

# Tracking the Untrackable: How to Track When Your Object Is Featureless

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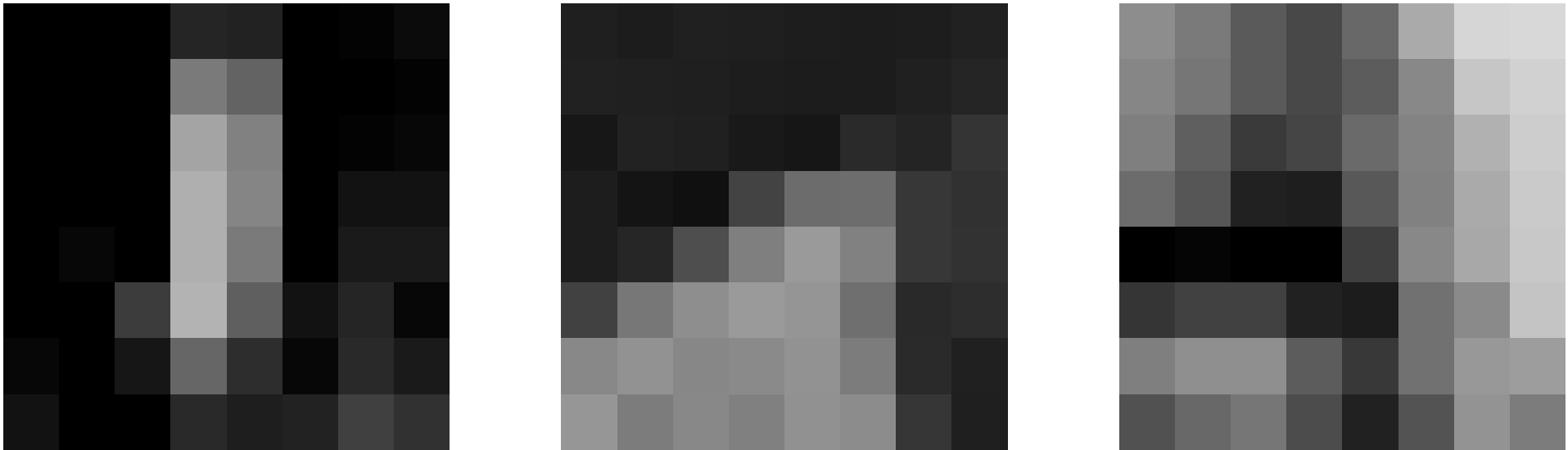
<http://cmp.felk.cvut.cz/>

<sup>2</sup>Centre for Vision, Speech and Signal Processing, University of Surrey

<http://surrey.ac.uk/cvssp/>



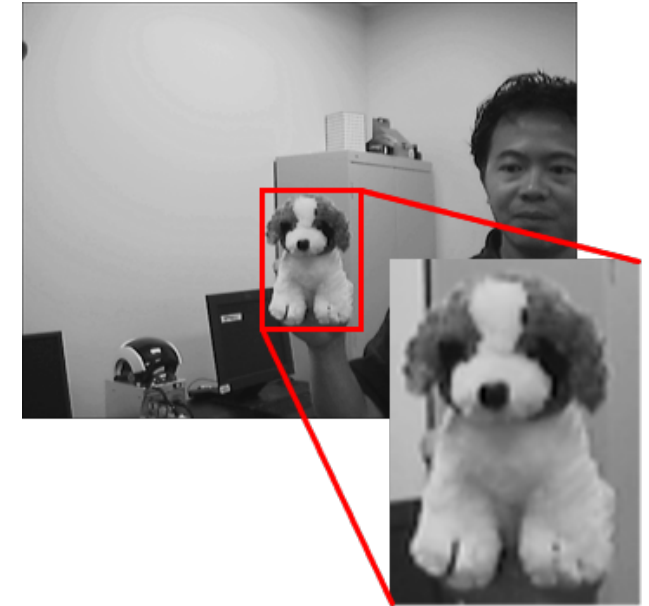
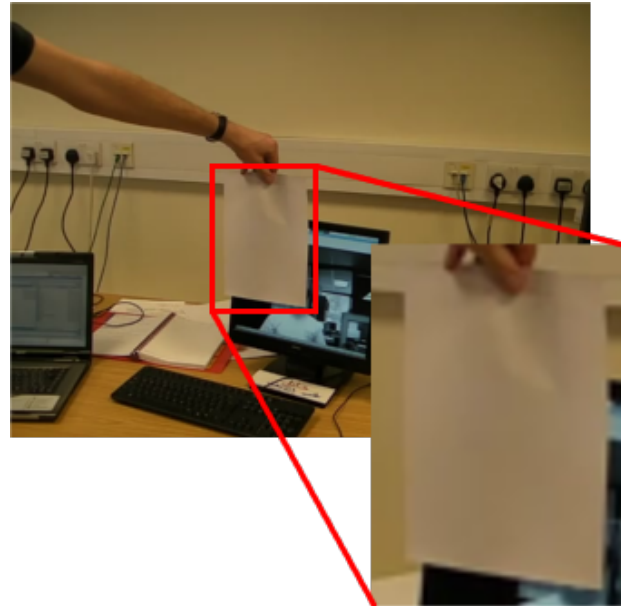
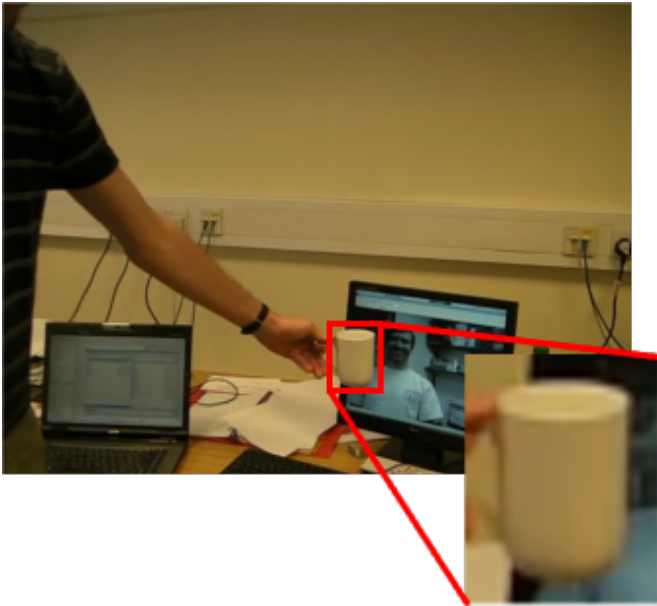
# Conventional Tracking



- ◆ well-localized features (blobs, corners,...)
- ◆ distinguishable from their neighbors

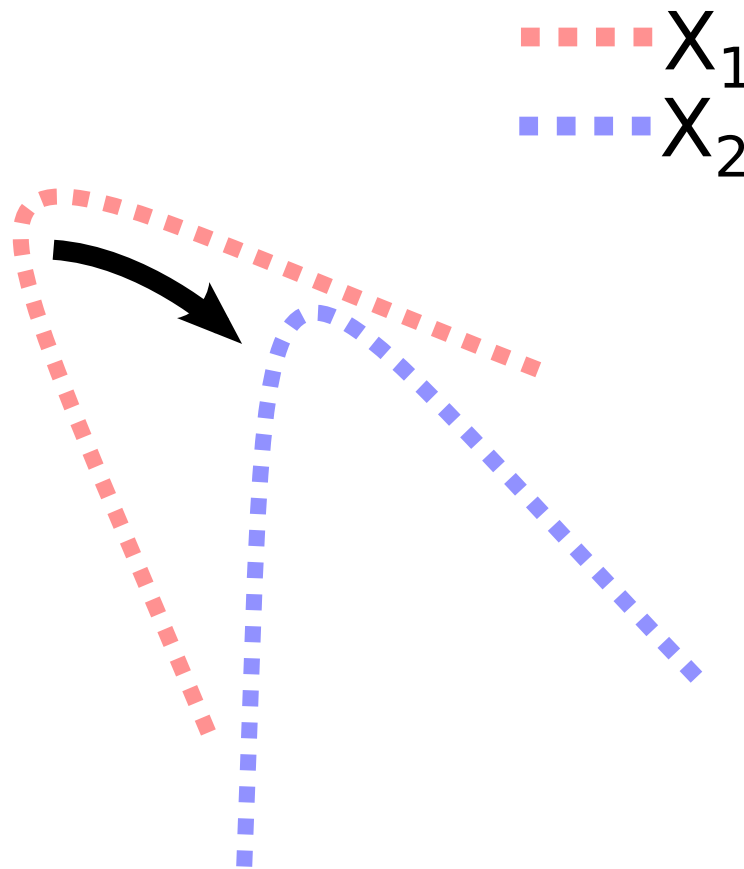
Moravec, H., Obstacle Avoidance and Navigation in the Real World by a Seeing Robot Rover, Tech Report CMU-RI-TR-3, Carnegie-Mellon University, Robotics Institute, 1980.

# Low-textured Objects?

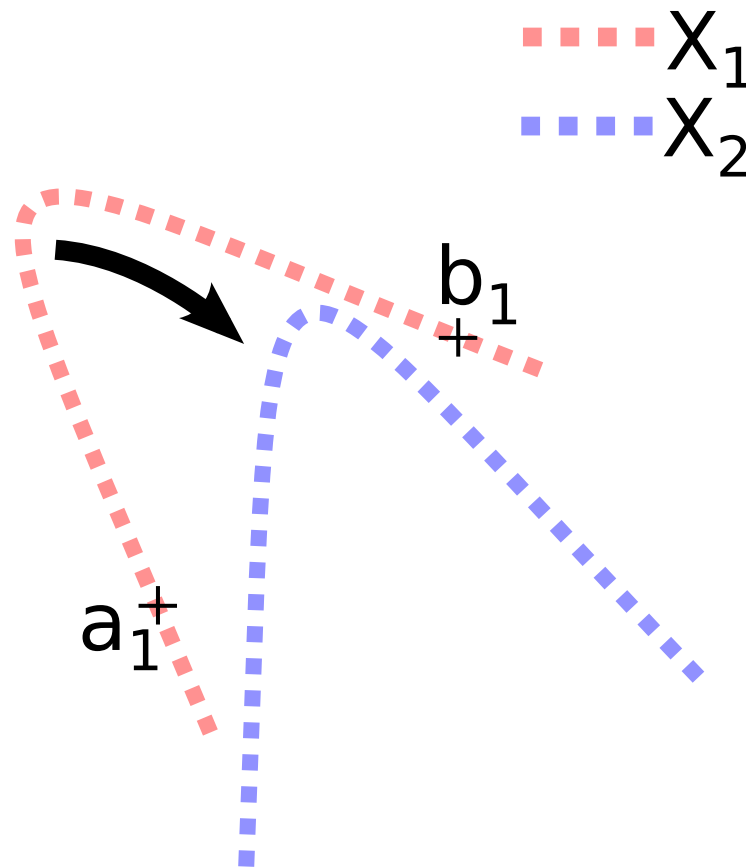


- ◆ some real-world objects have little of such features
- ◆ or these features lie on the object boundaries and are affected by the background

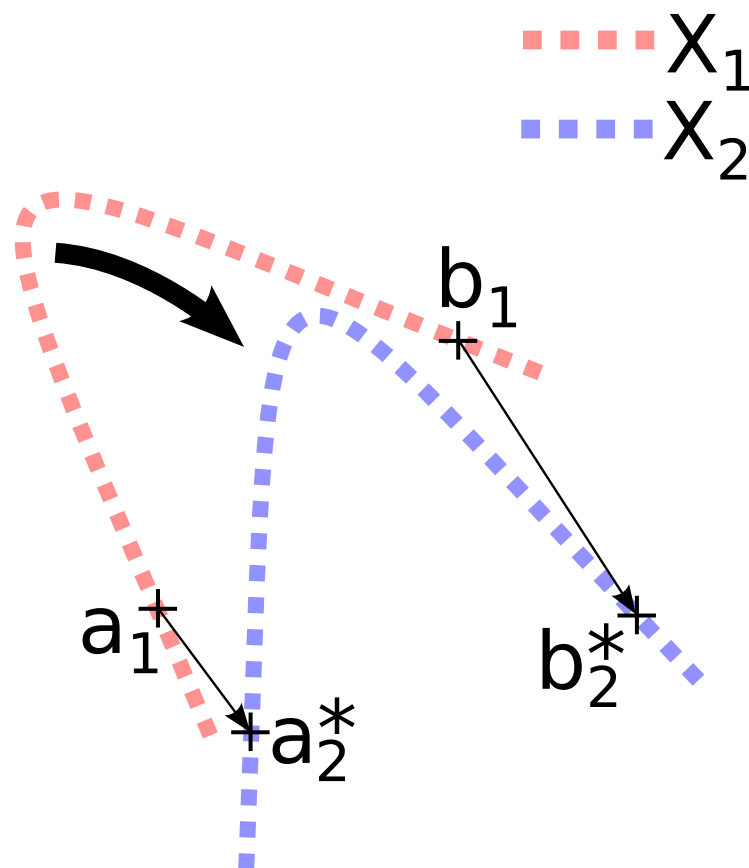
# The Aperture Problem



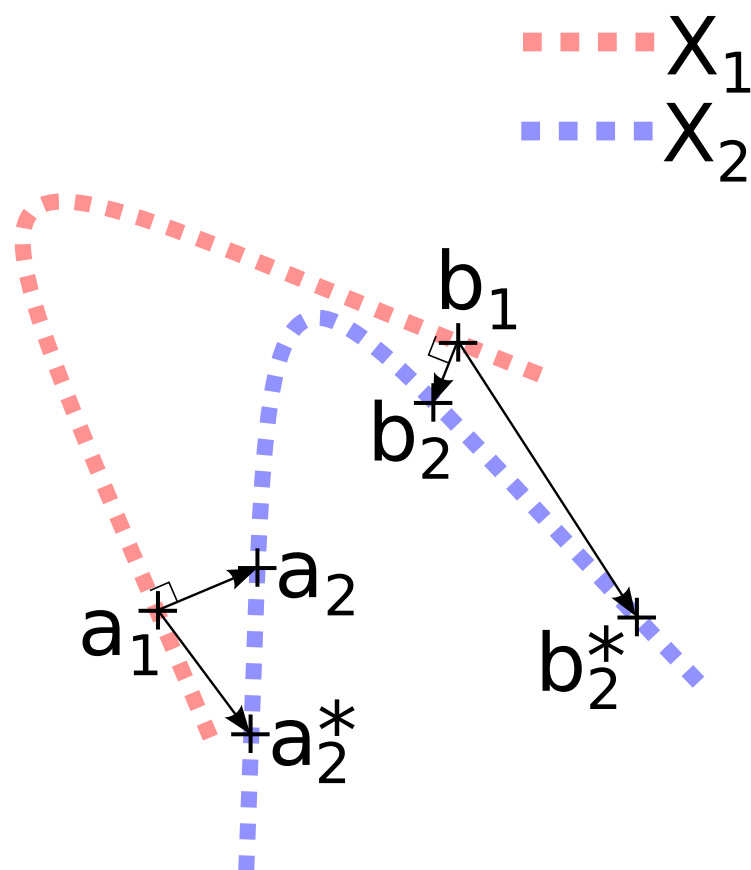
# The Aperture Problem



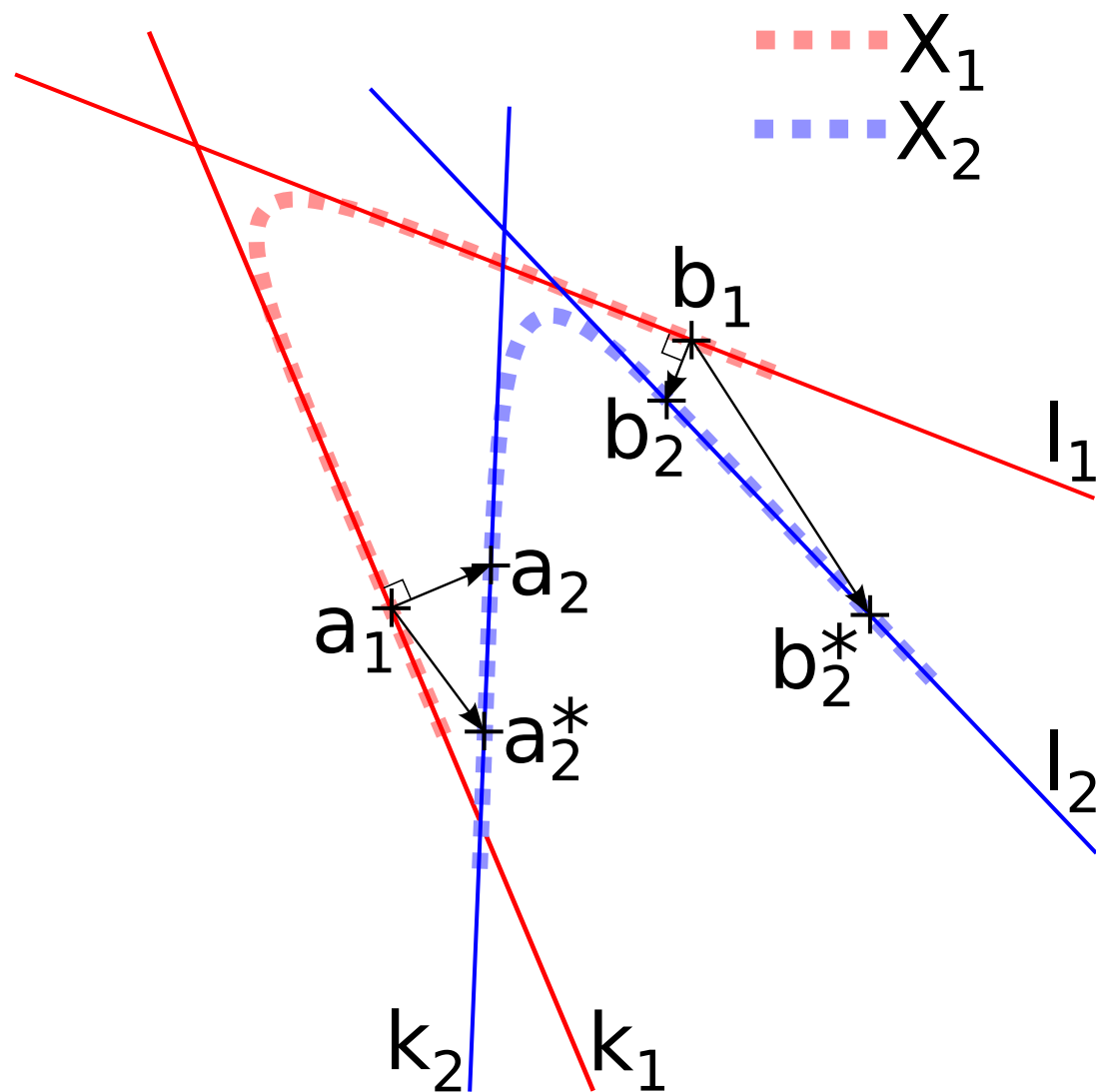
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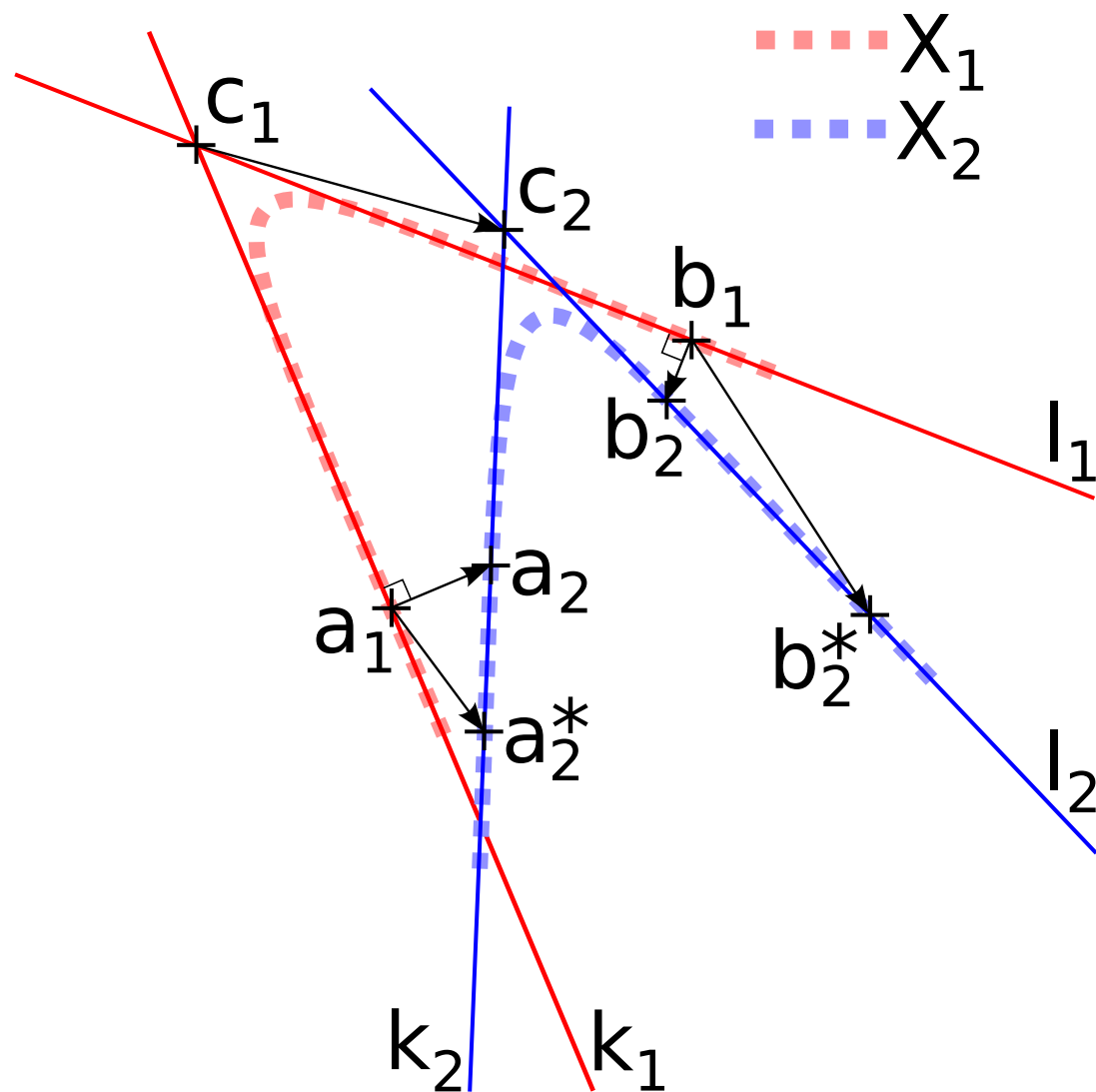


