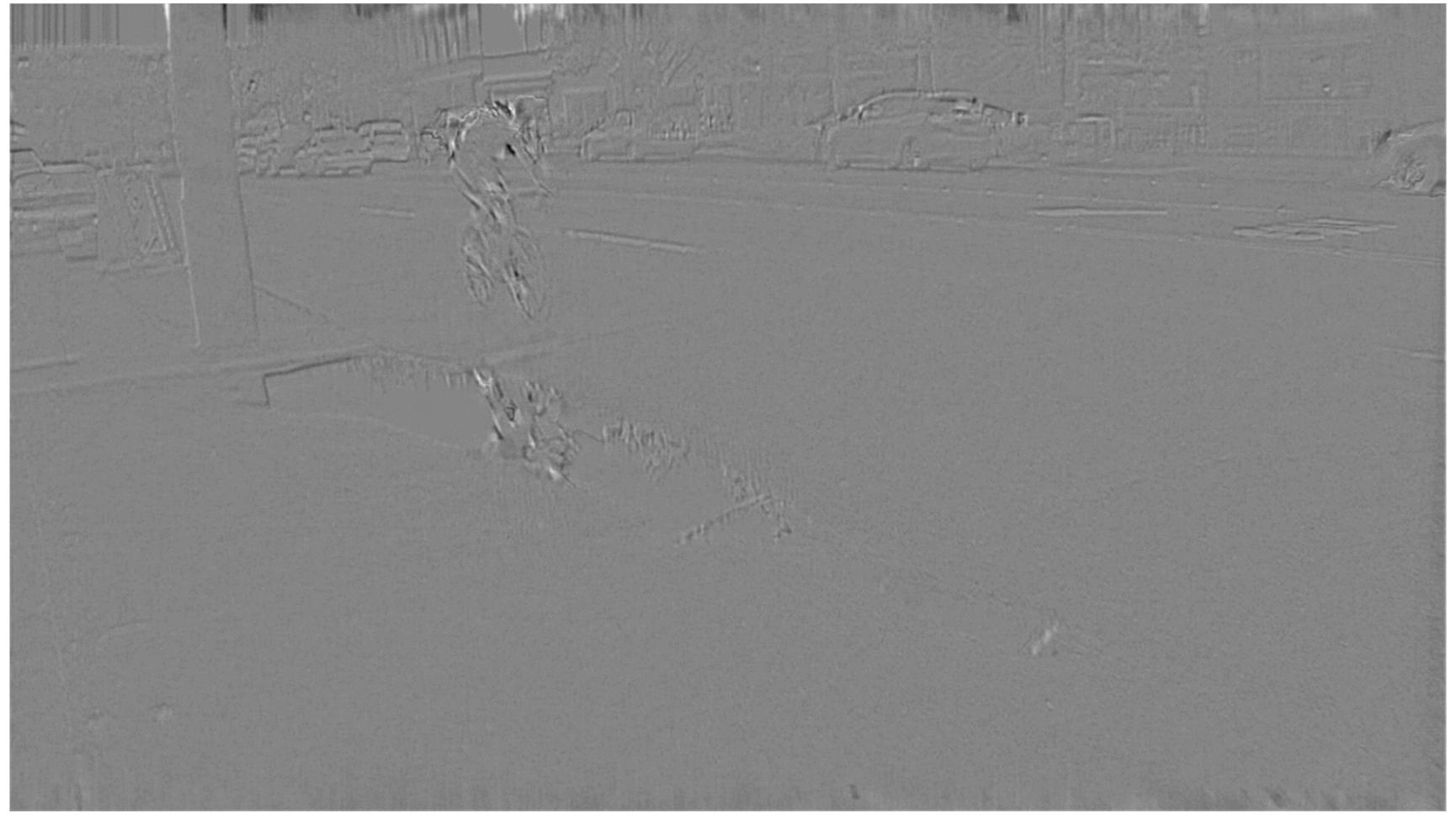
Filters and responses from F0 at DBN+Flow, that appear to find edges.