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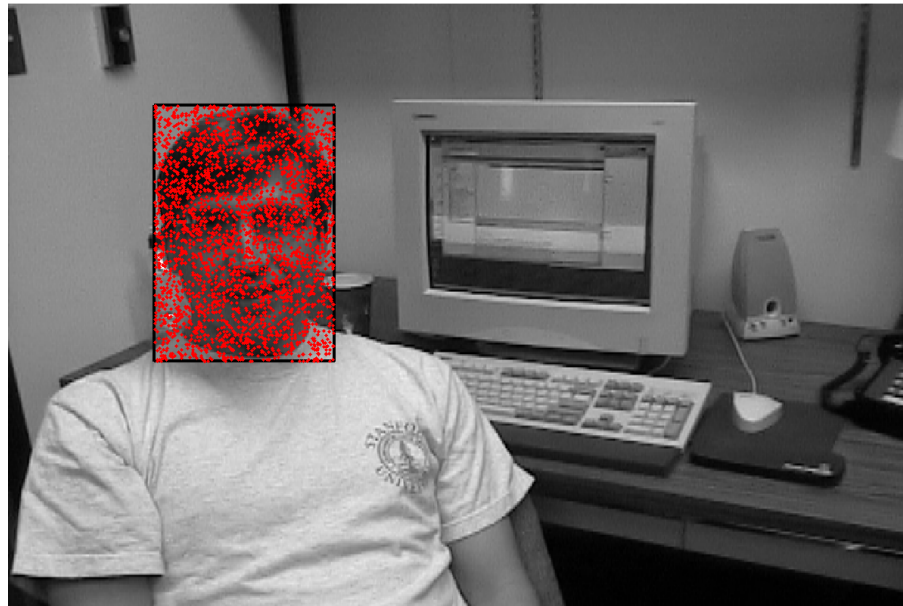
# Points and Lines

## Points

- ◆ edge points (edgels)
- ◆ randomly generated, then converged to the edges

## Lines

- ◆ tangents to the edges



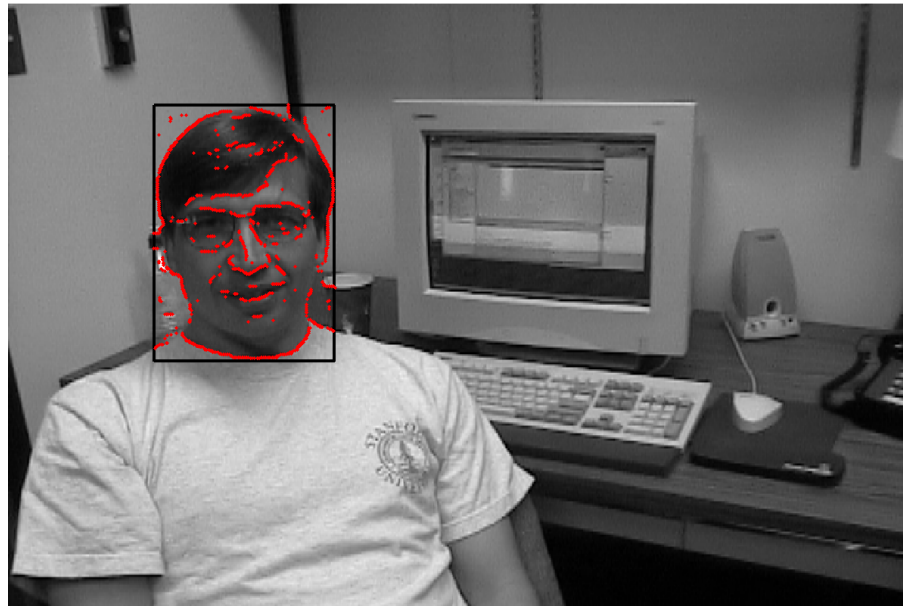
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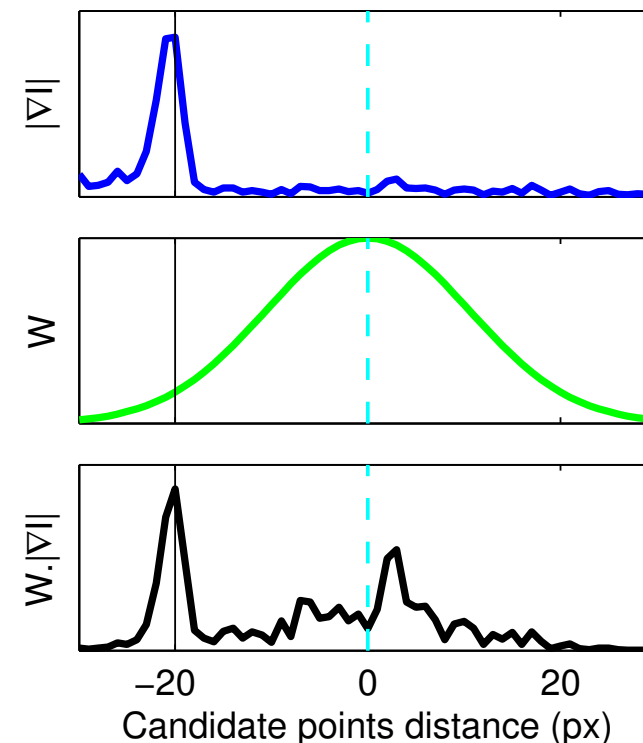
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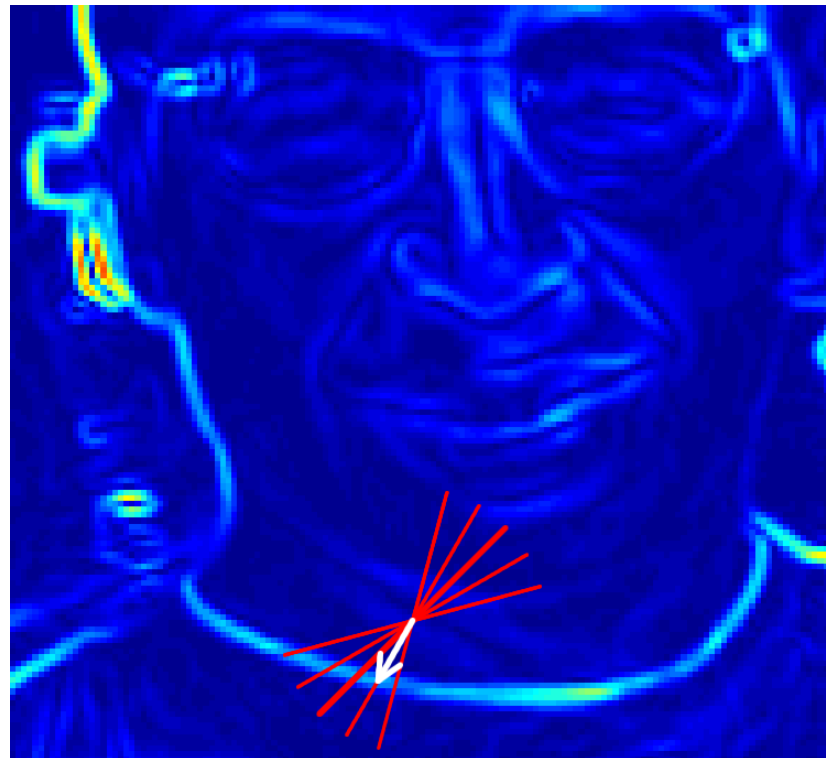
# Selecting Tangent Points

- ◆ used to find strong edges
- ◆ weak edges only when there are no strong ones near
- ◆ 1D search window in gradient direction
- ◆ weights: **magnitude of gradient** and **distance**



# Matching Tangents

- ◆ used to obtain correspondences
- ◆ candidates are tangent points at local maxima of gradients in multiple directions; correspondence verification criteria are: **gradient angle**, **appearance similarity** and **distance**.



# Line Correspondences

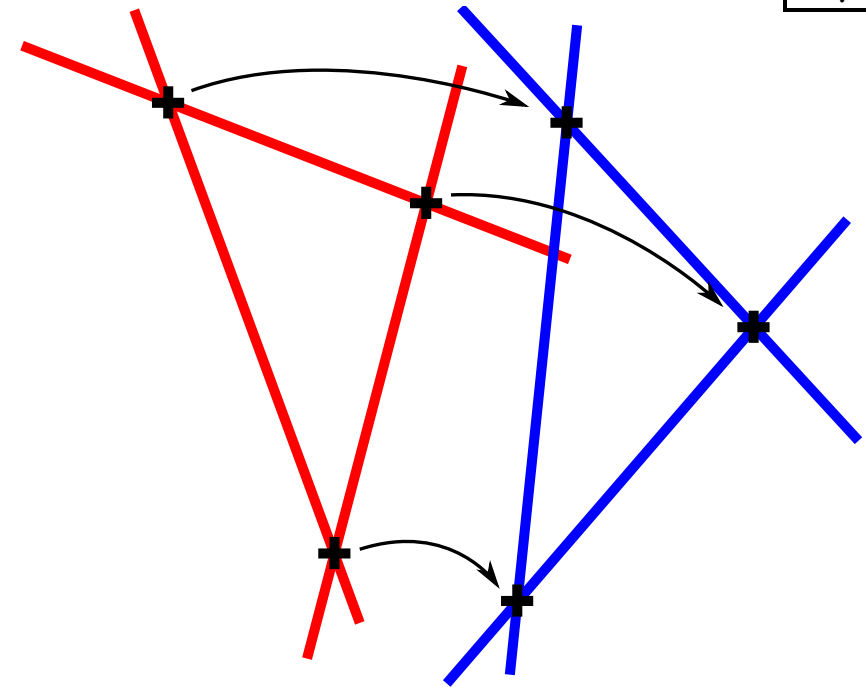
## Estimation of transformation

- ◆ 2 constraints from one correspondence
- ◆ however, 2 correspondences are not enough (scale ambiguity)
- ◆ at least 3 correspondences for a **similarity** transformation (4 DoF), either directly or by intersection correspondences

# RANSAC – RANdom SAmple Consensus

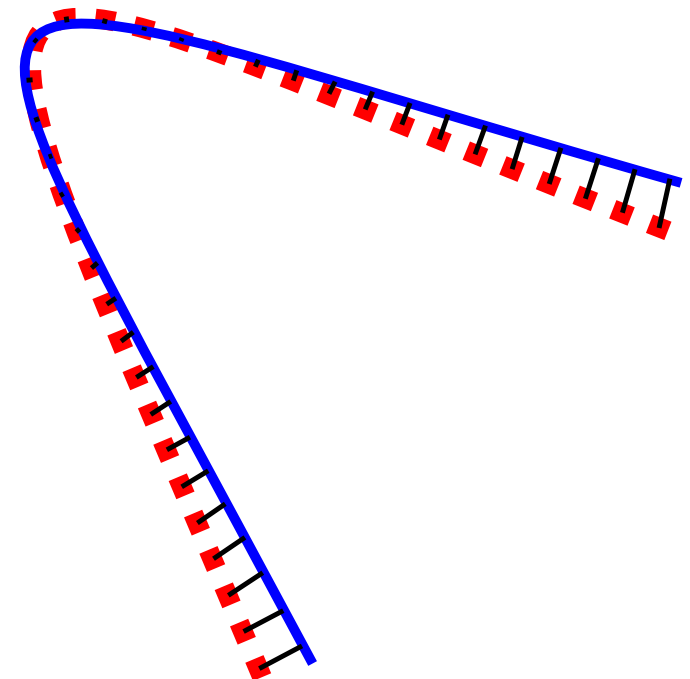
## Random Sample

- ◆ minimal sample:  
3 line correspondences
- ◆ transformation by point  
correspondences of intersections



## Consensus

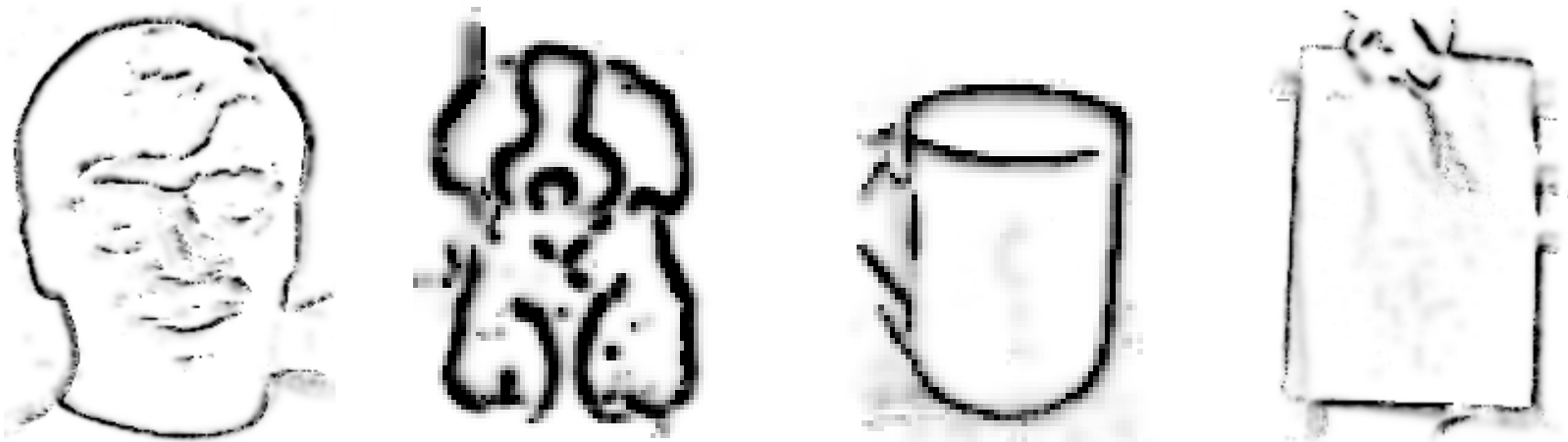
- ◆ inlier ratio insufficient  
indicator of quality
- ◆ used together with  
image **evidence** of points  
(Chamfer distance)





# Online Learning Points Reliability

- ◆ remembering positions of edge points whose tangents are often transformed to edges  
 ⇒ field of observed point qualities
- ◆ gives an estimate of object structure
- ◆ fitting to points decreases drift



# Model

- ◆ Encapsulates whole state of frame-to-frame tracking:
  - geometric information:  
position of tracker, points, lines,
  - field of observed point qualities.
- ◆ Initial model: user-given tracked bounding box, generated points.
- ◆ Online update at every frame.

# Using Multiple Models

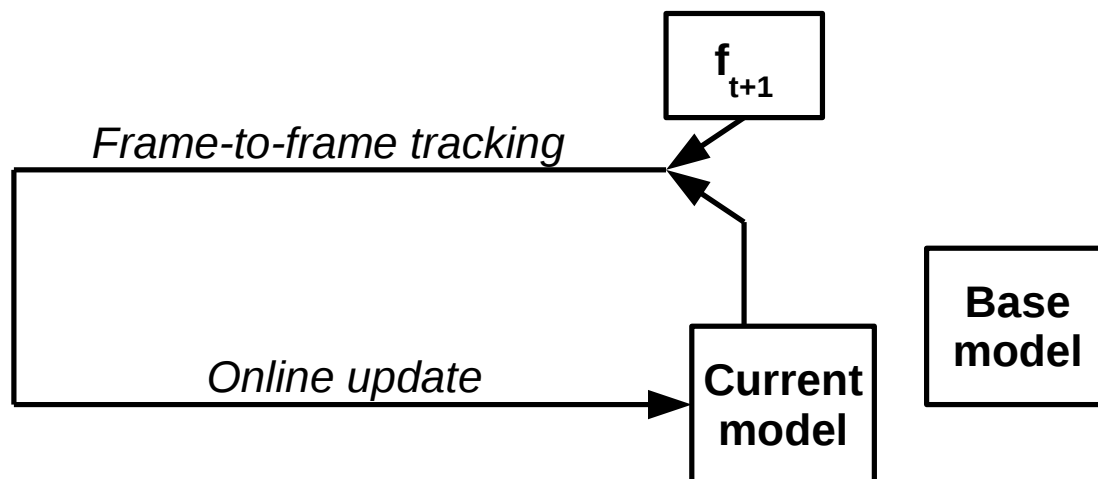
Current model = state of the tracker.

Current  
model



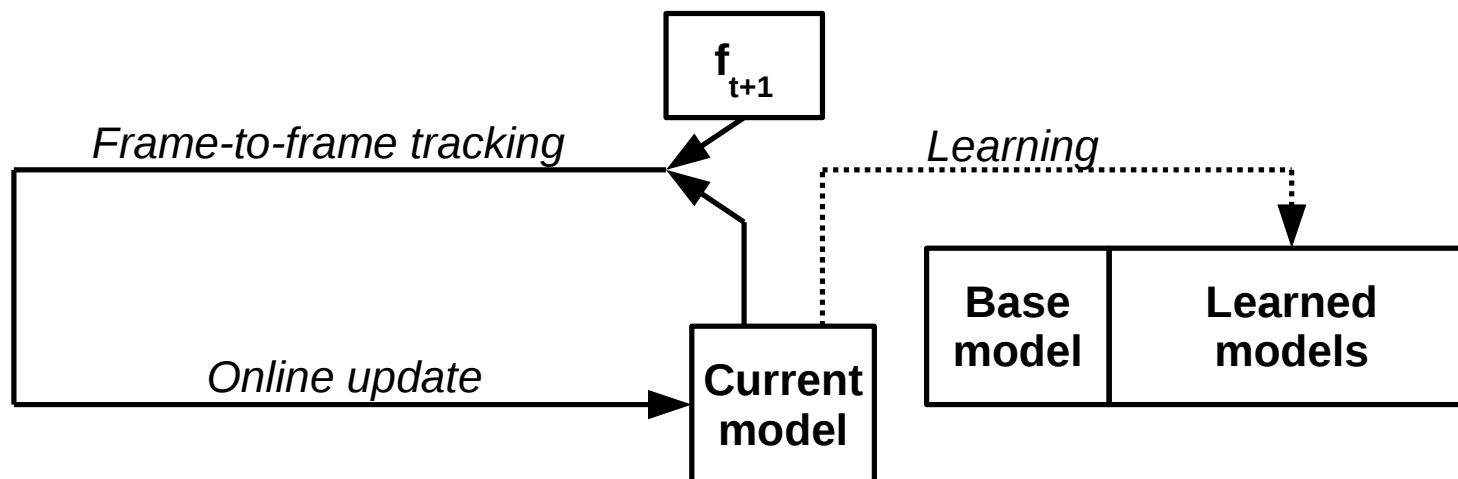
# Using Multiple Models

Base model = initial state at the first frame.



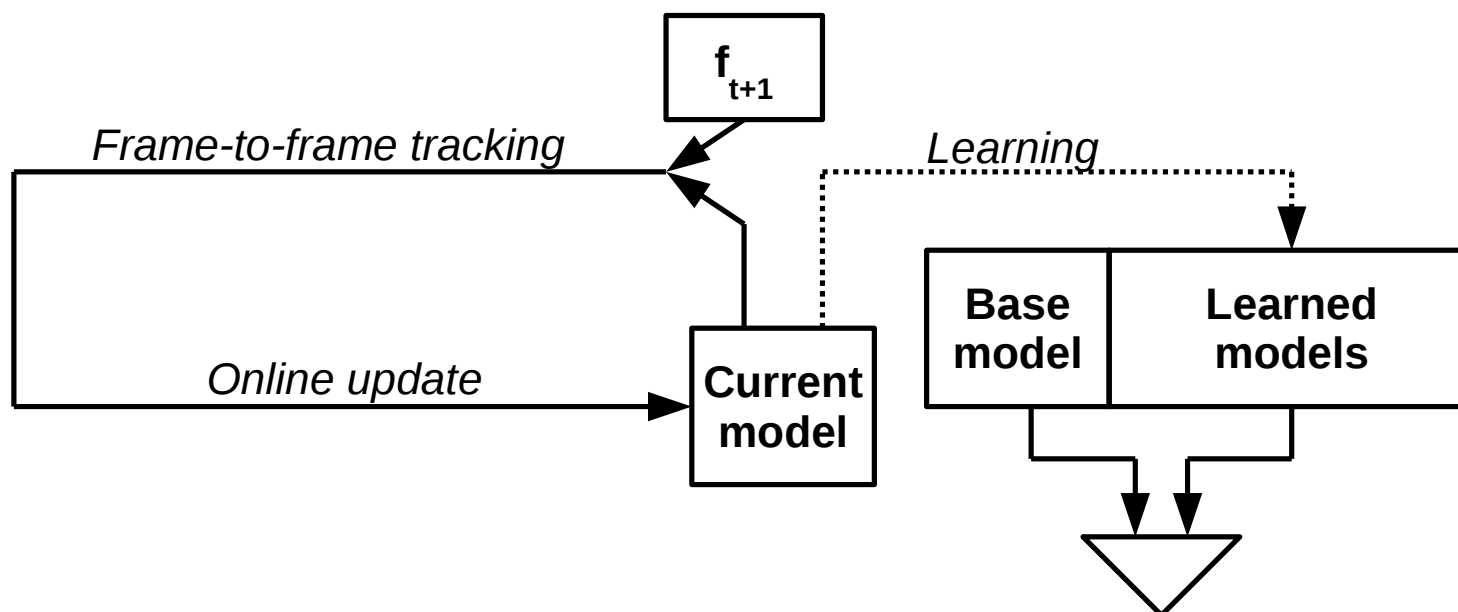
# Using Multiple Models

Models with high confidence are stored for later use.



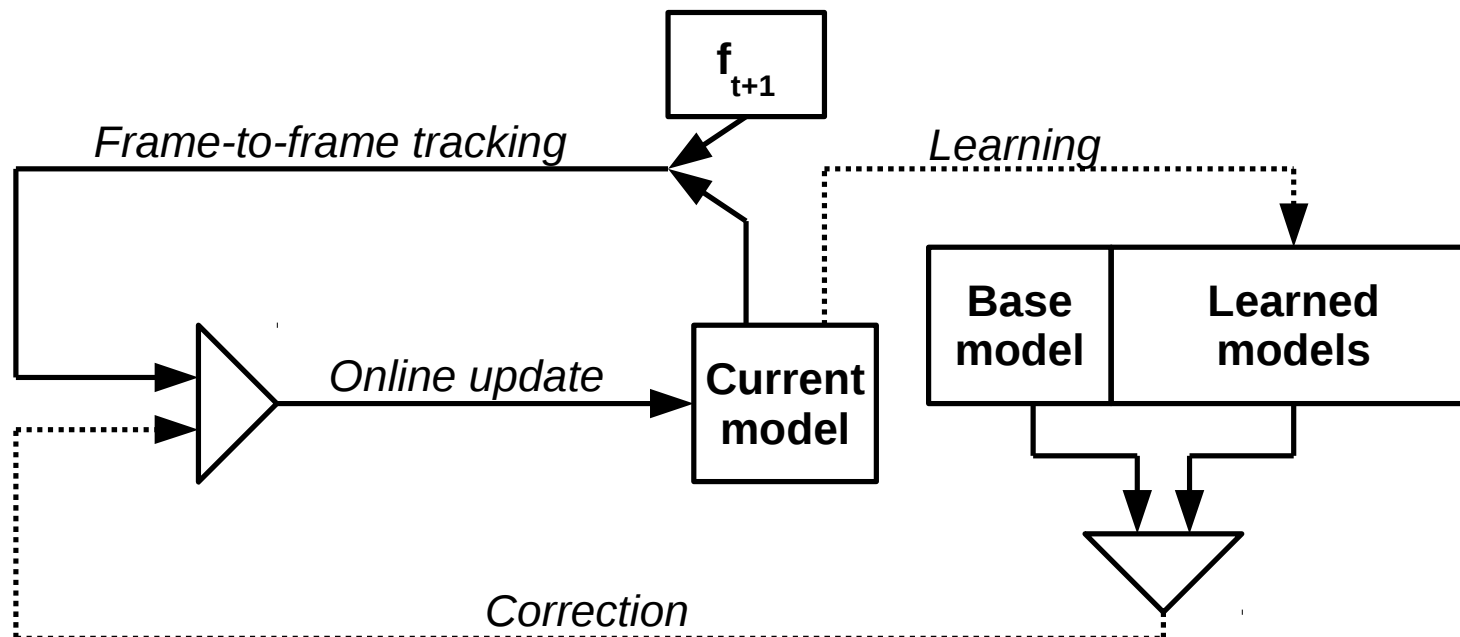
# Using Multiple Models

On sudden drop of **current** model score, the **base** model and possibly other **learned** models are tried for a **correction**.



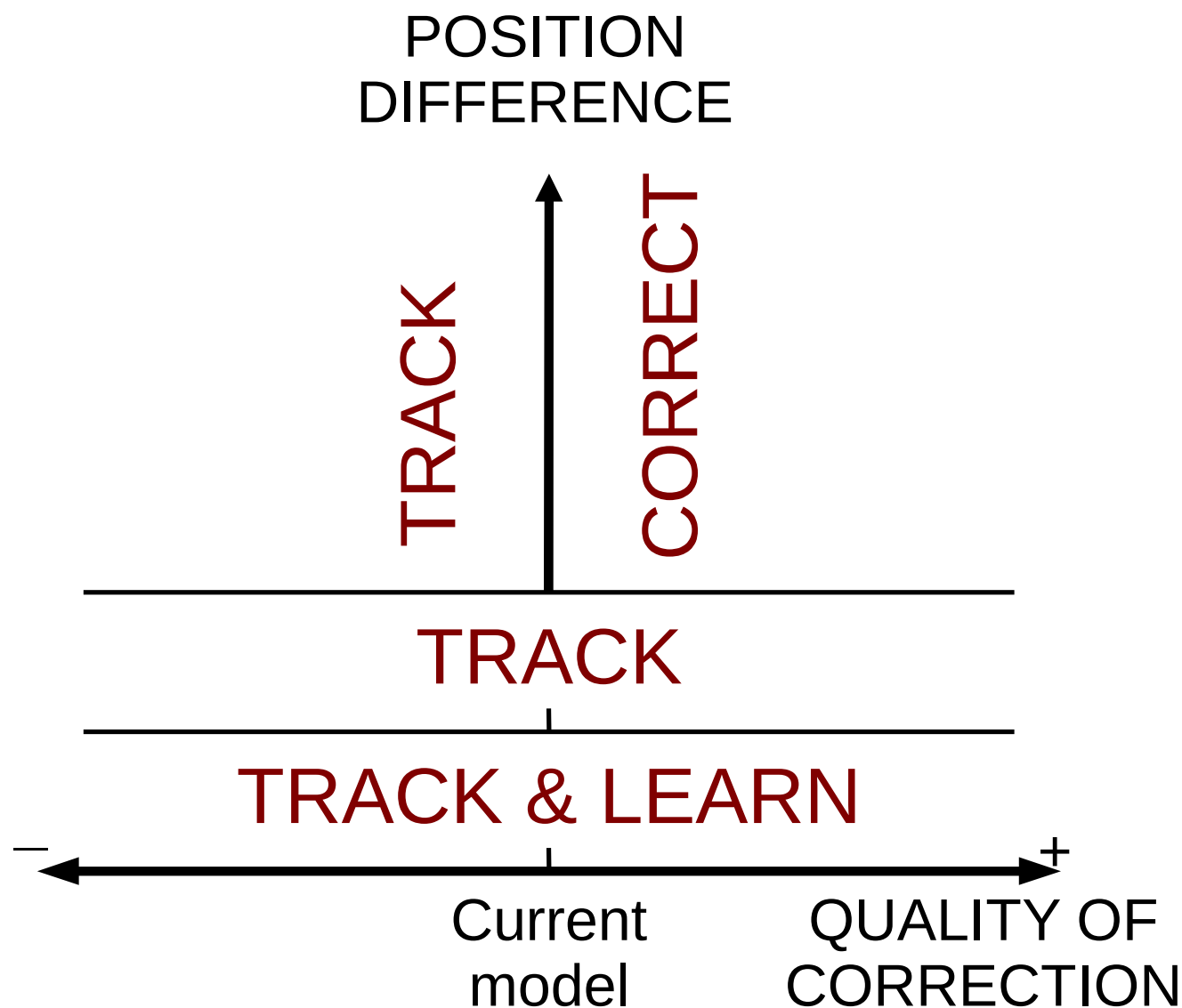
# Using Multiple Models

The best of **correction** models is then compared with **current** model and tracker pose is recovered if necessary.





# Possible Events In Correction of the Tracker



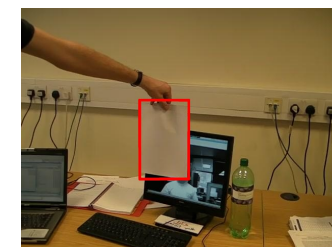
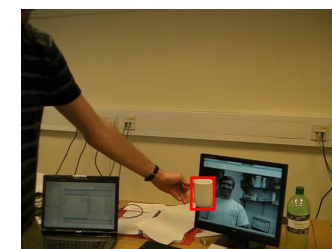
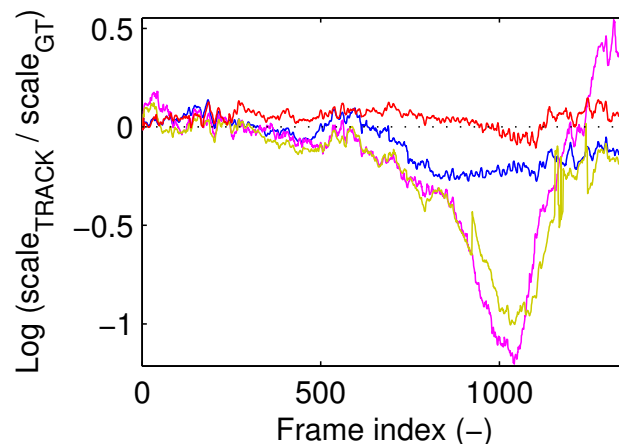
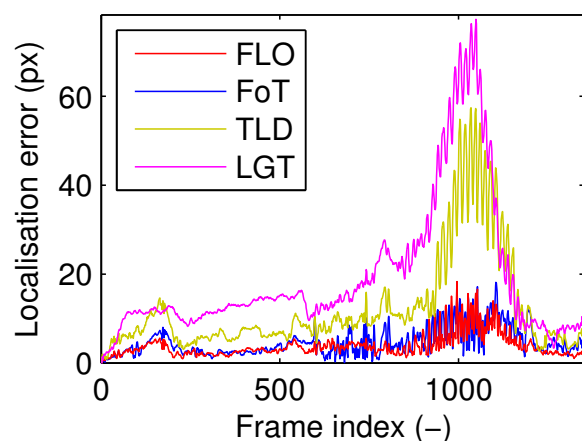
# Experimental Evaluation

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Compared with FoT (LK, Vojíř, '10), LGT (Čehovin, '11), TLD (Kálal, '11), results are competitive or superior.

Measured:

- ◆ localization error (distance of center)
- ◆ scale error (ratio of size to GT size)

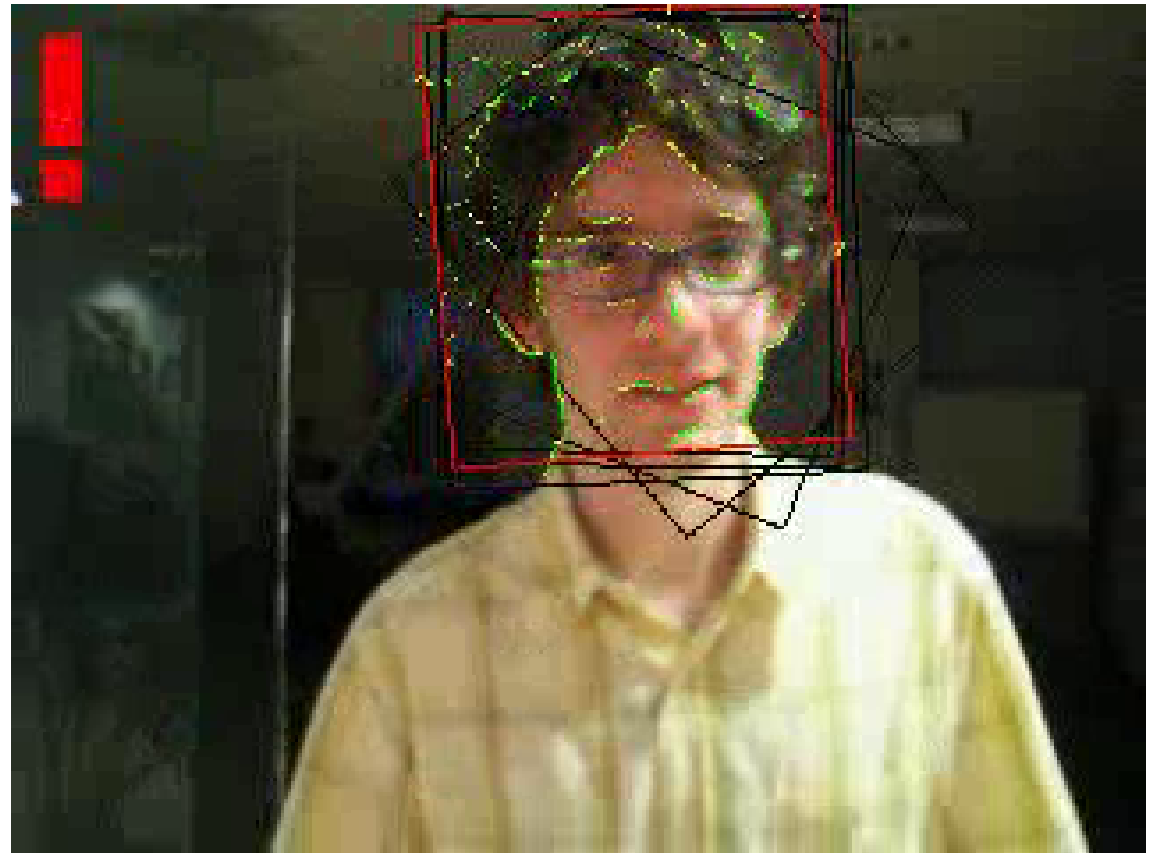
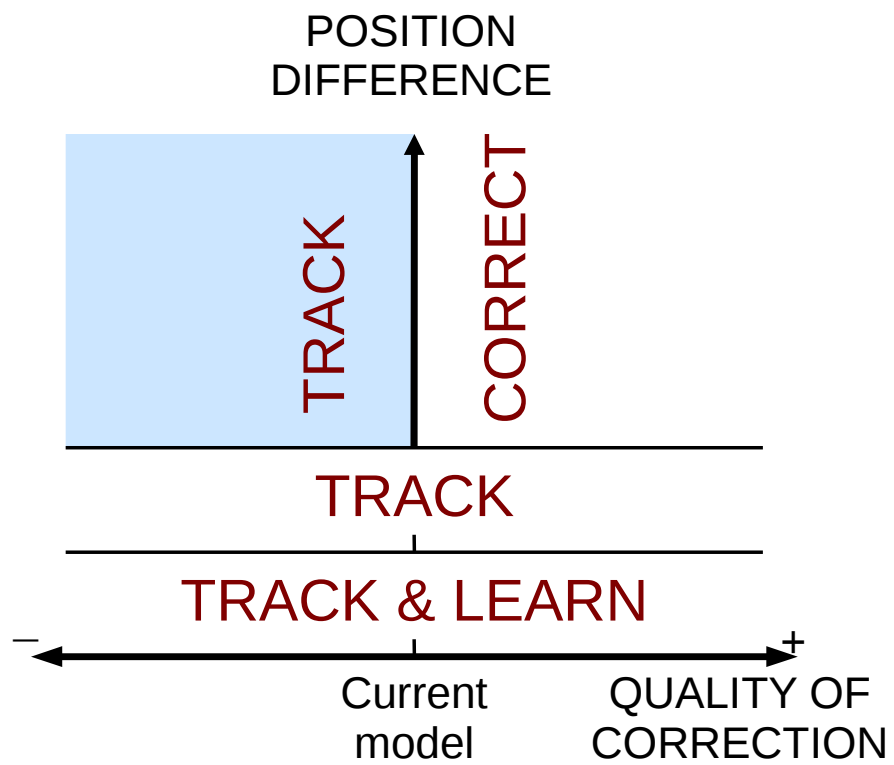


Matas, J. and Vojíř, T.: Robustifying the flock of trackers. Proc. of CVWW 2011.

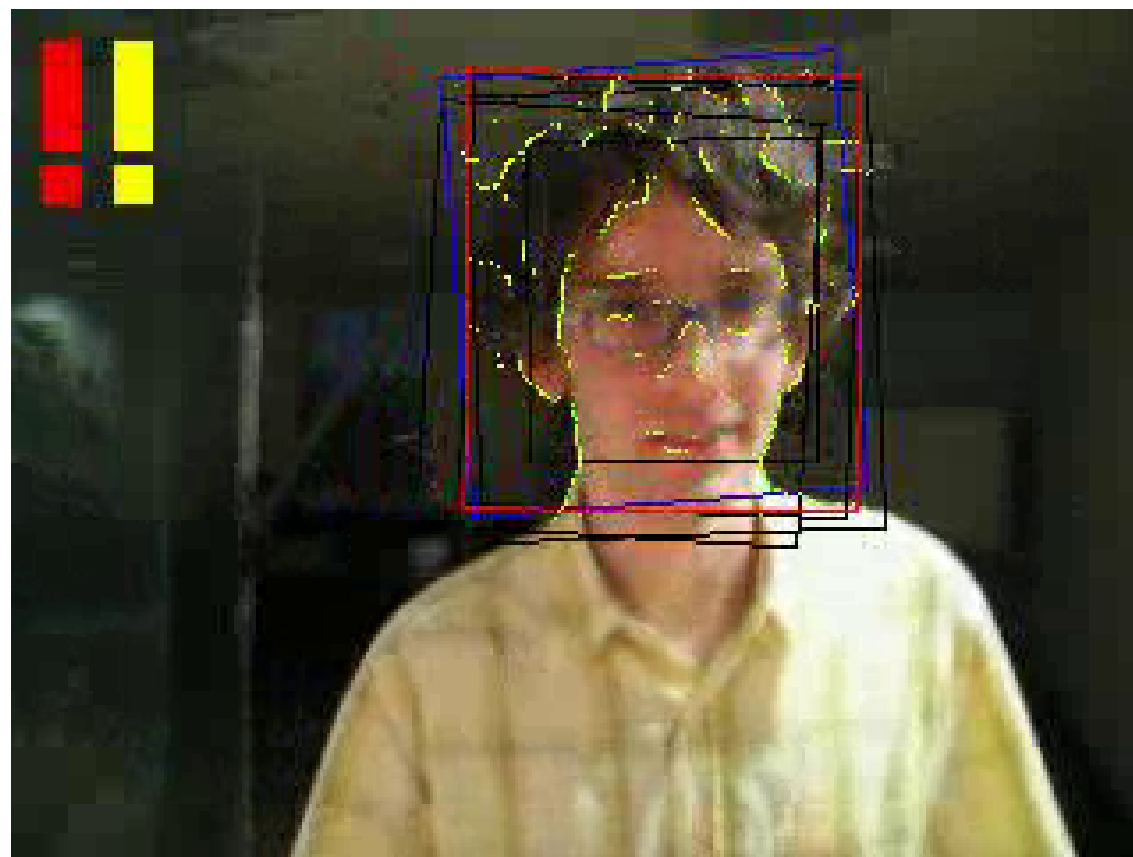
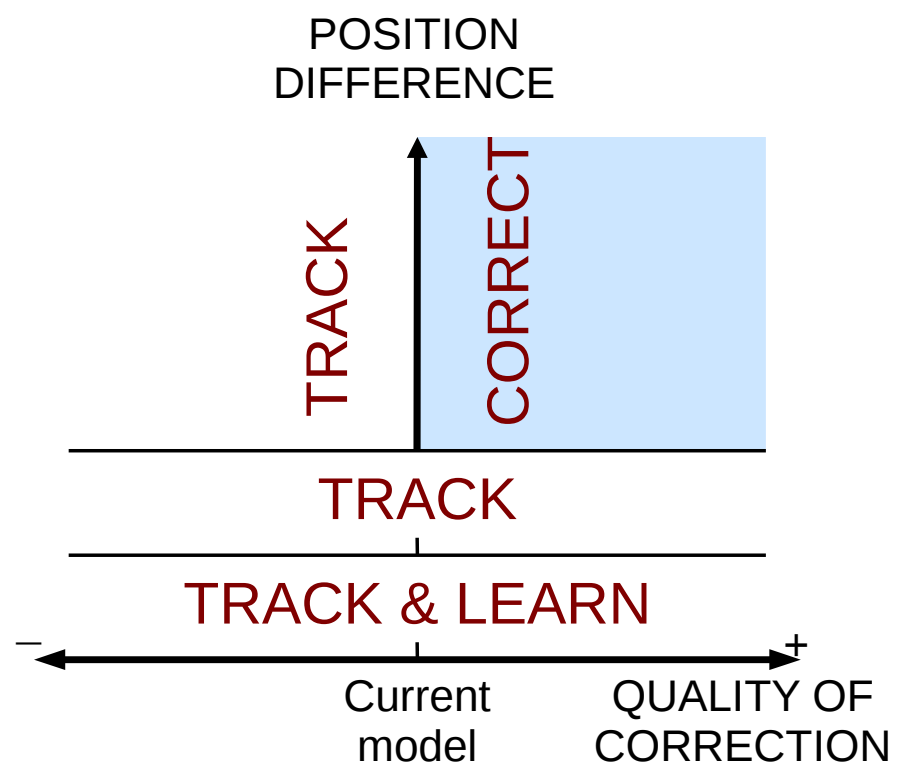
Cehovin, L. *et al.*: An adaptive coupled-layer visual model for robust visual tracking. Proc. of ICCV 2011.

Kálal, Z. *et al.*: P-N learning: Bootstrapping binary classifiers by structural constraints. Proc. of CVPR 2010.

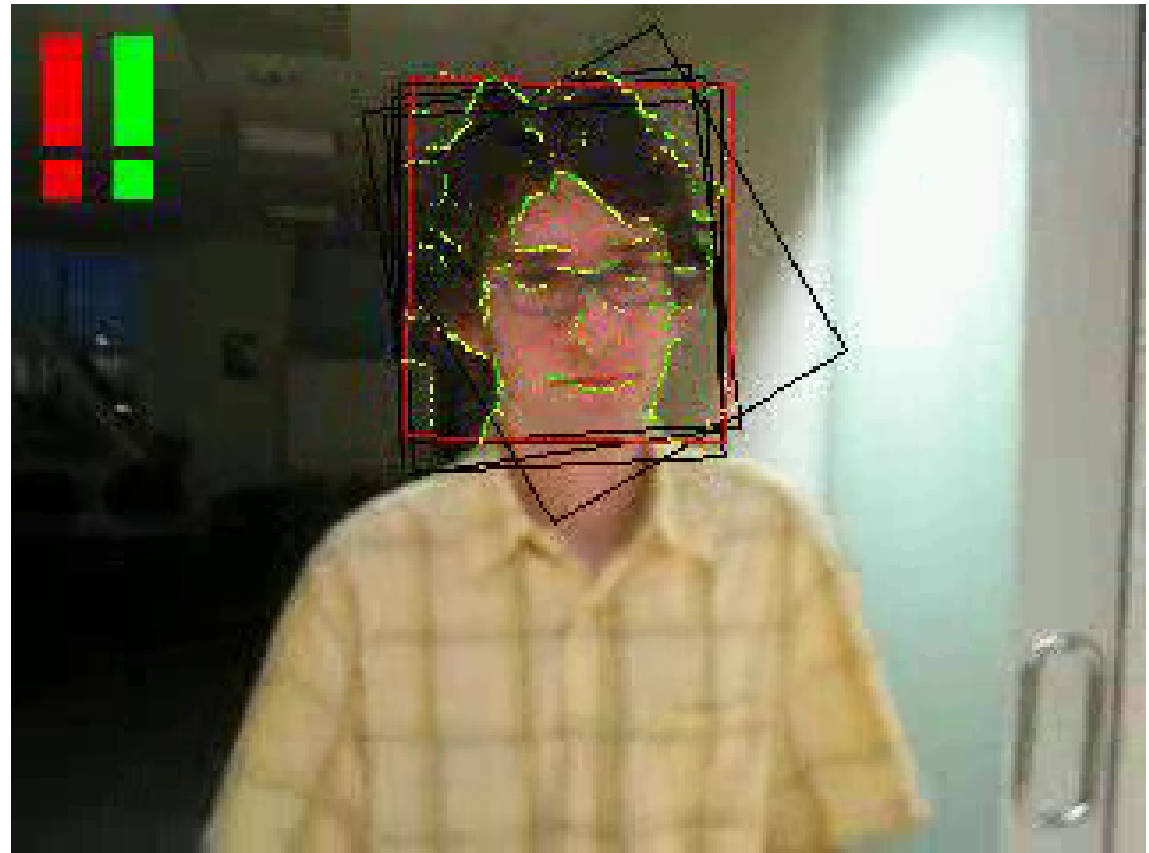
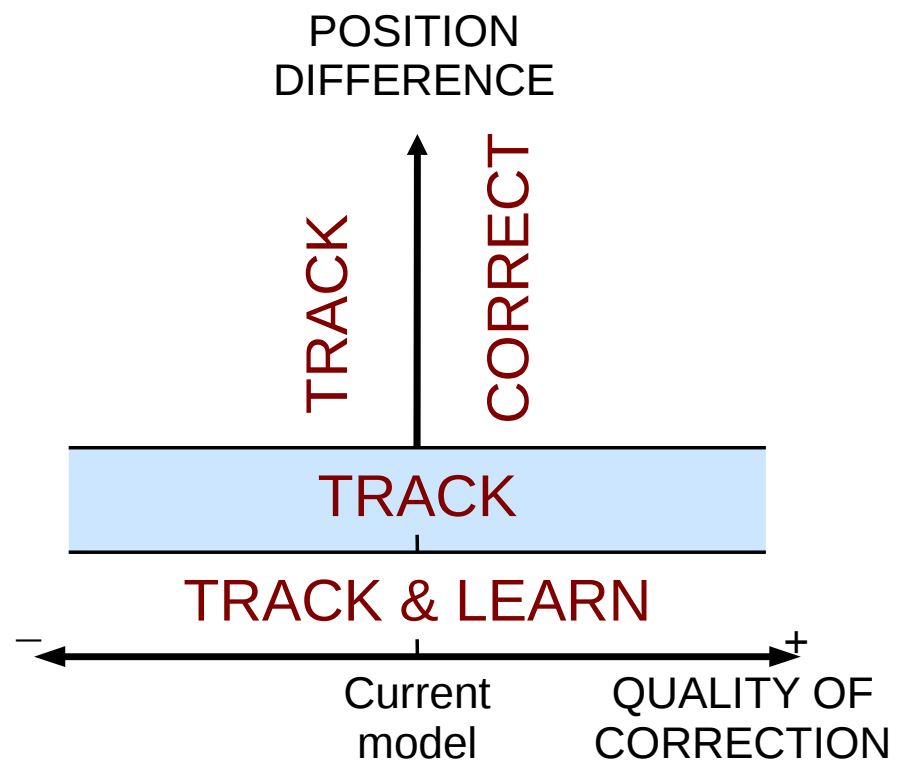
# Video Example