Filters and responses from F0 at DBN+Flow, that appear to detect warping artifacts.

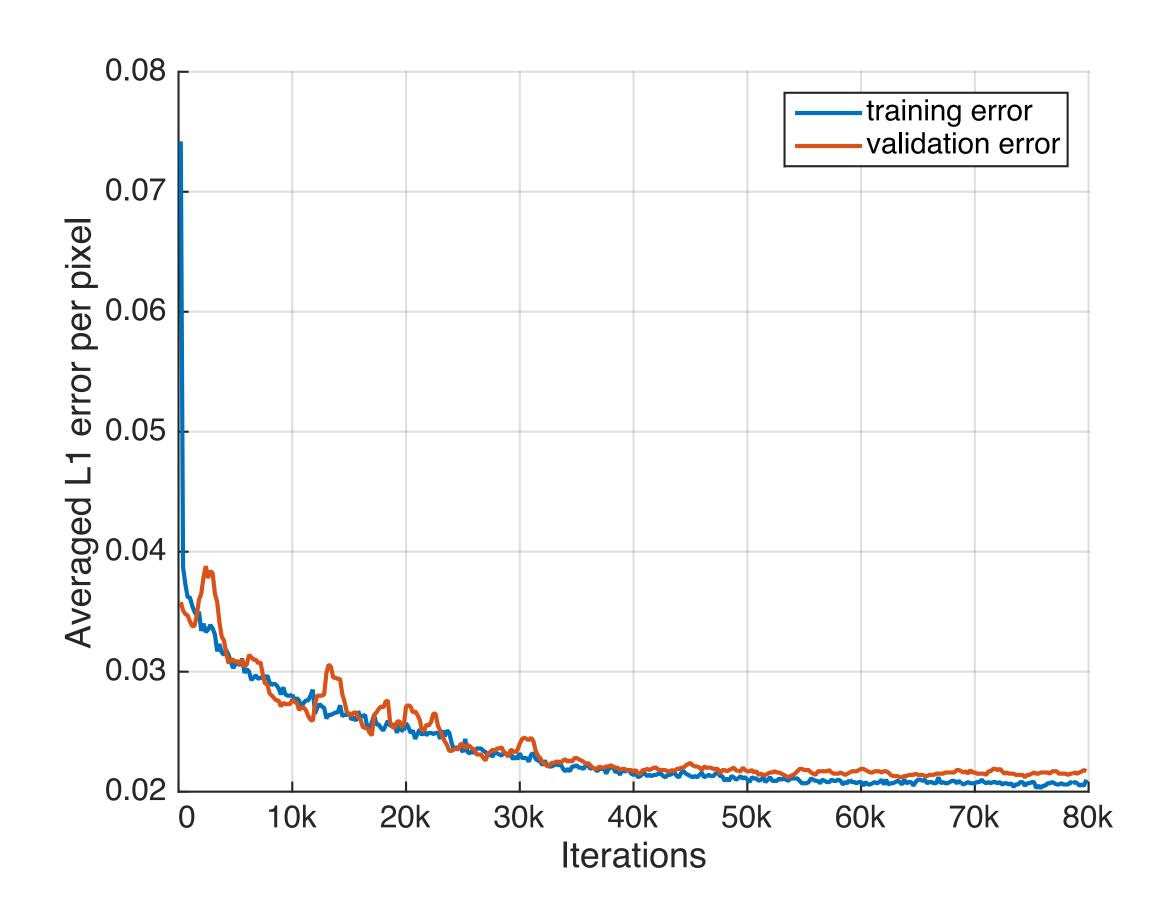




Filters and responses from F0 at DBN+Flow, that appear to detect warping artifacts.

Convergence

• A convergence plot is given here.



Convergence, DBN+Flow