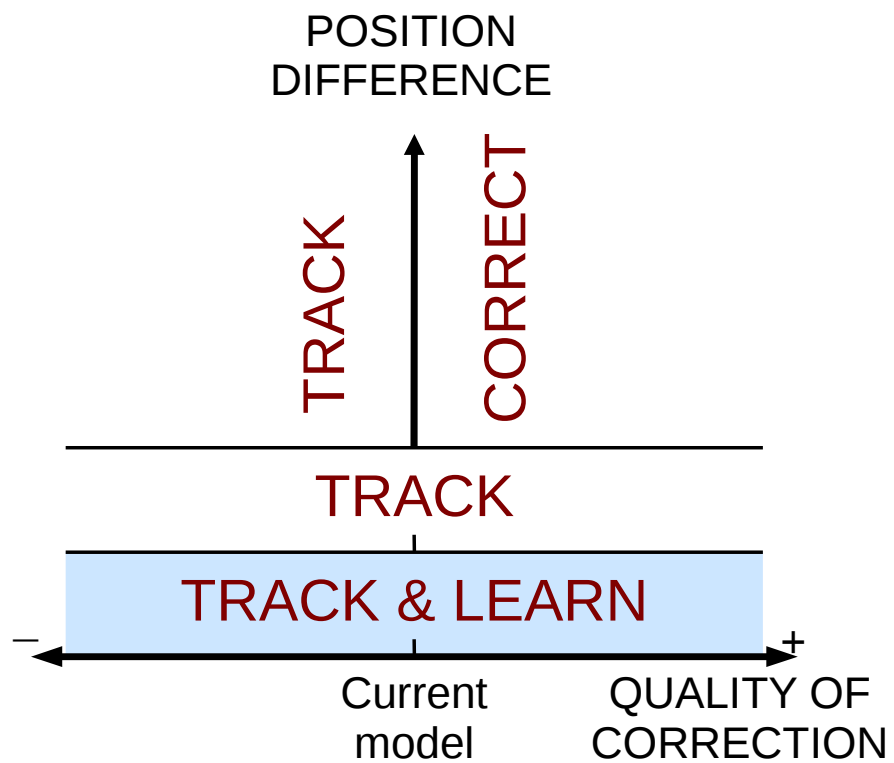
# Video Example



# Video Example



# Video Example



# Video Example

David.avi

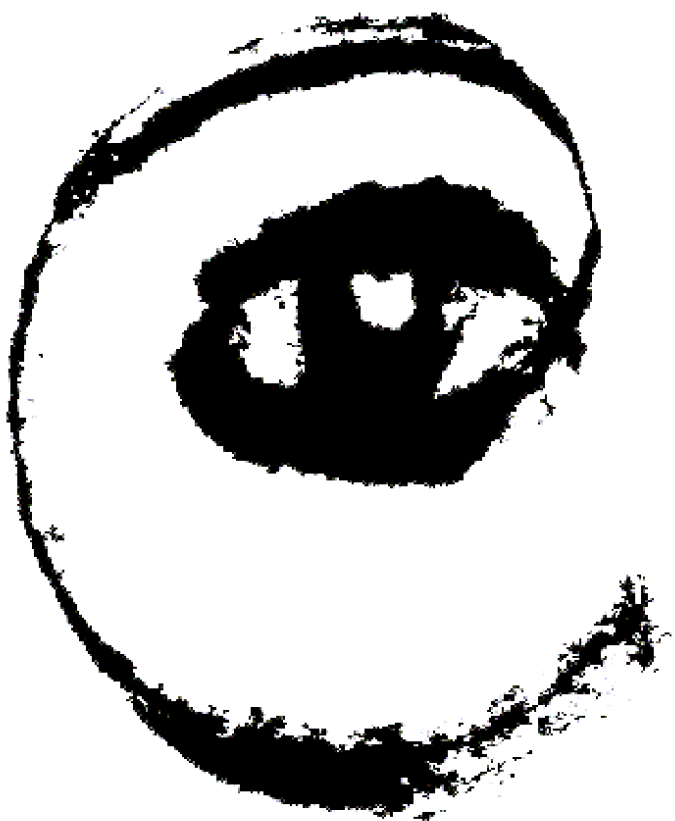
# Conclusions

- ◆ designed, implemented and tested the FLO tracker using a novel idea of tracking tangent lines
- ◆ competitive on scenes with the sufficient number of distinguishable features
- ◆ superior on scenes without such features
- ◆ disadvantage: still short-term tracking (no redetection, once lost = forever lost)

Thanks for attention! Any questions?

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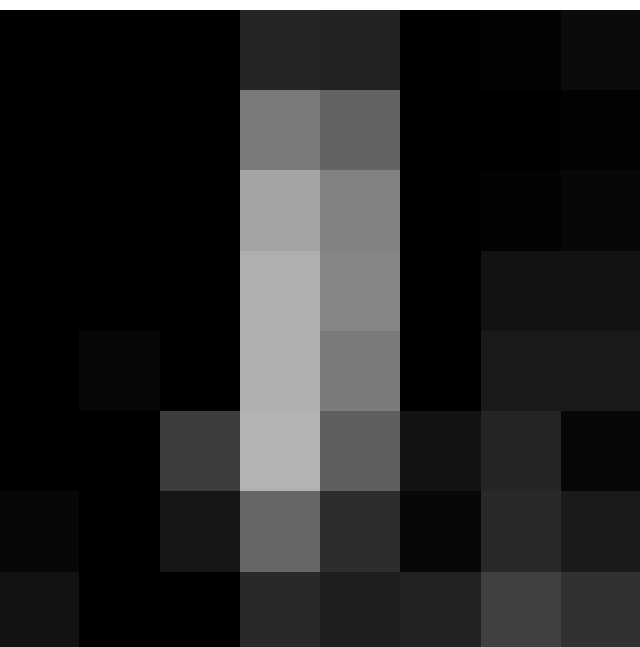


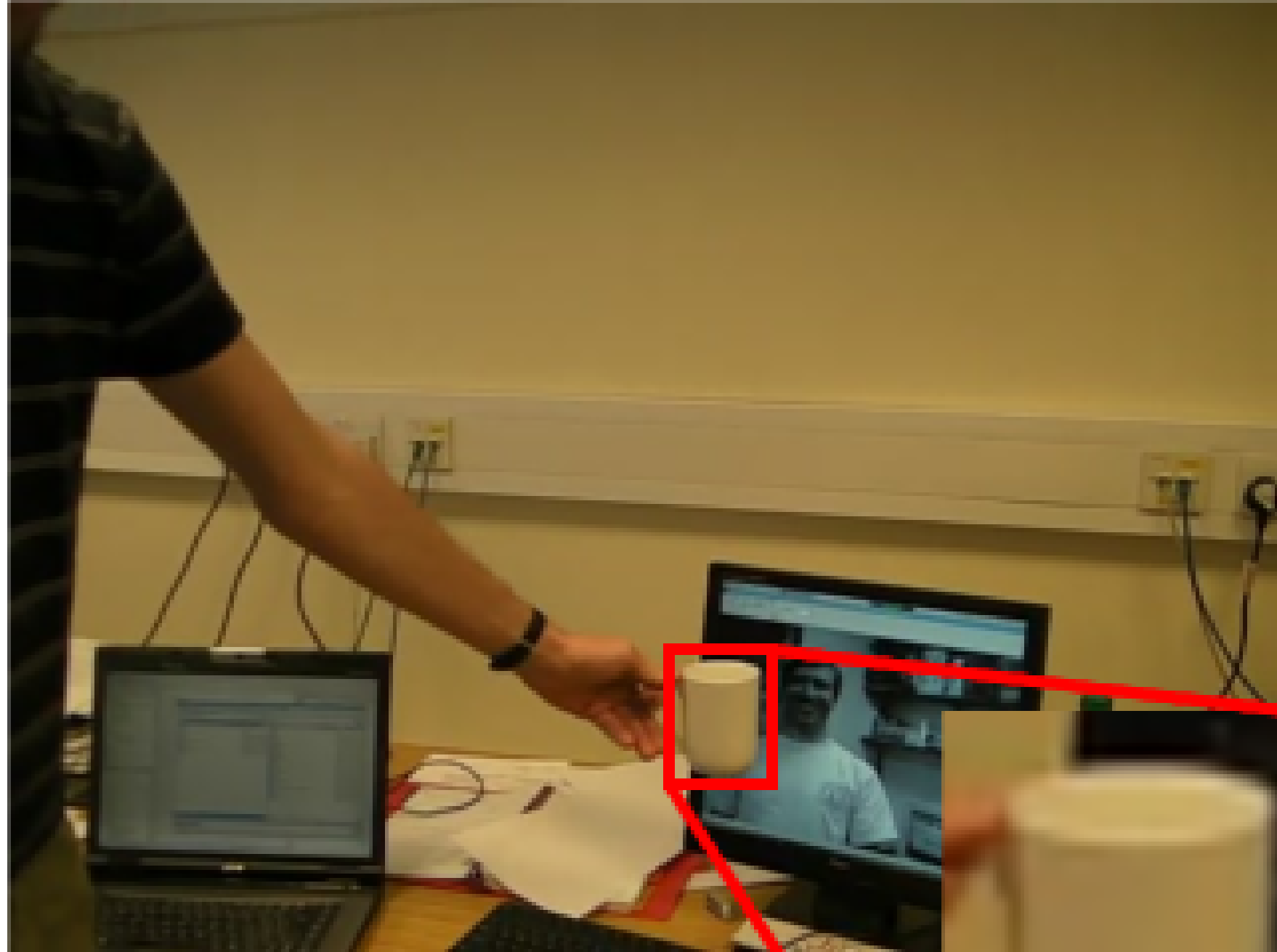


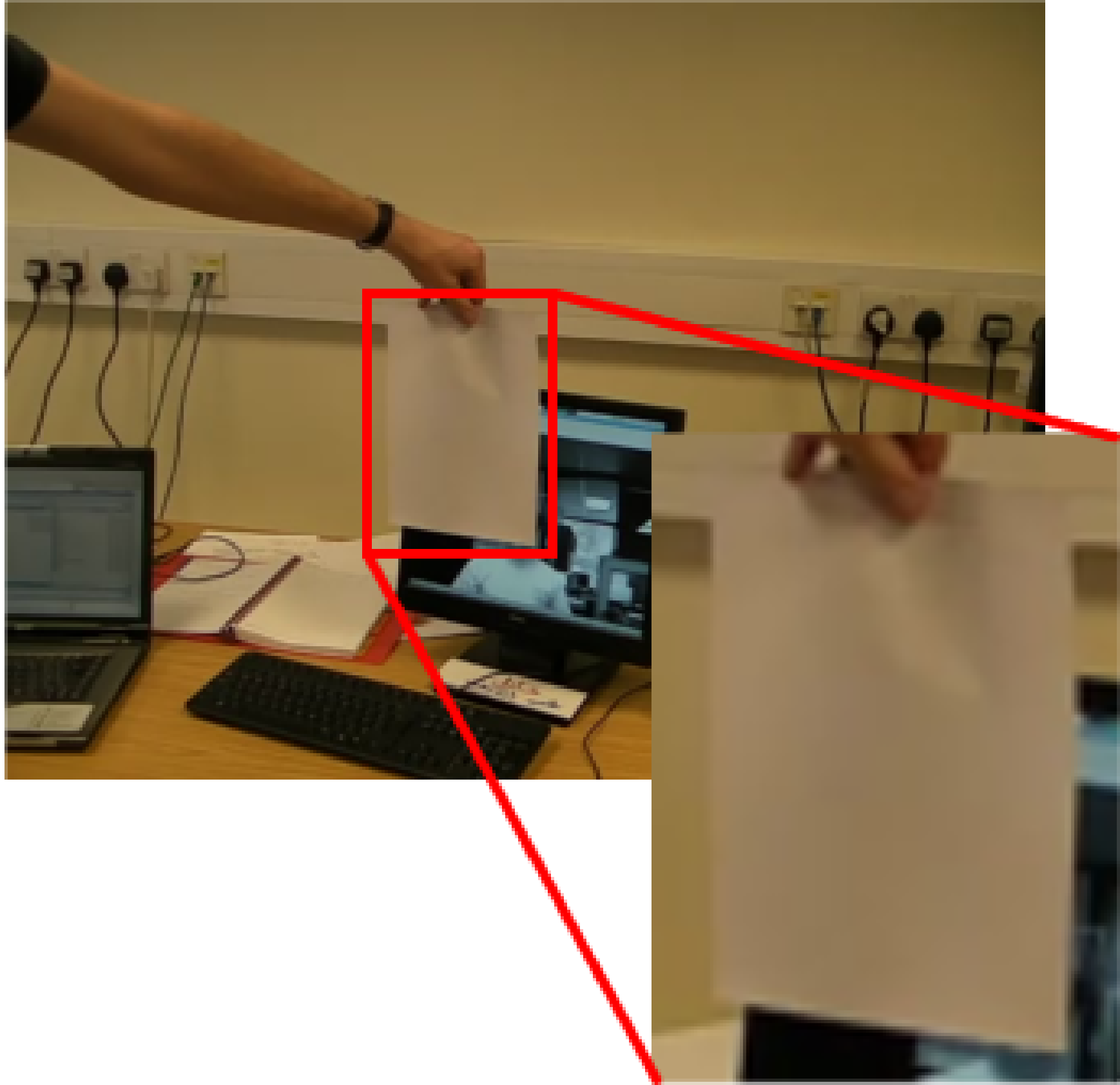
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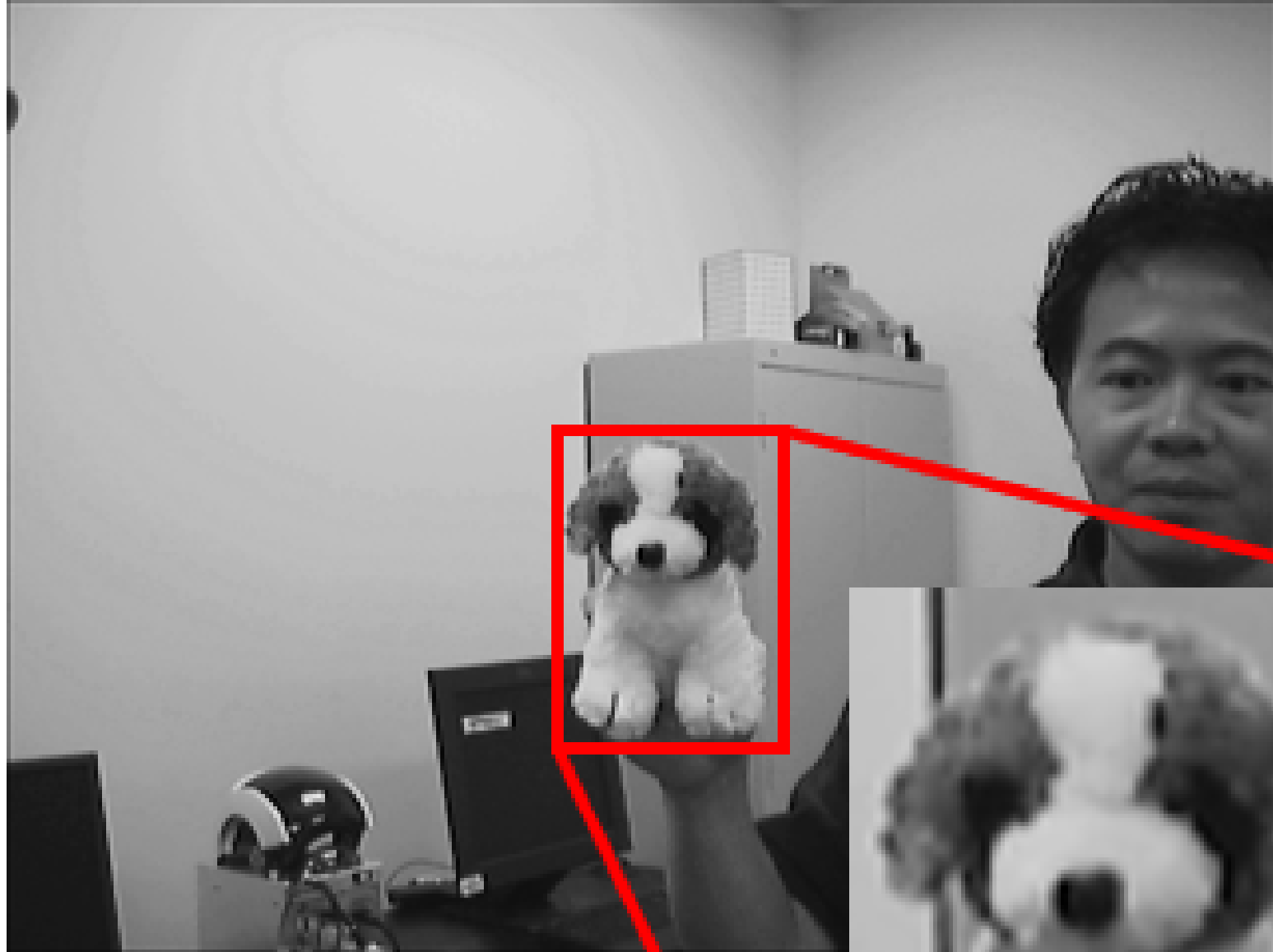


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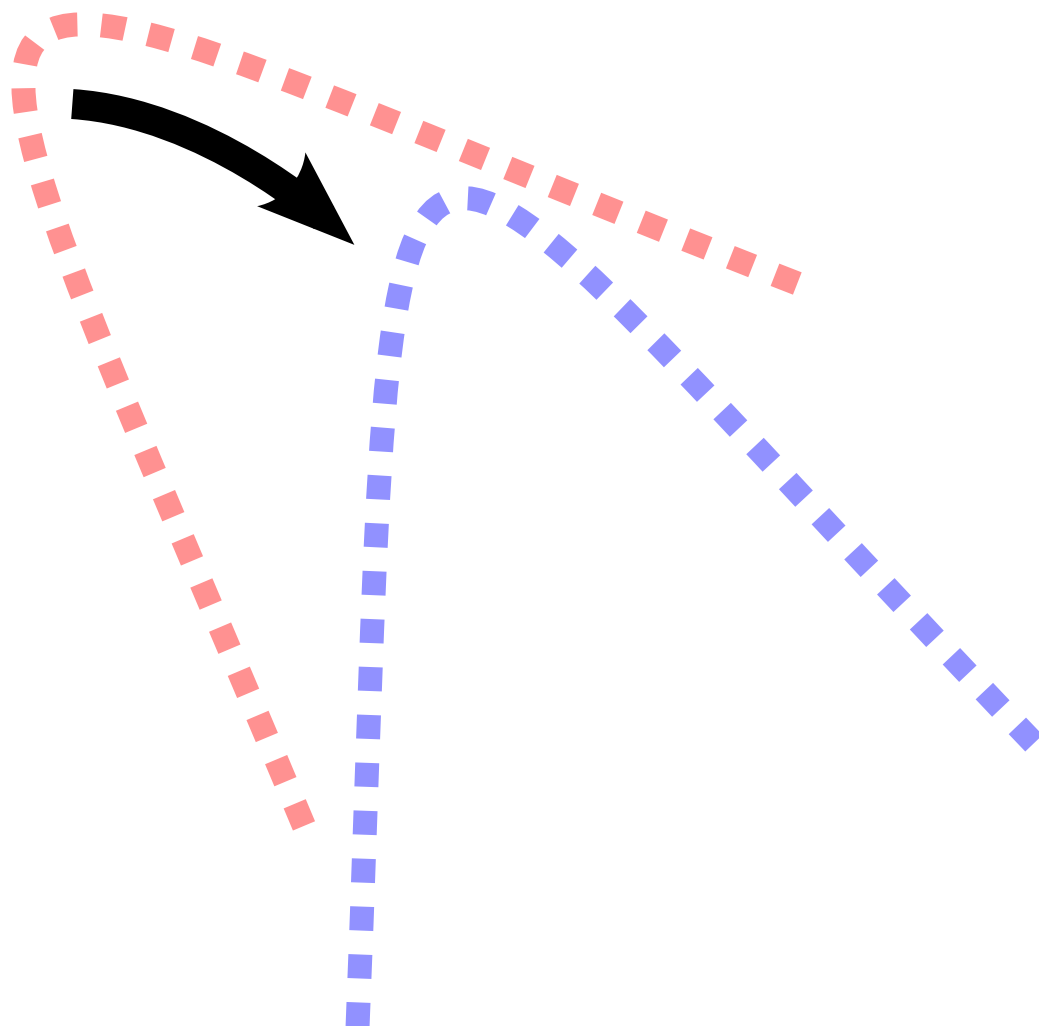




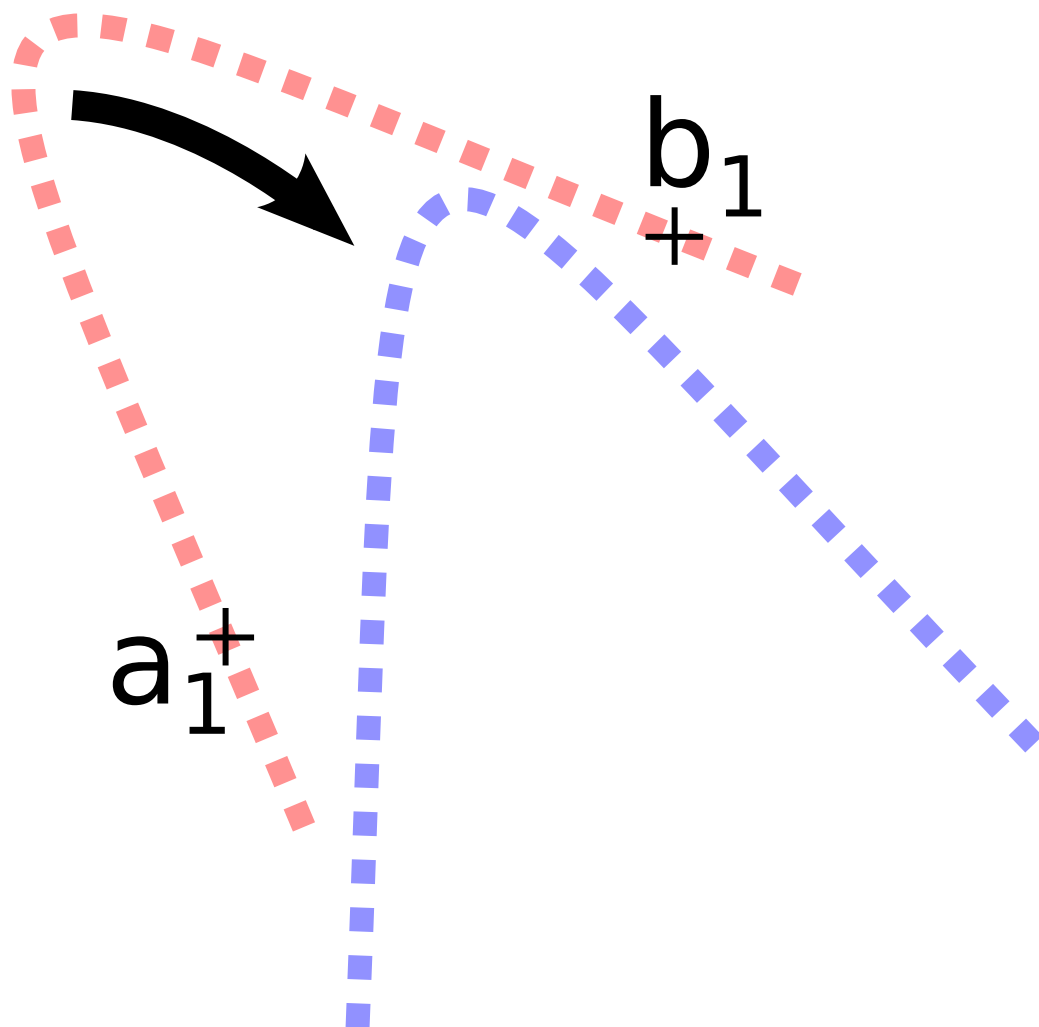




$X_1$   
 $X_2$

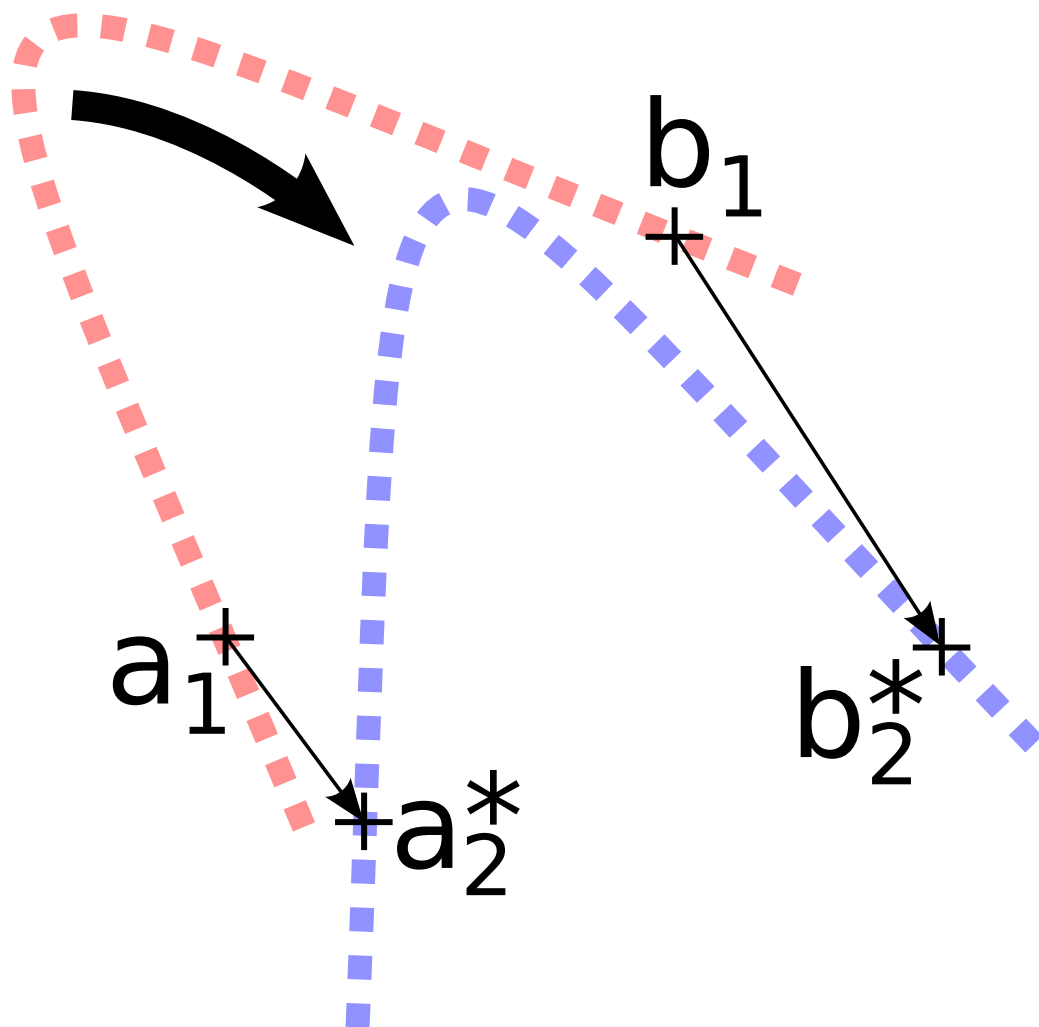


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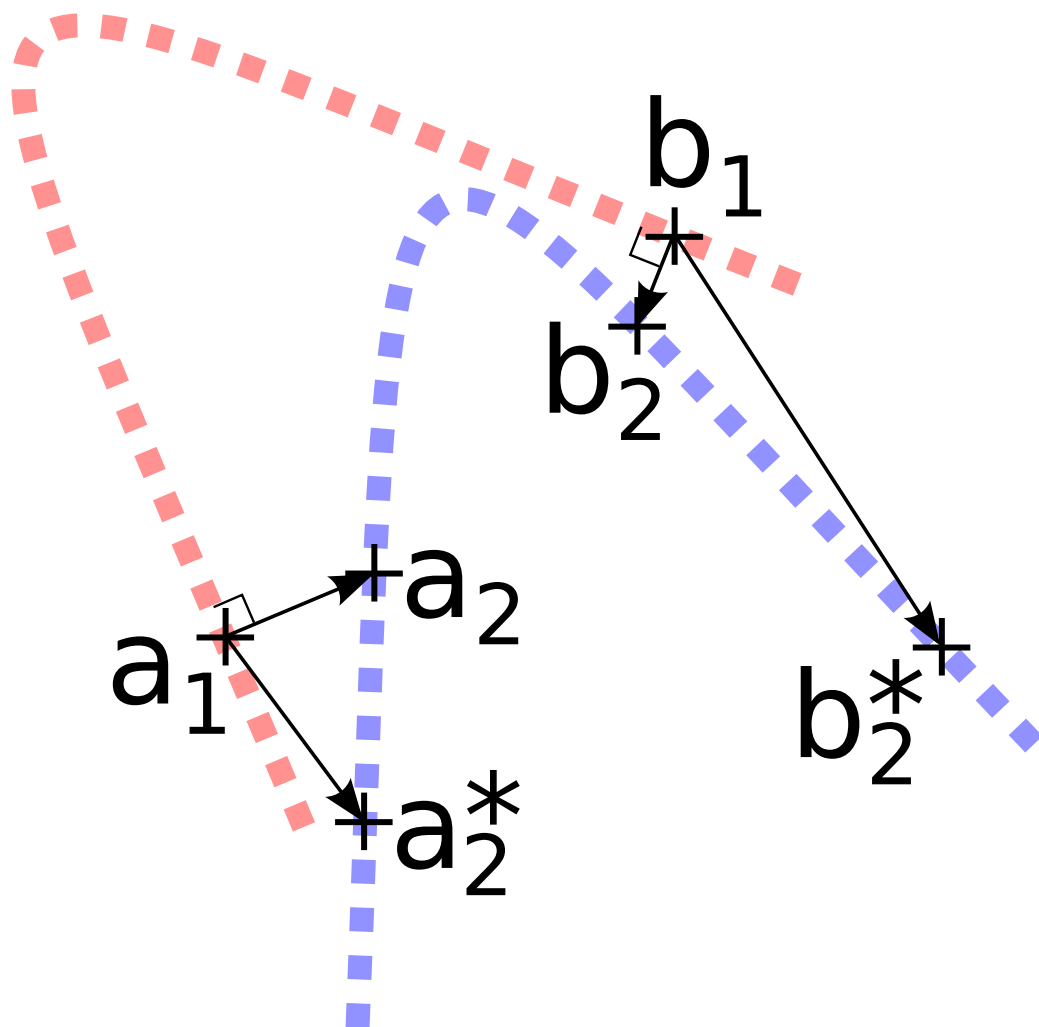


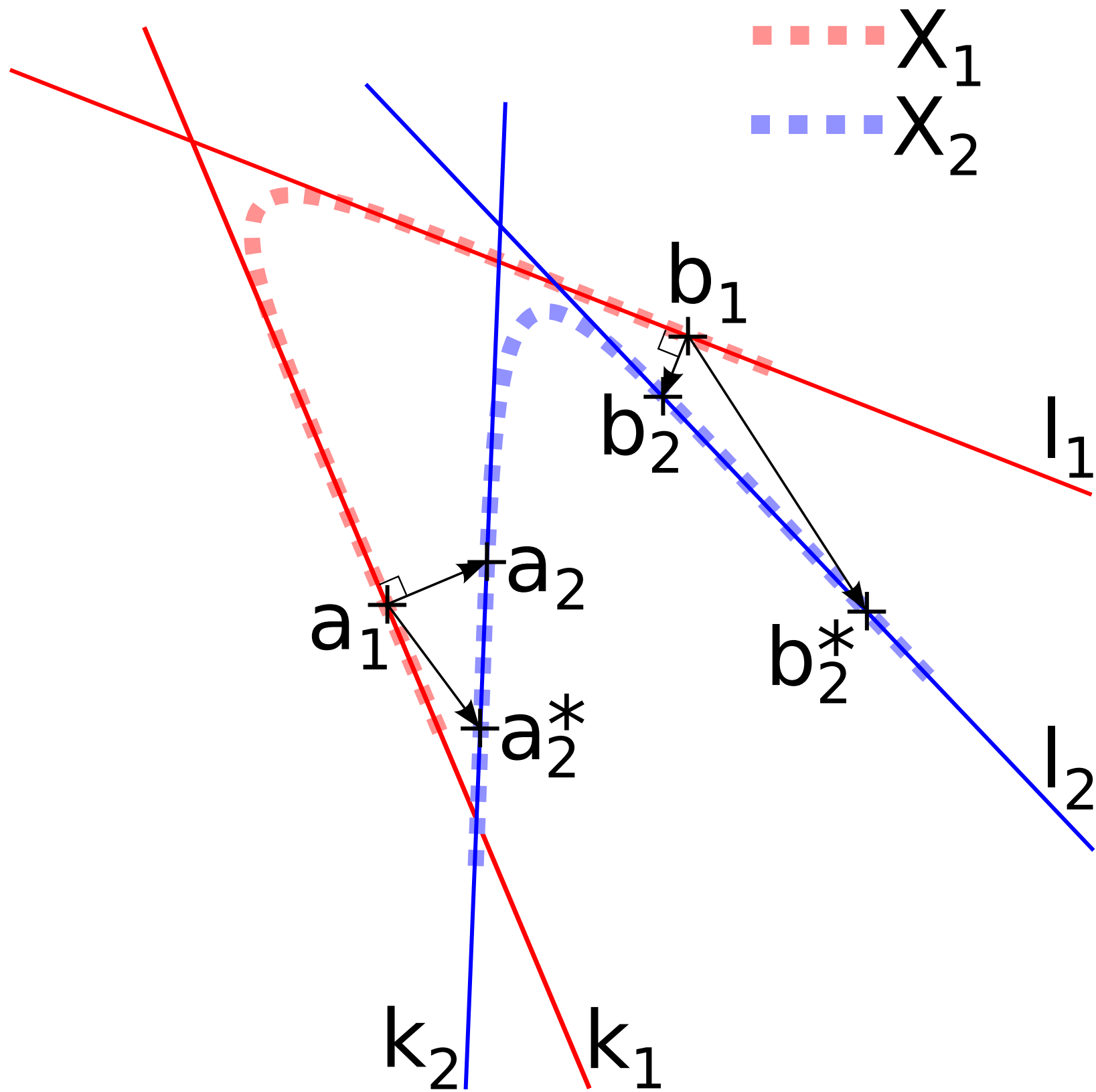


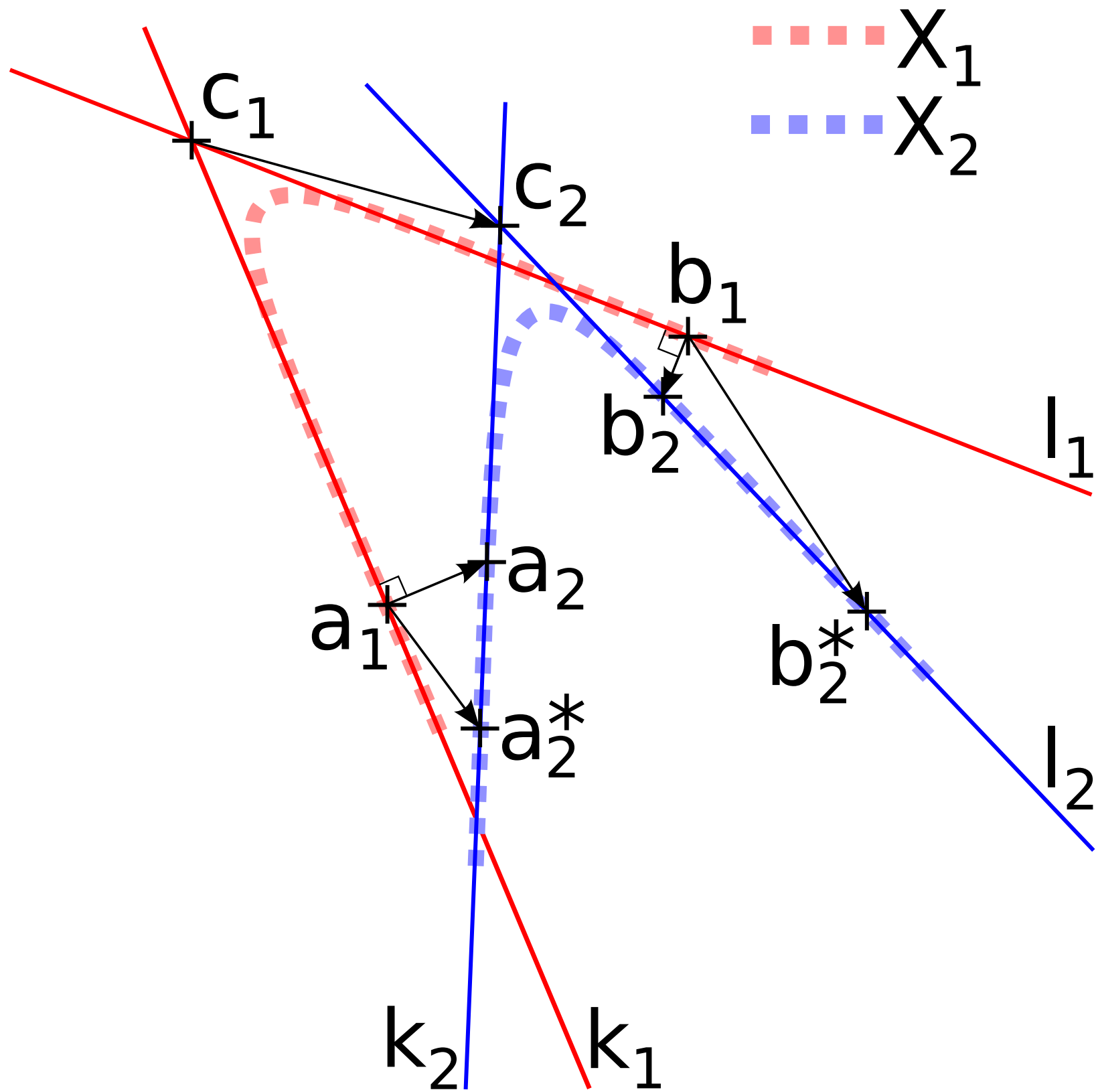
$X_1$   
 $X_2$

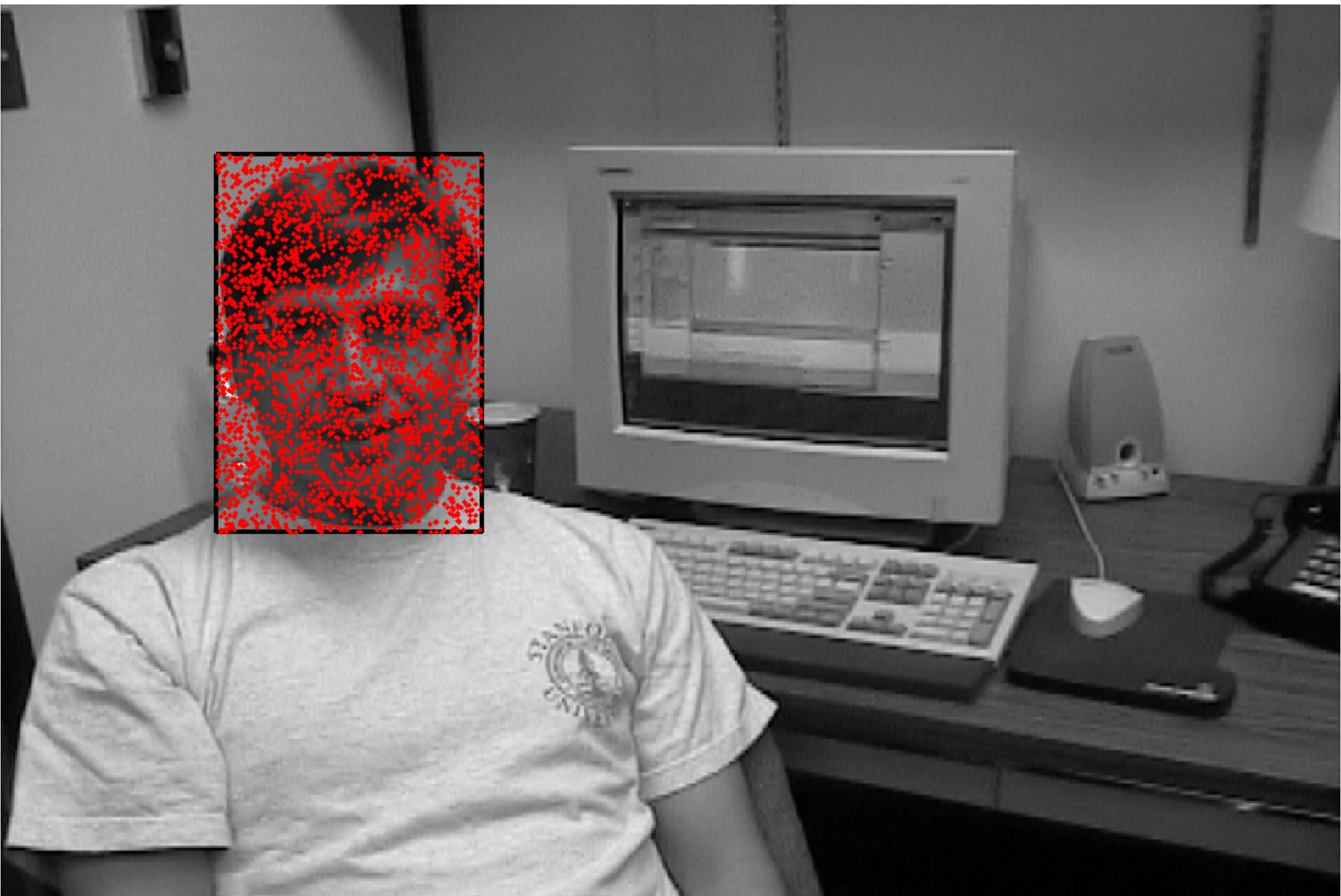


$X_1$   
 $X_2$





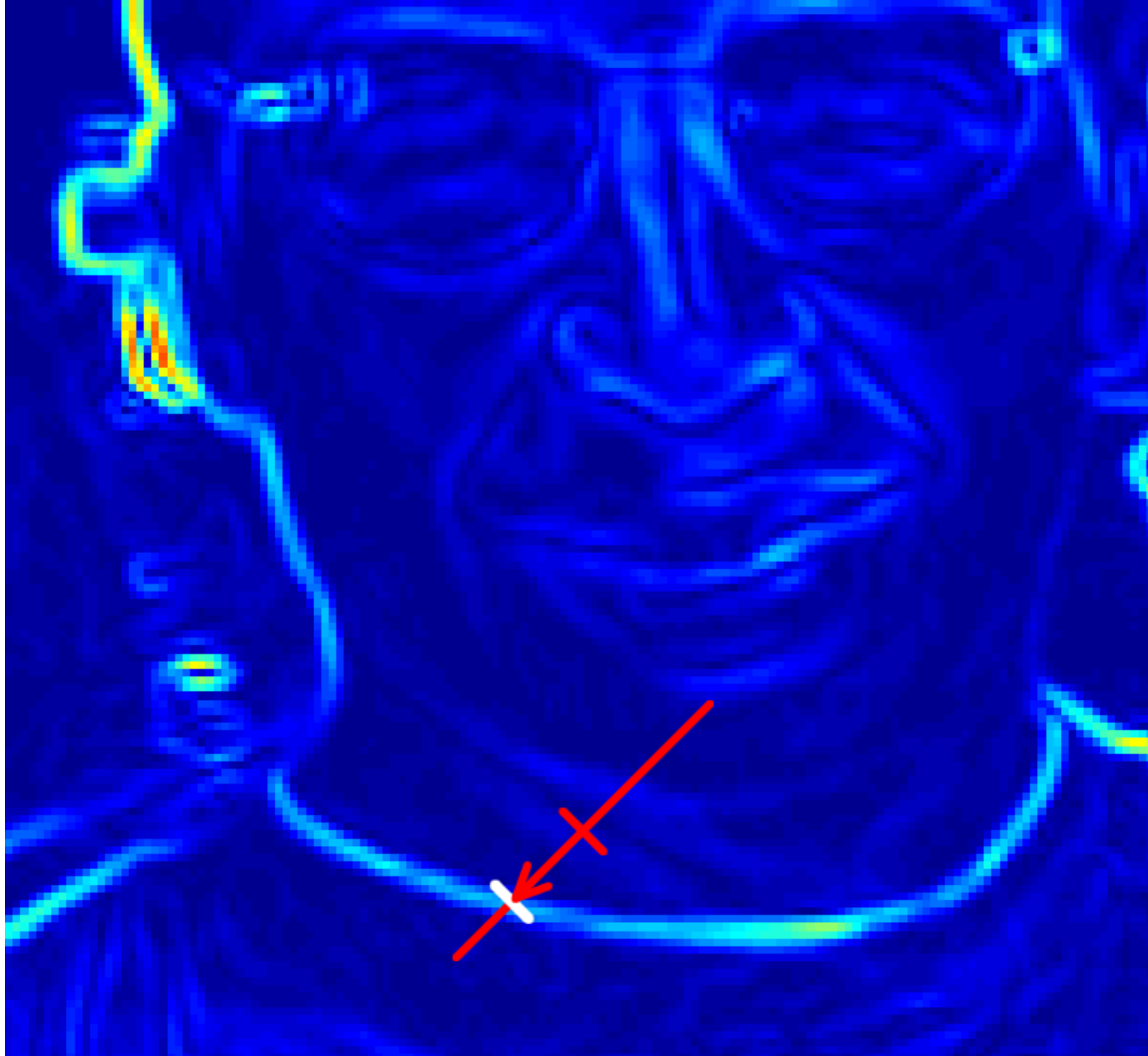




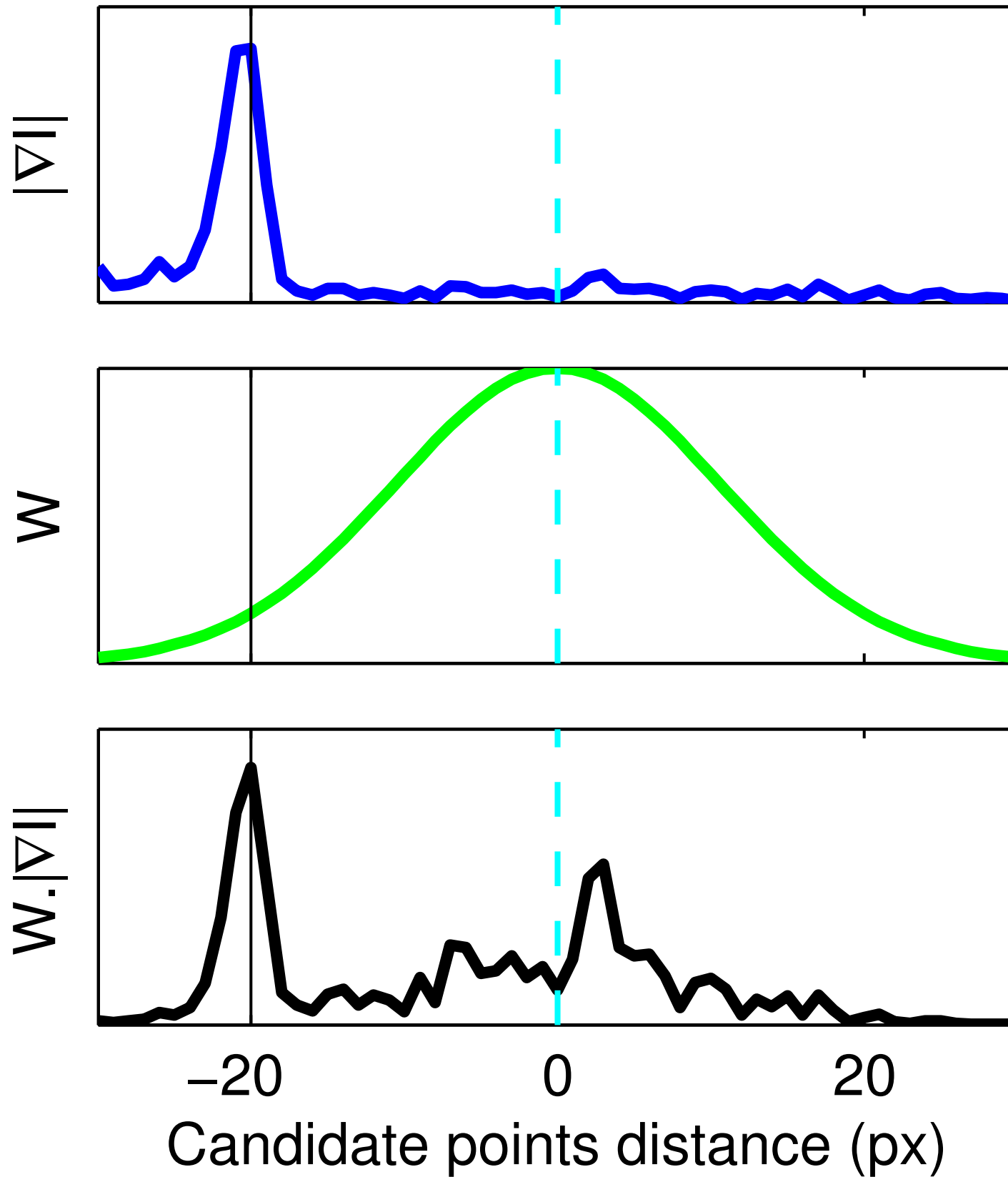


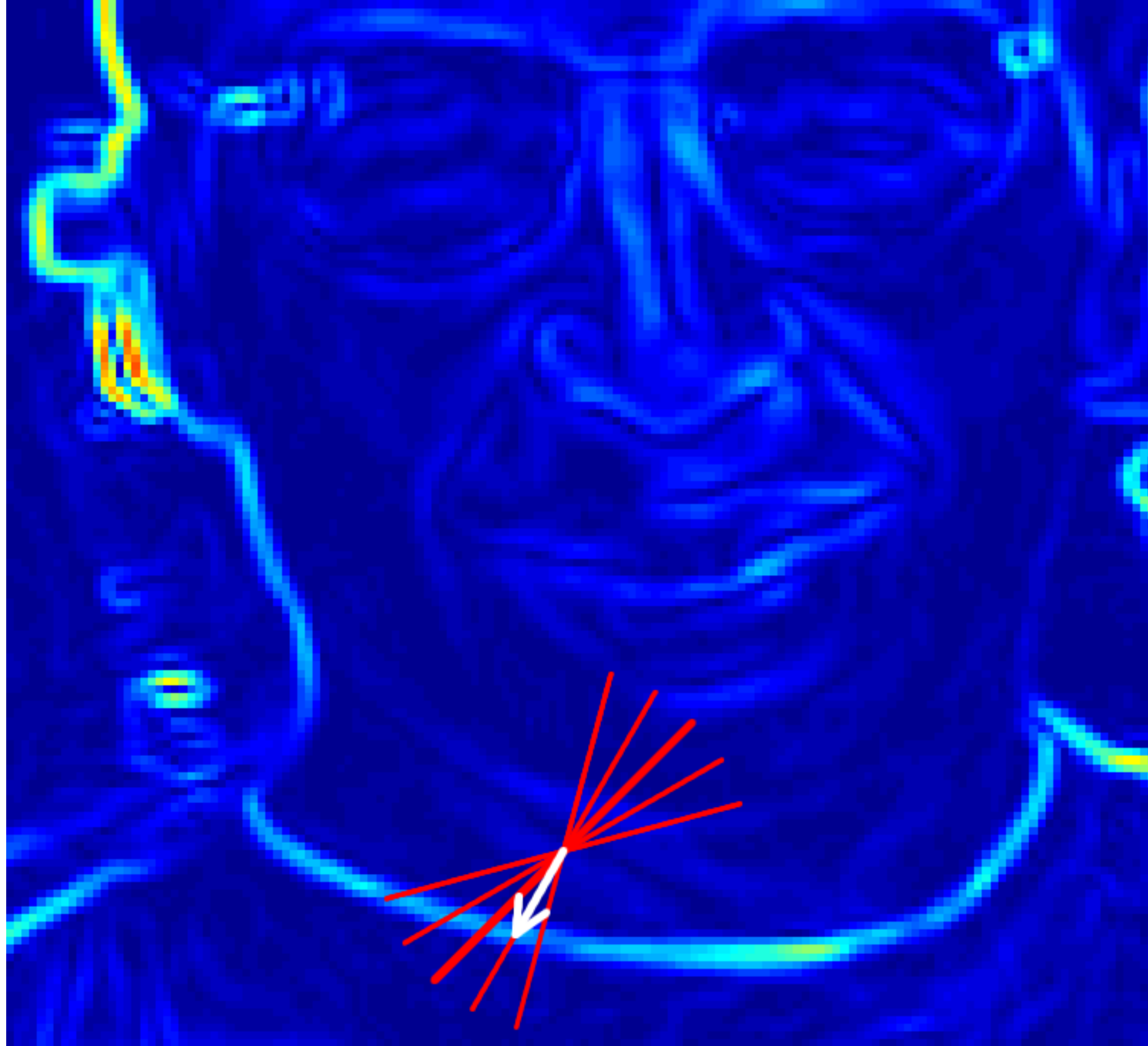


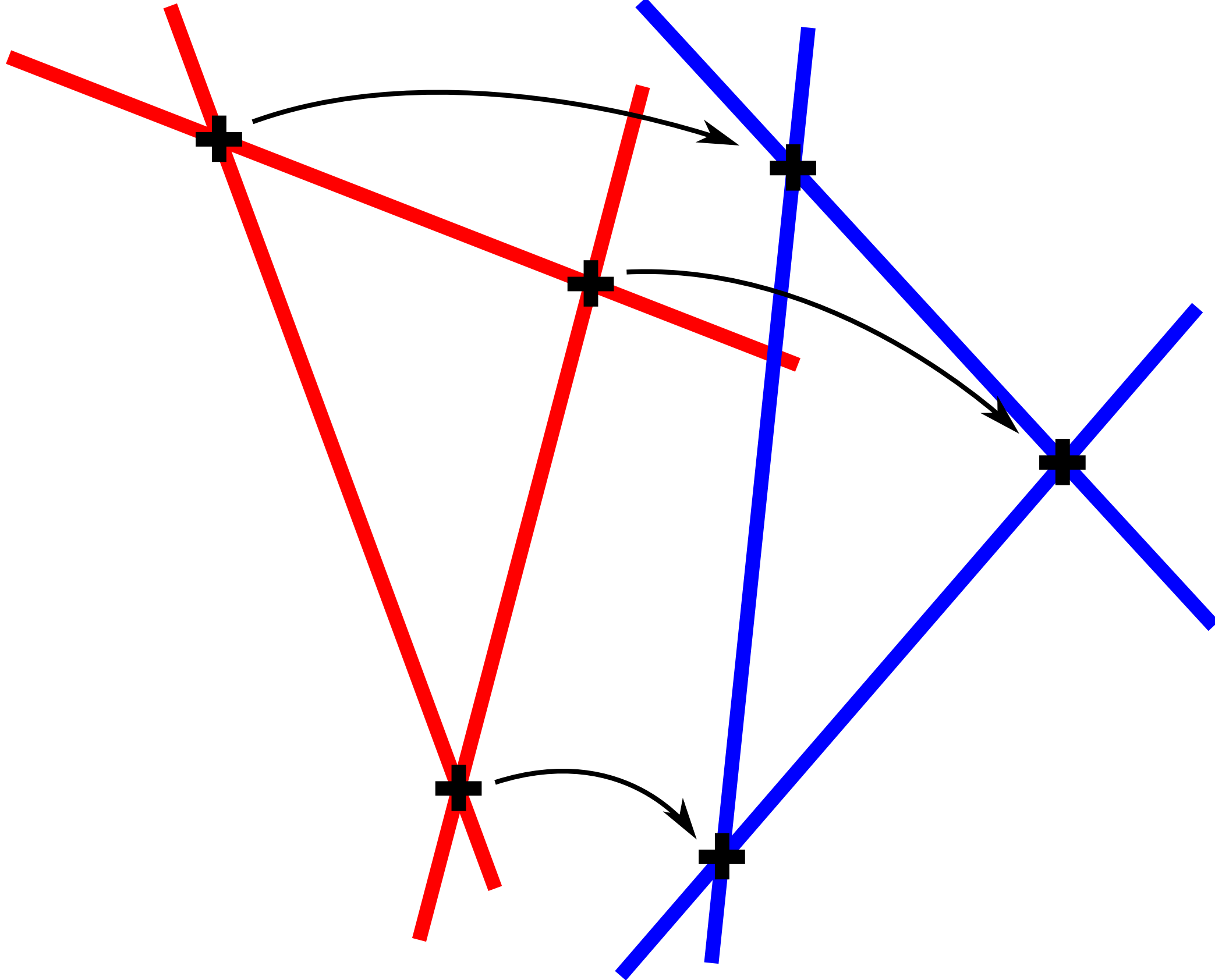


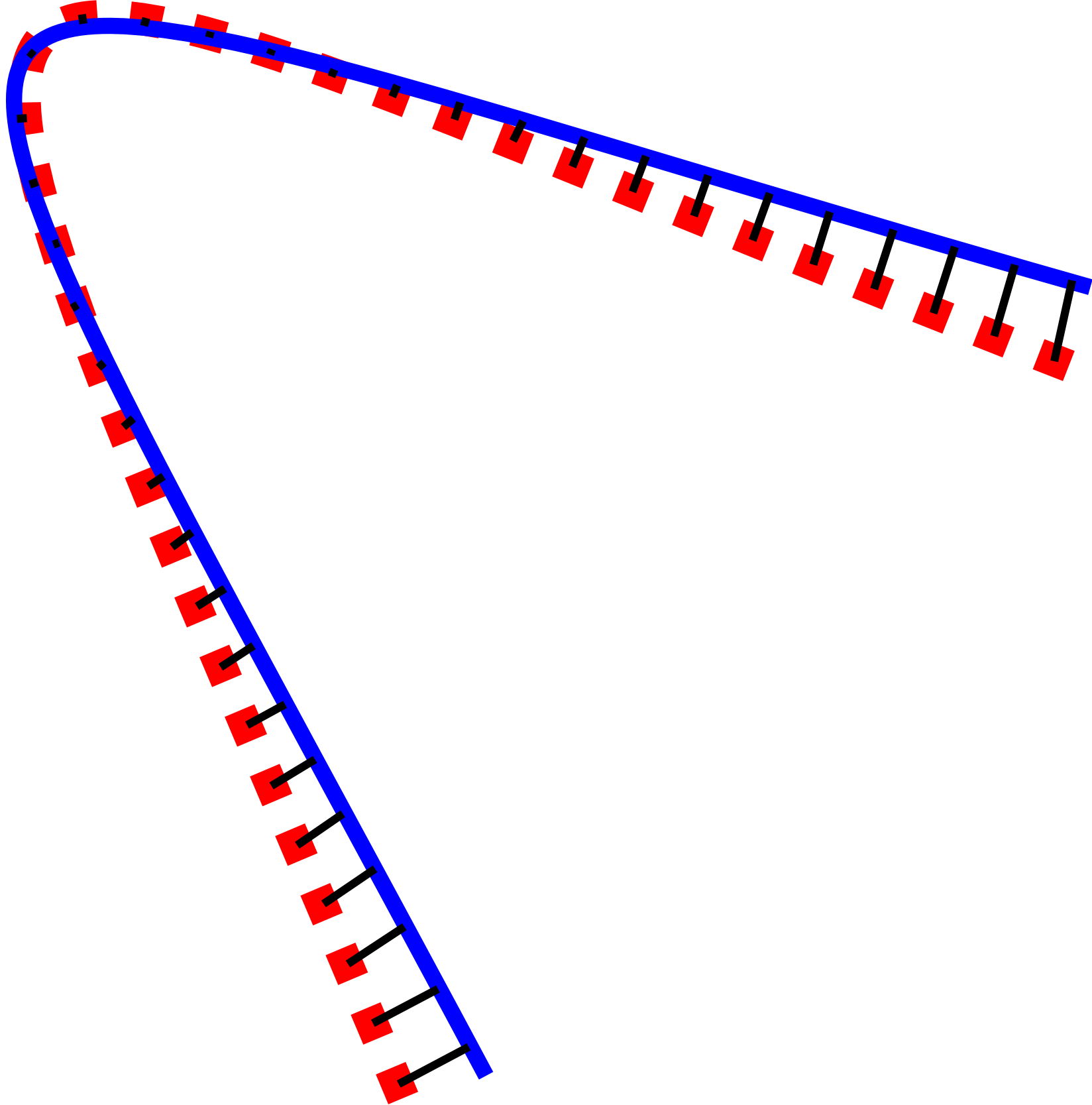






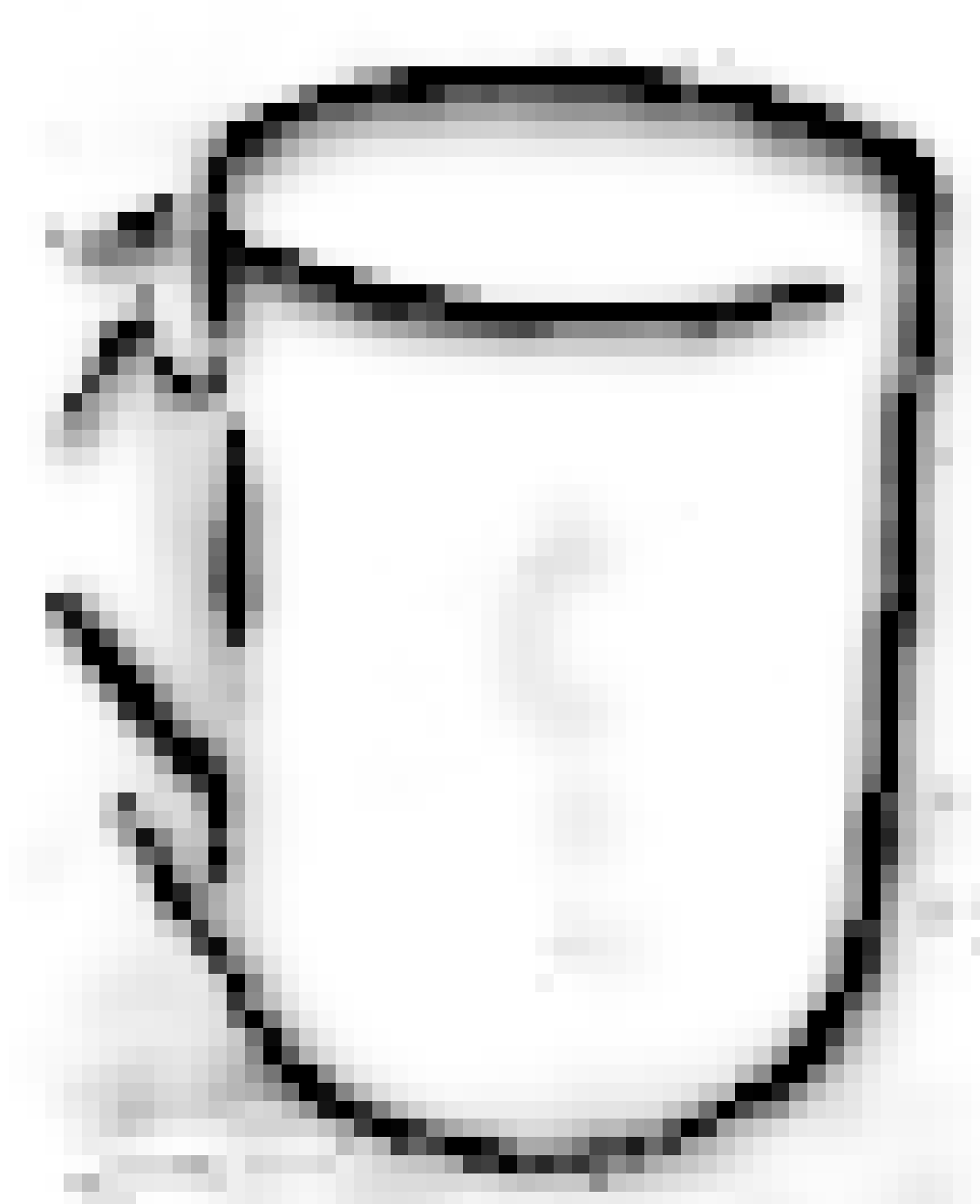








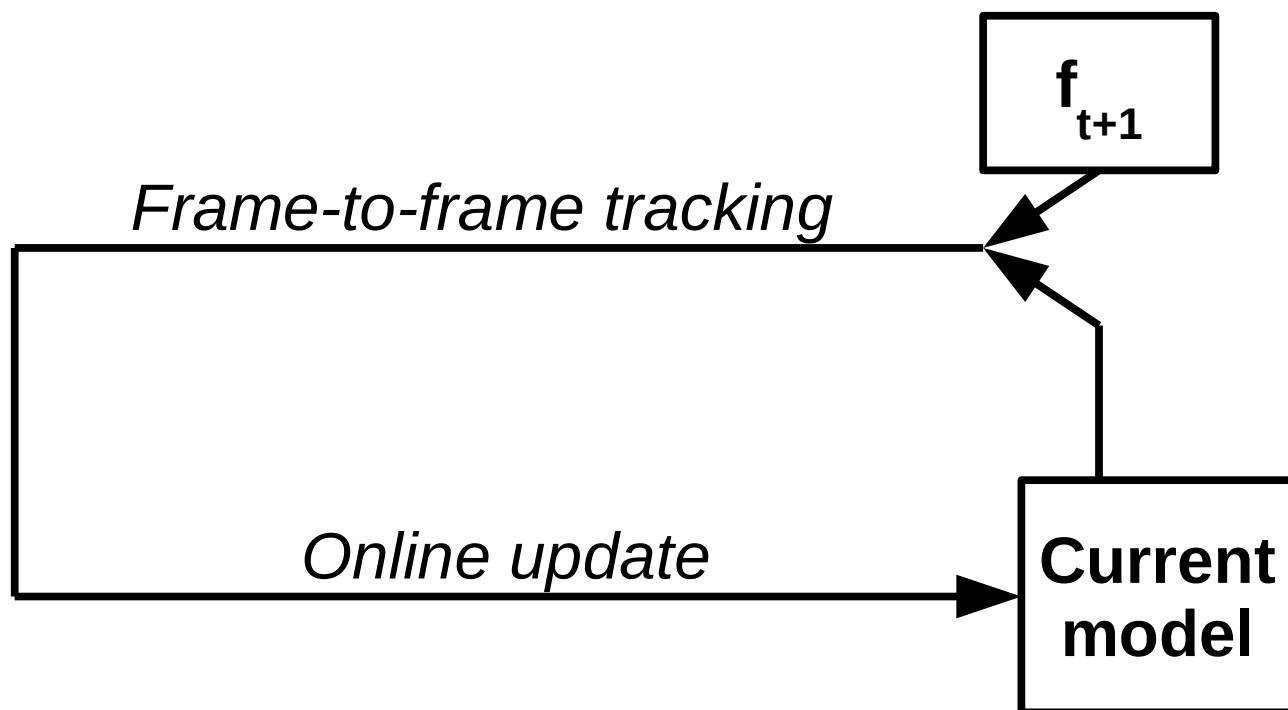


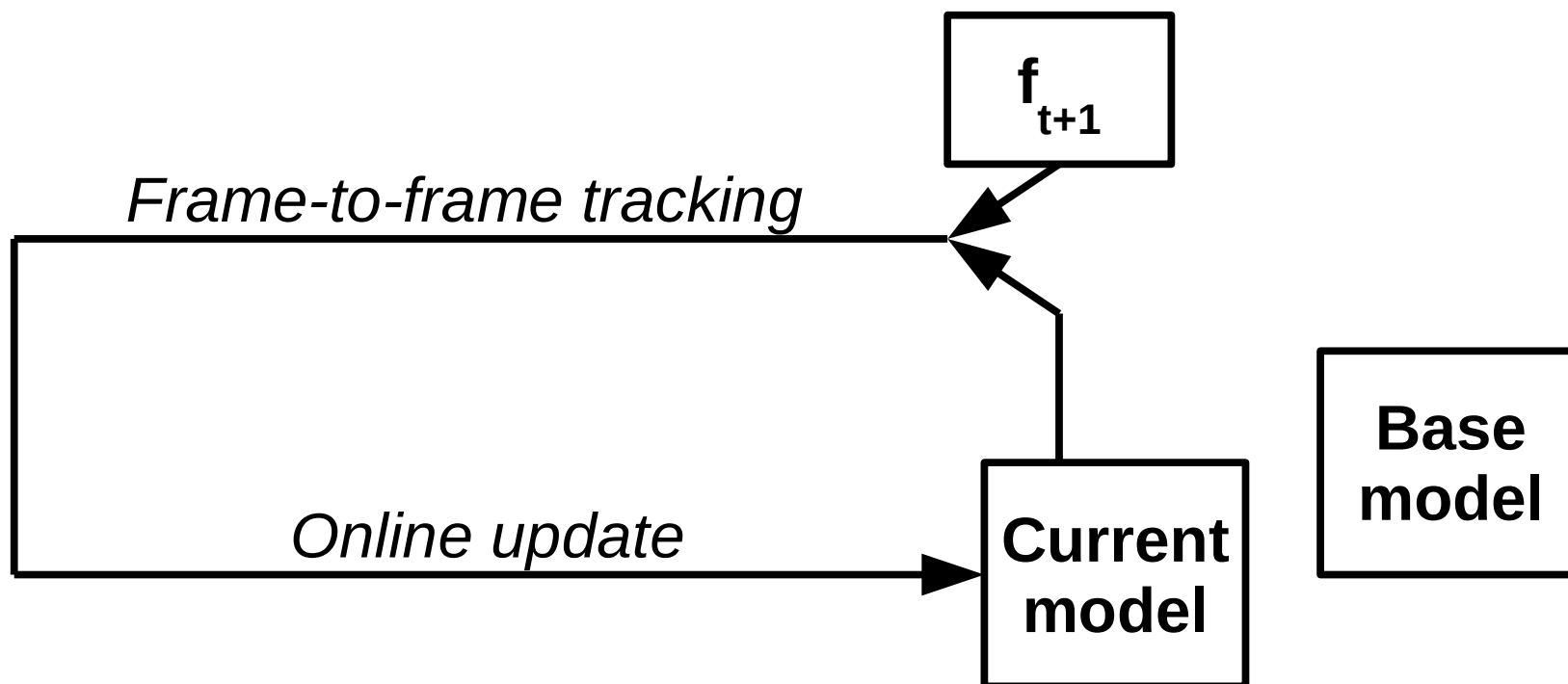


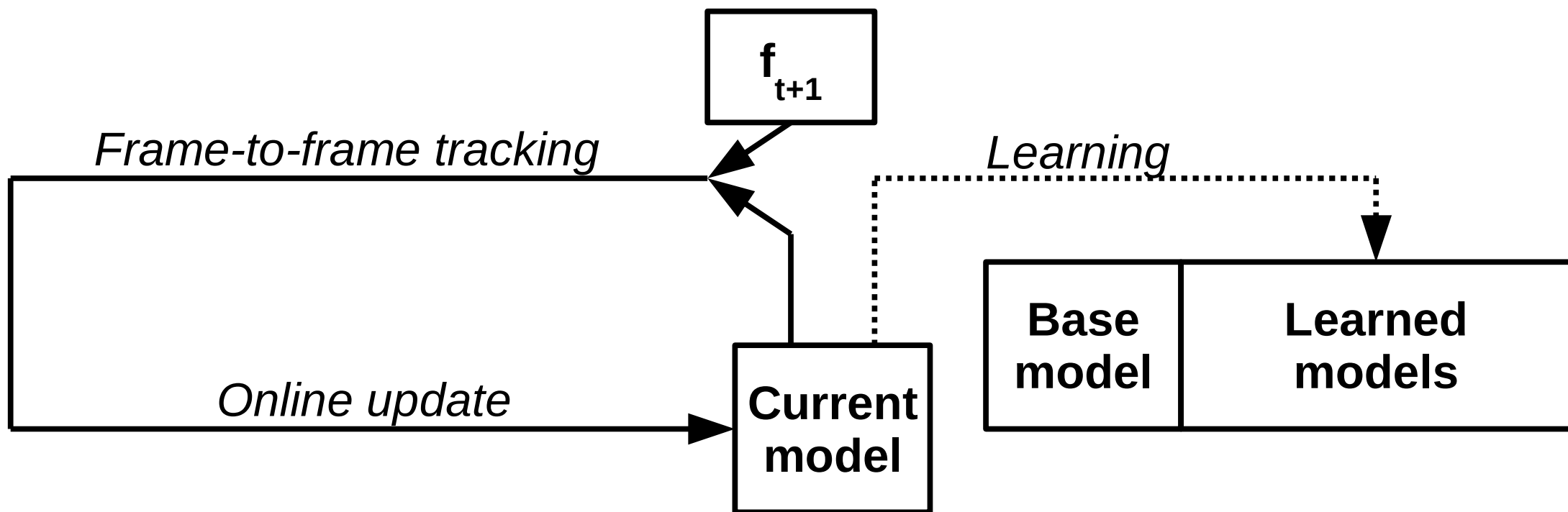


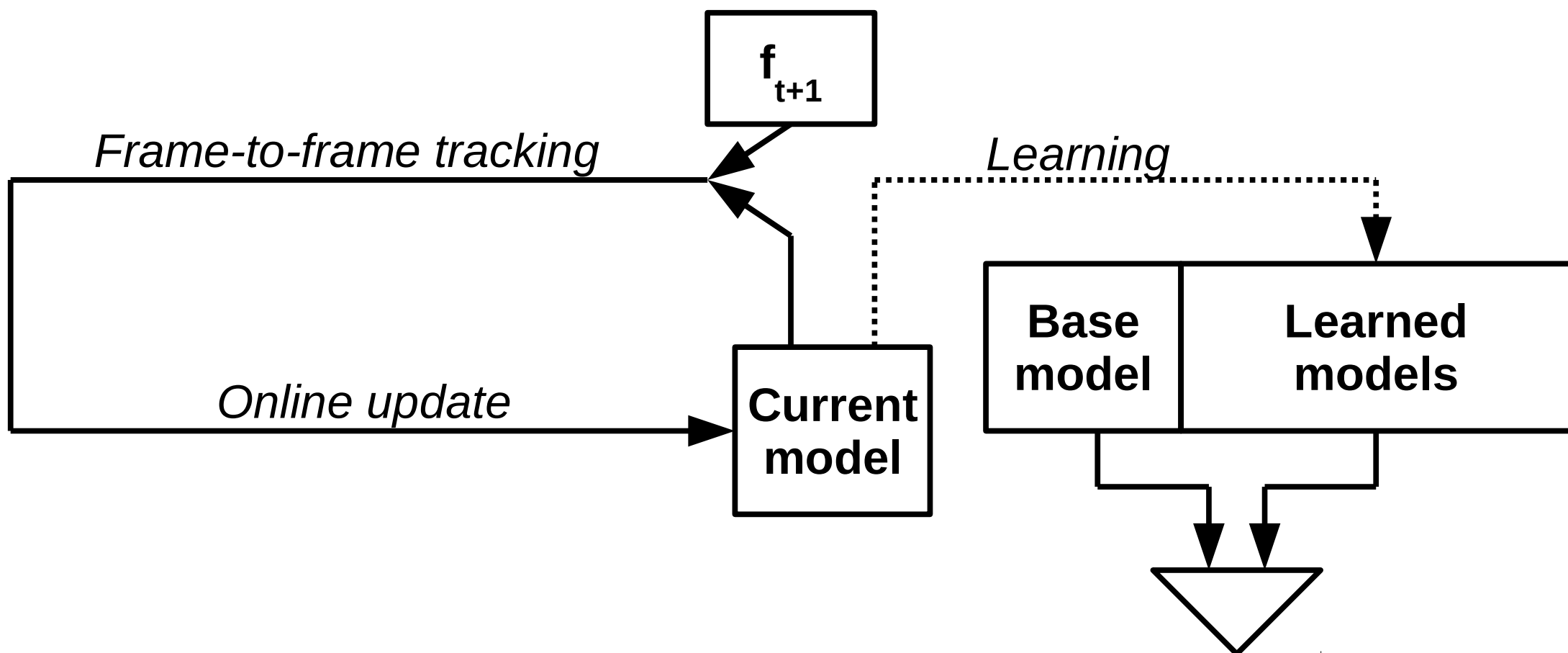


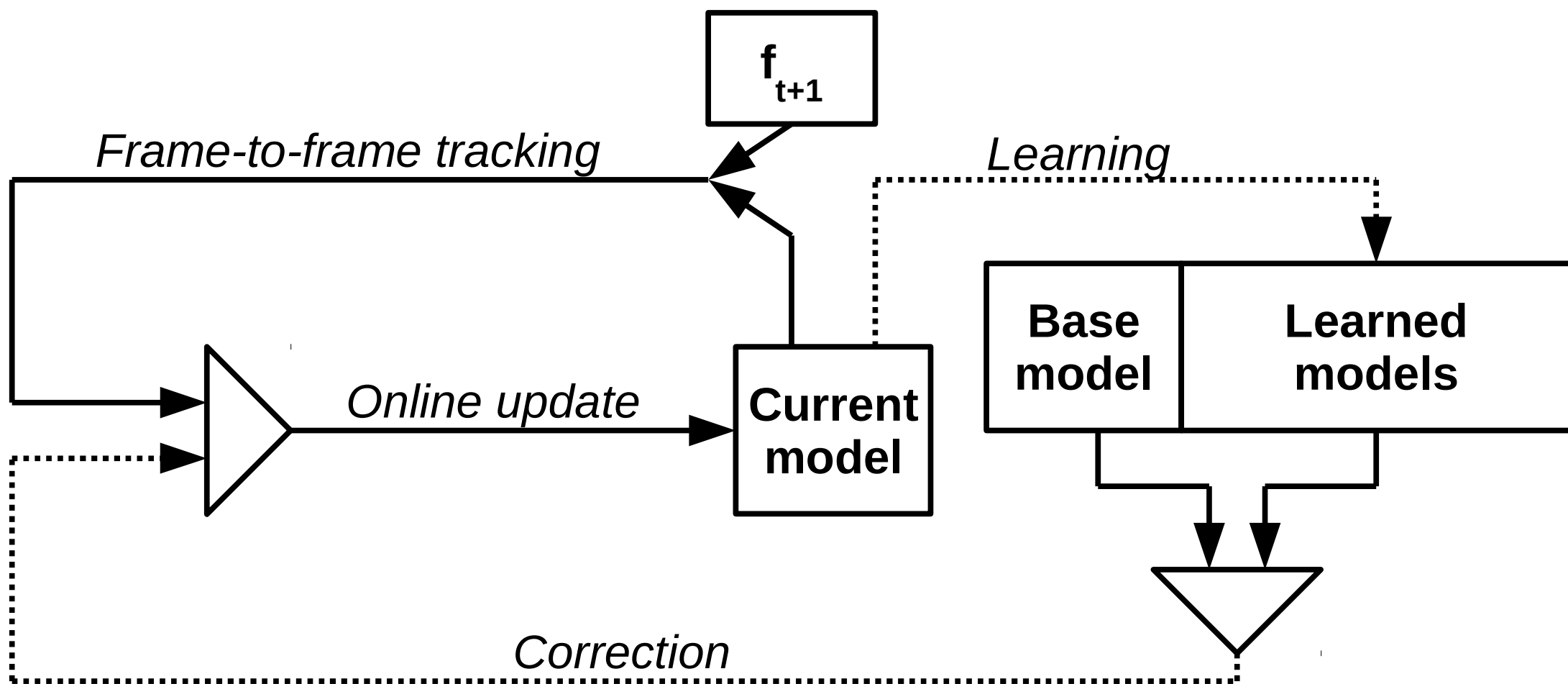
**Current  
model**











POSITION  
DIFFERENCE

TRACK

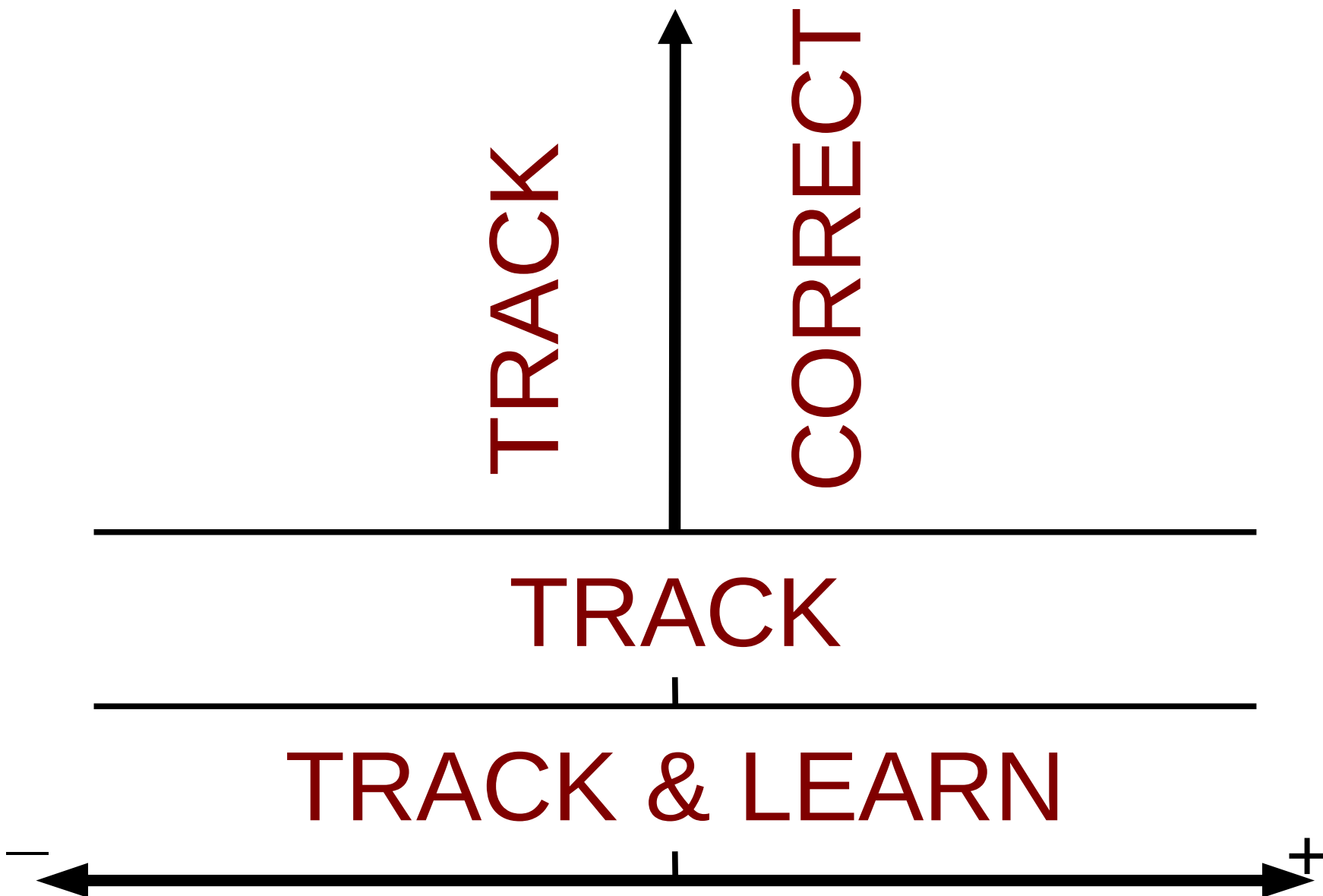
CORRECT

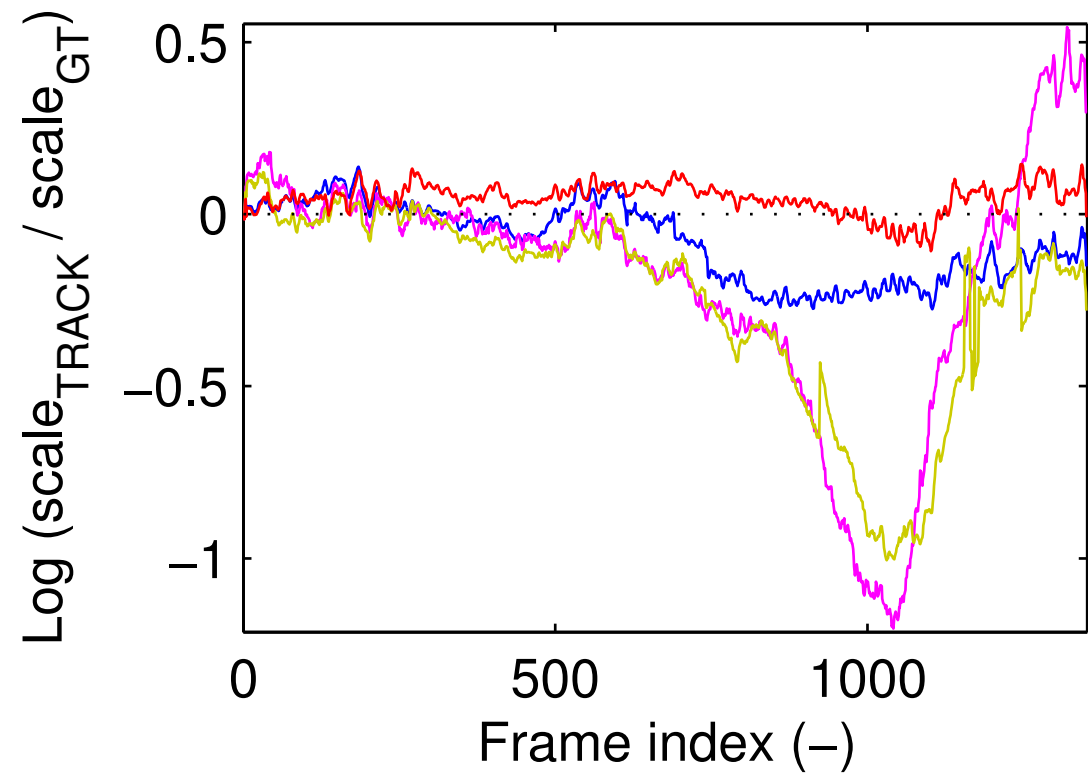
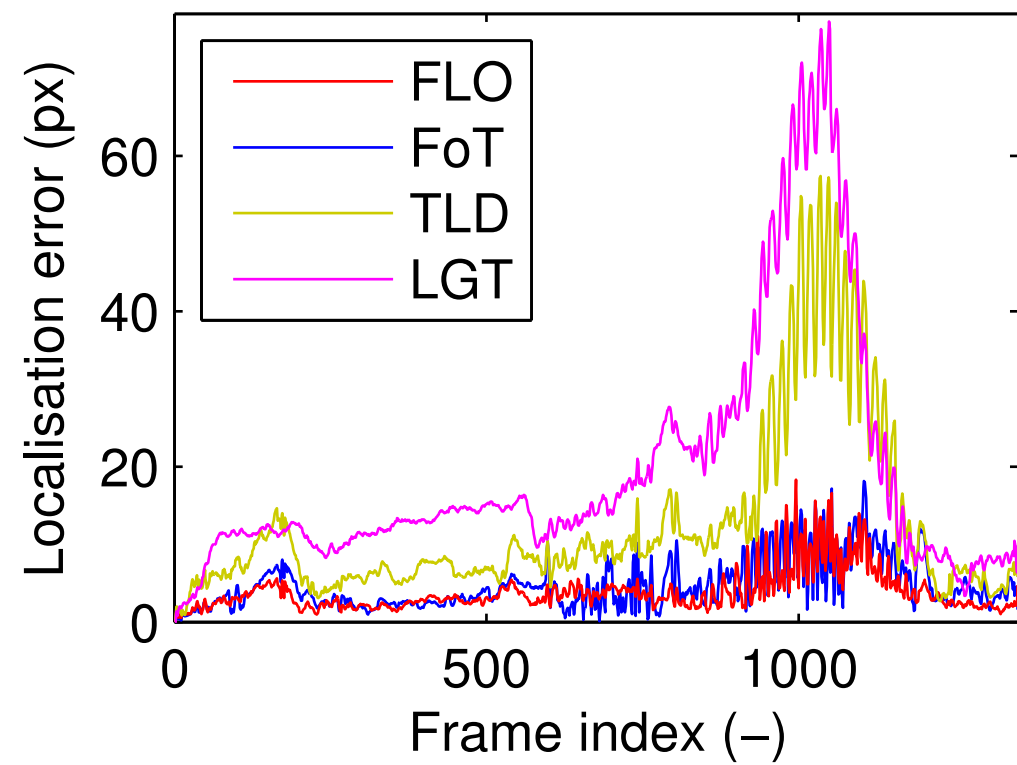
TRACK

TRACK & LEARN

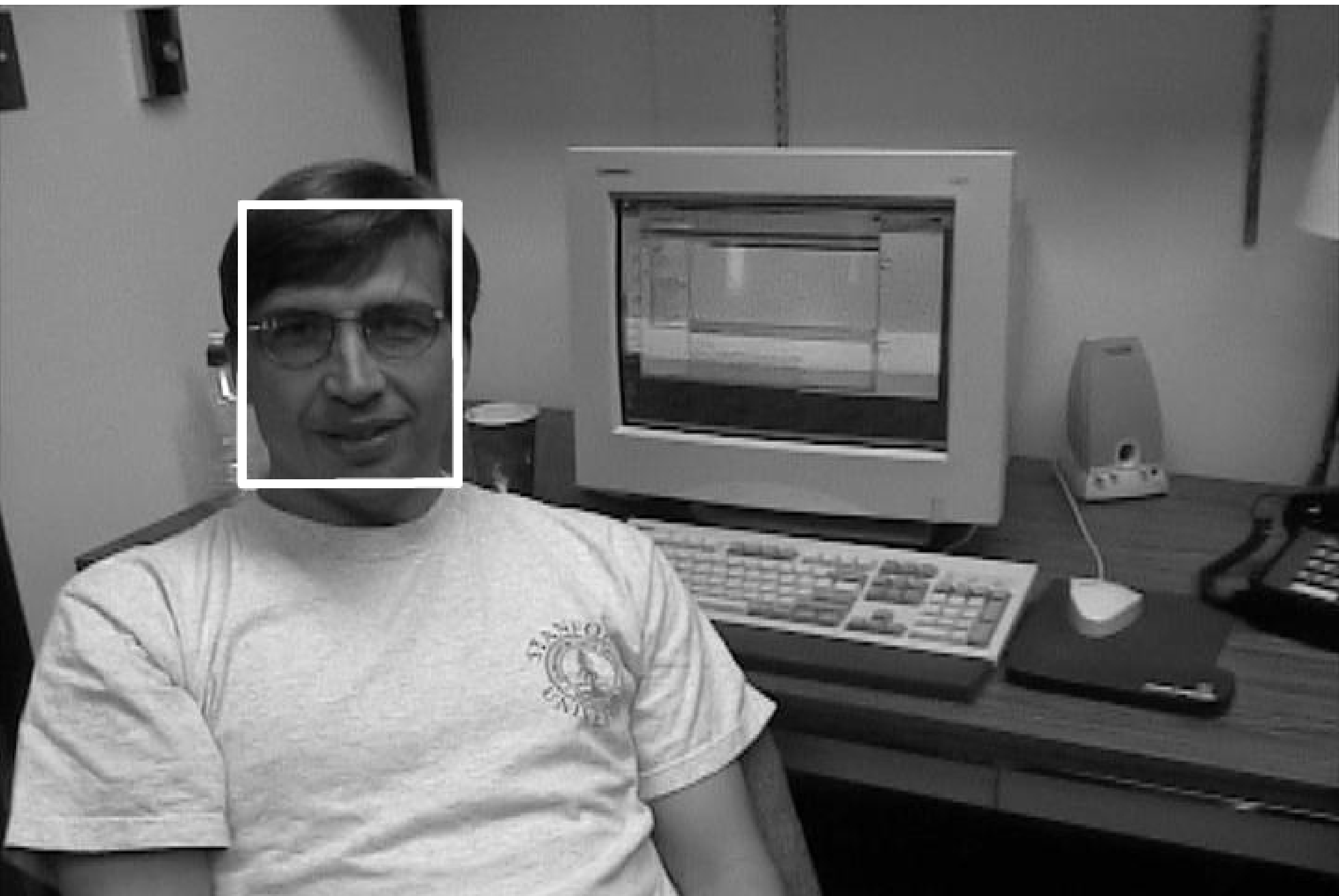
Current  
model

QUALITY OF  
CORRECTION

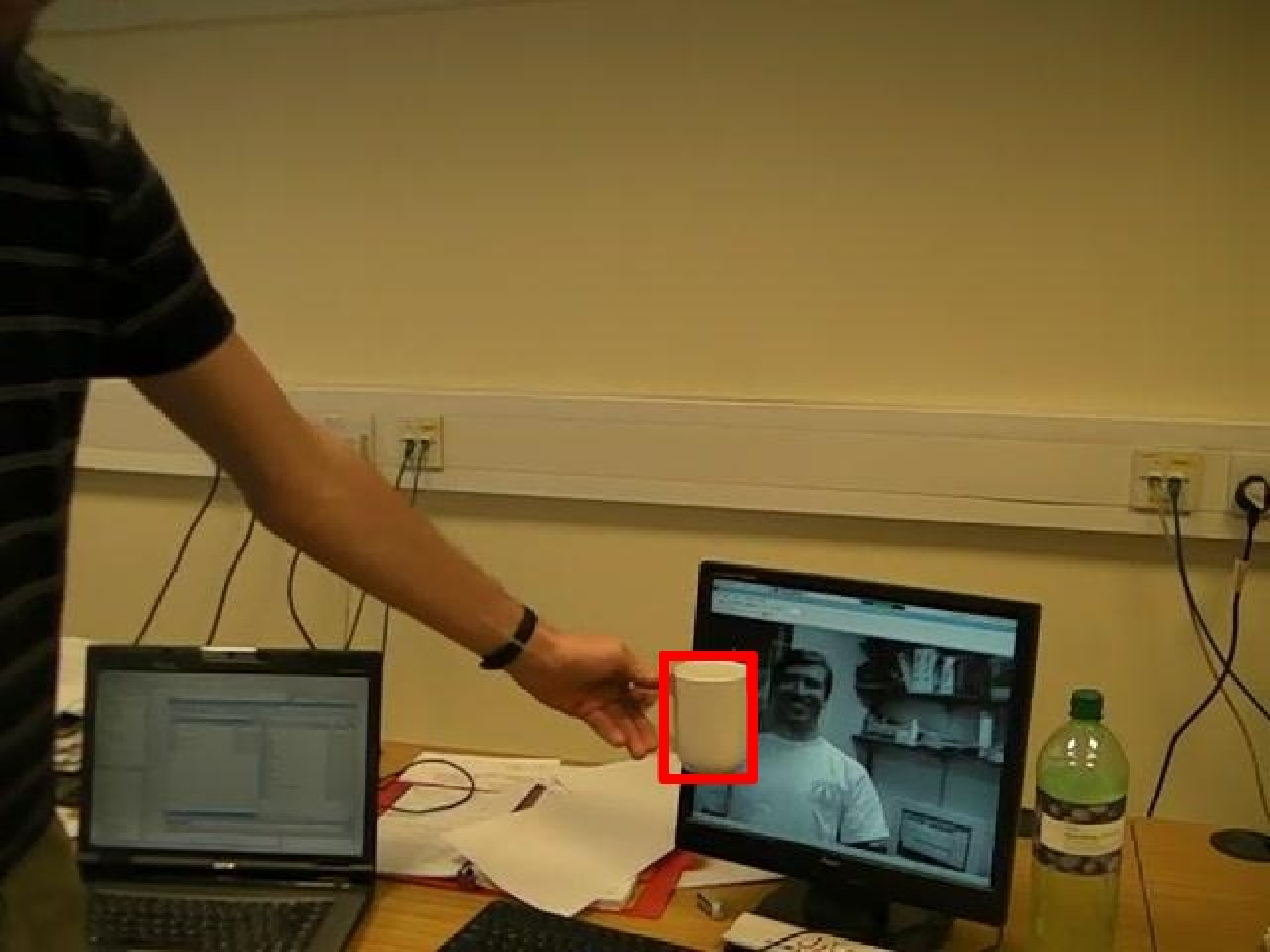


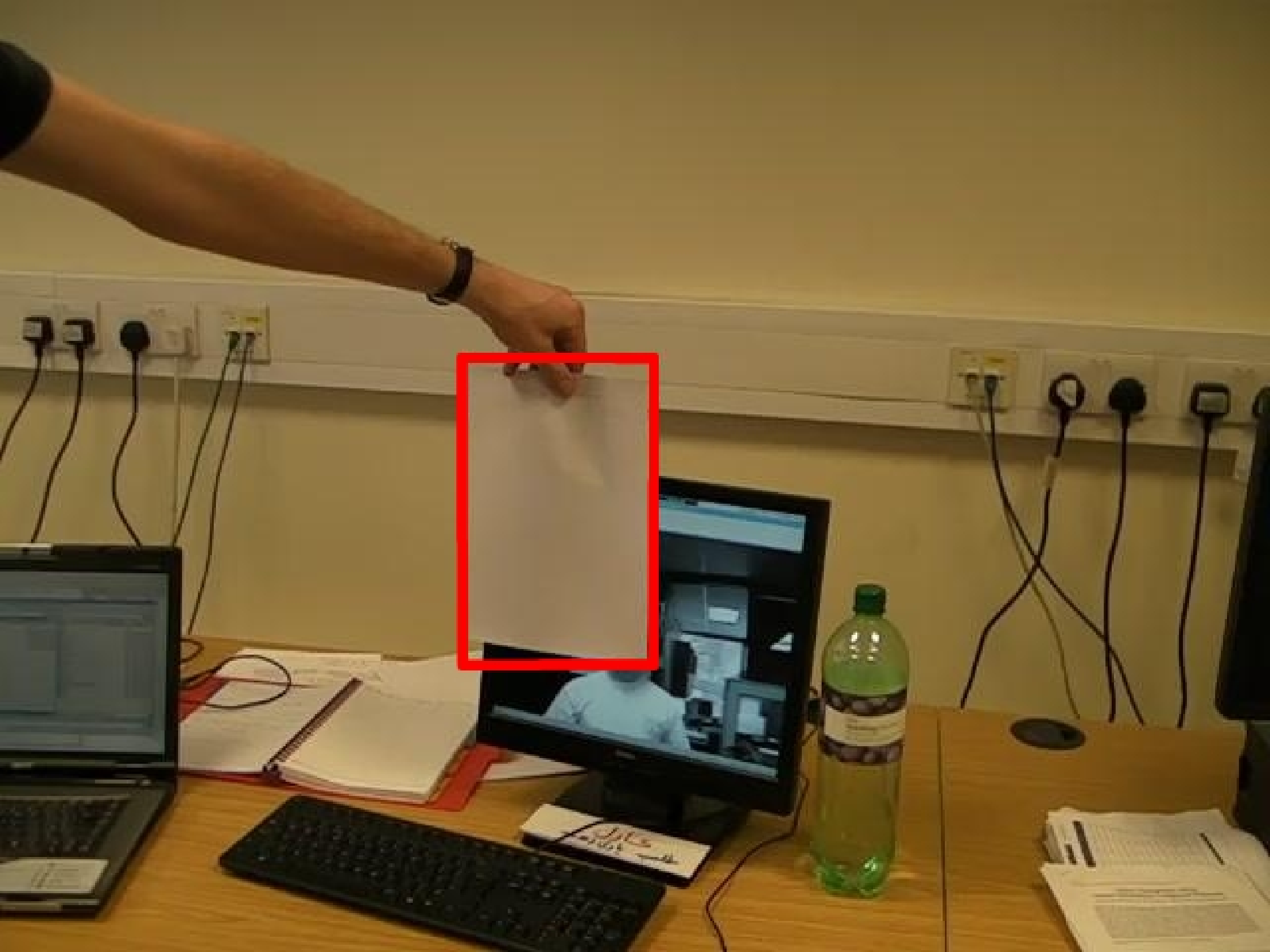


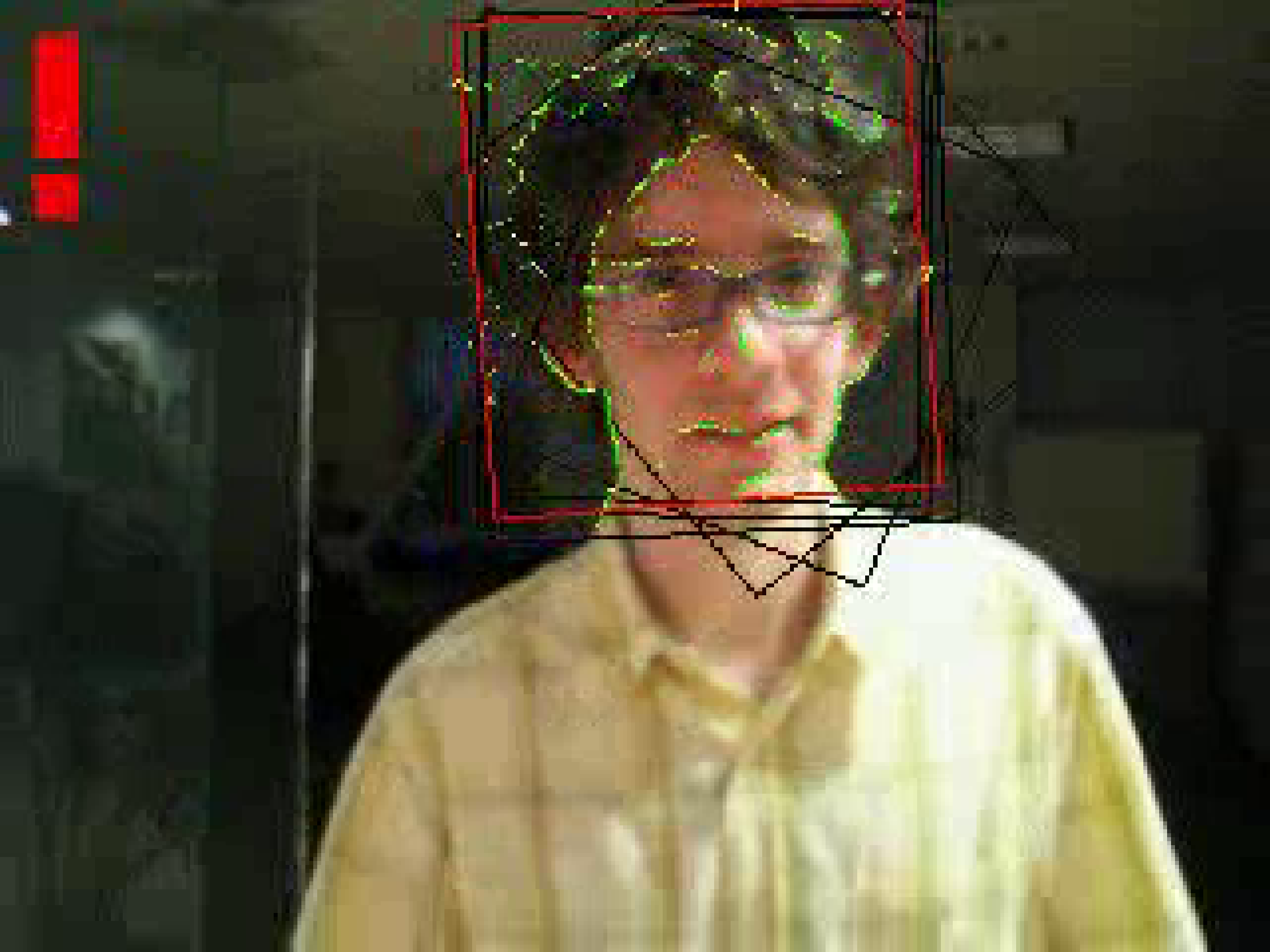




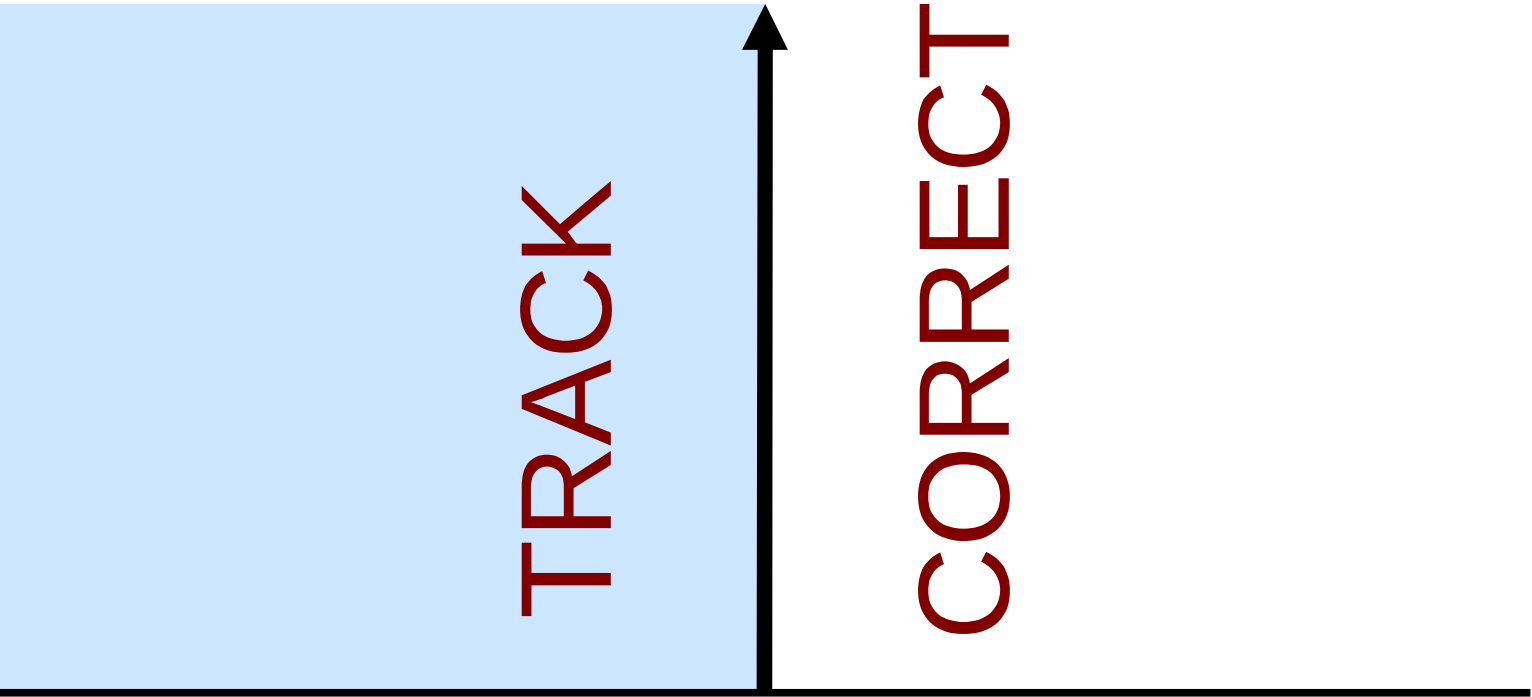






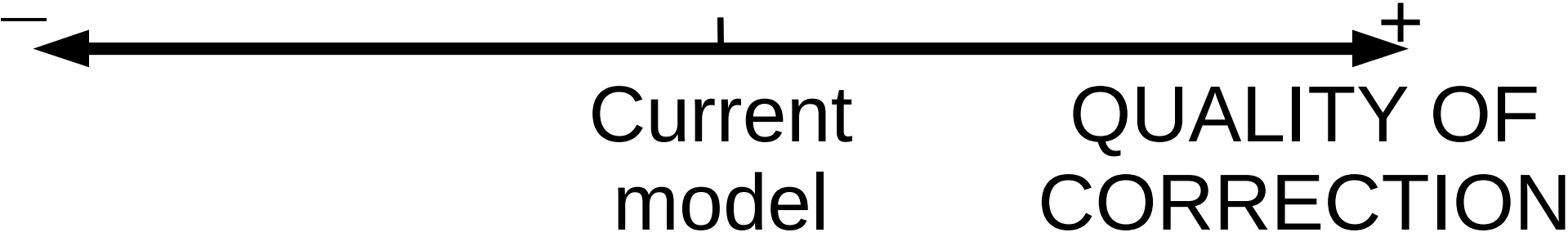


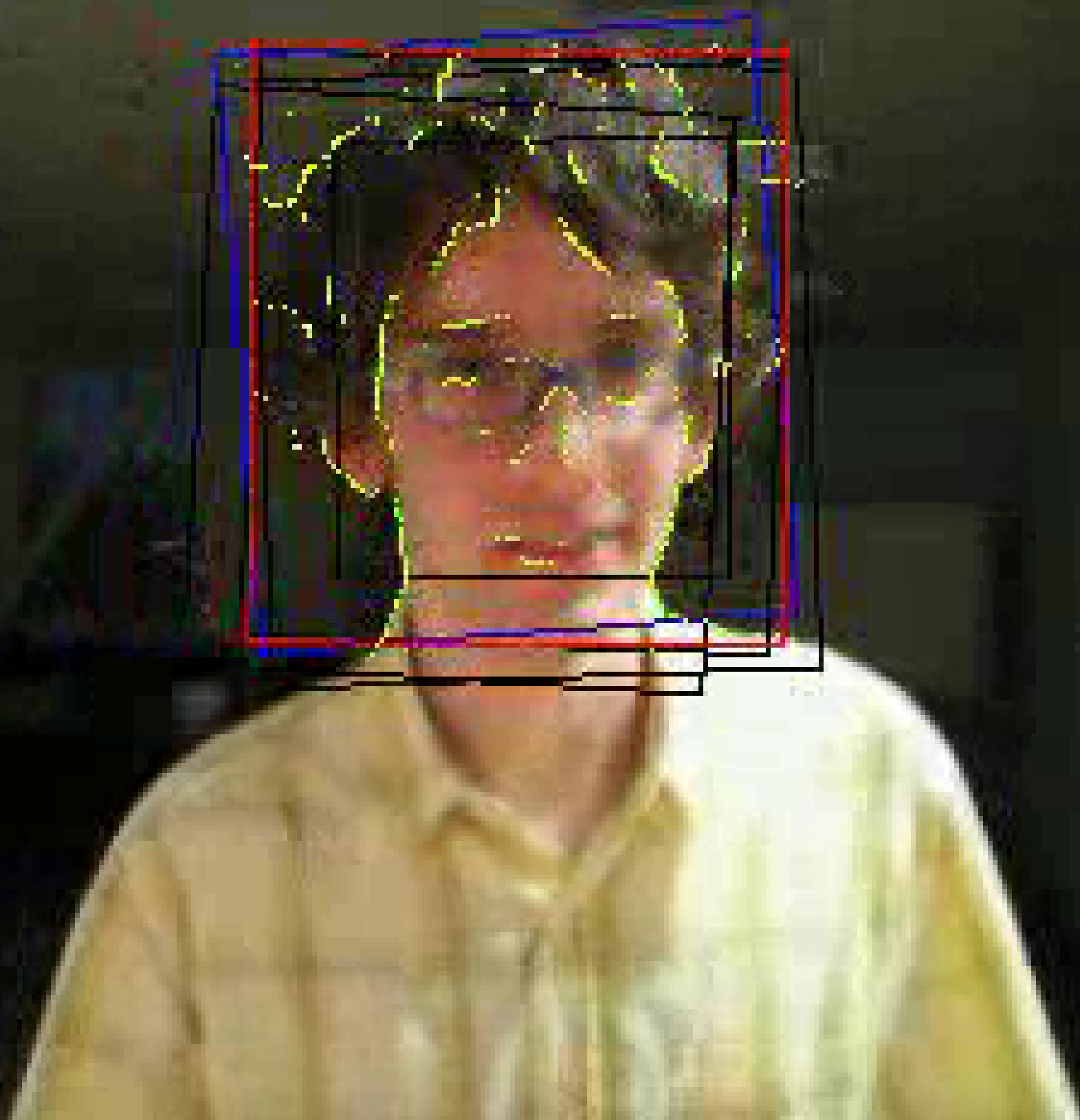
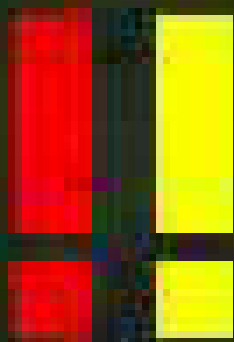
POSITION  
DIFFERENCE



TRACK

TRACK & LEARN

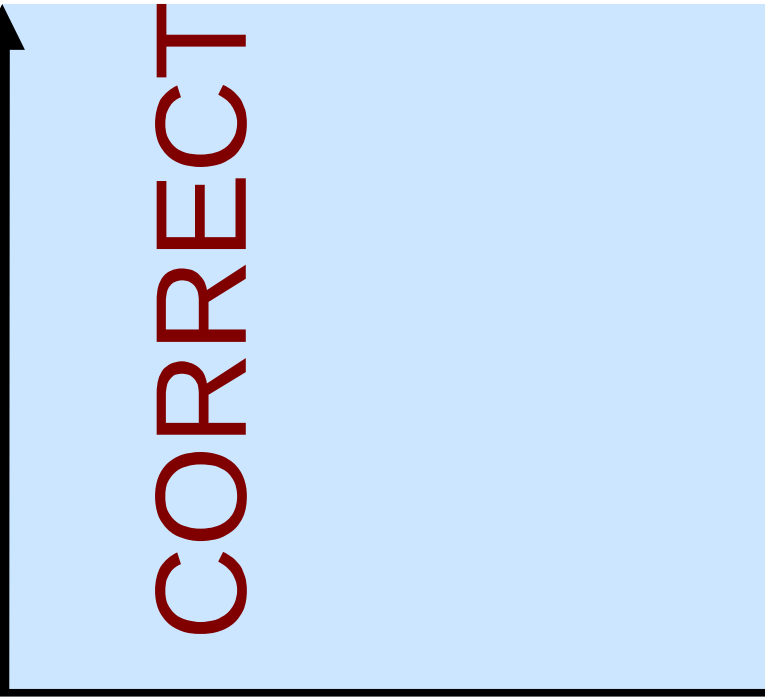




POSITION  
DIFFERENCE

TRACK

CORRECT



TRACK

TRACK & LEARN



Current  
model

QUALITY OF  
CORRECTION





POSITION  
DIFFERENCE

TRACK

CORRECT



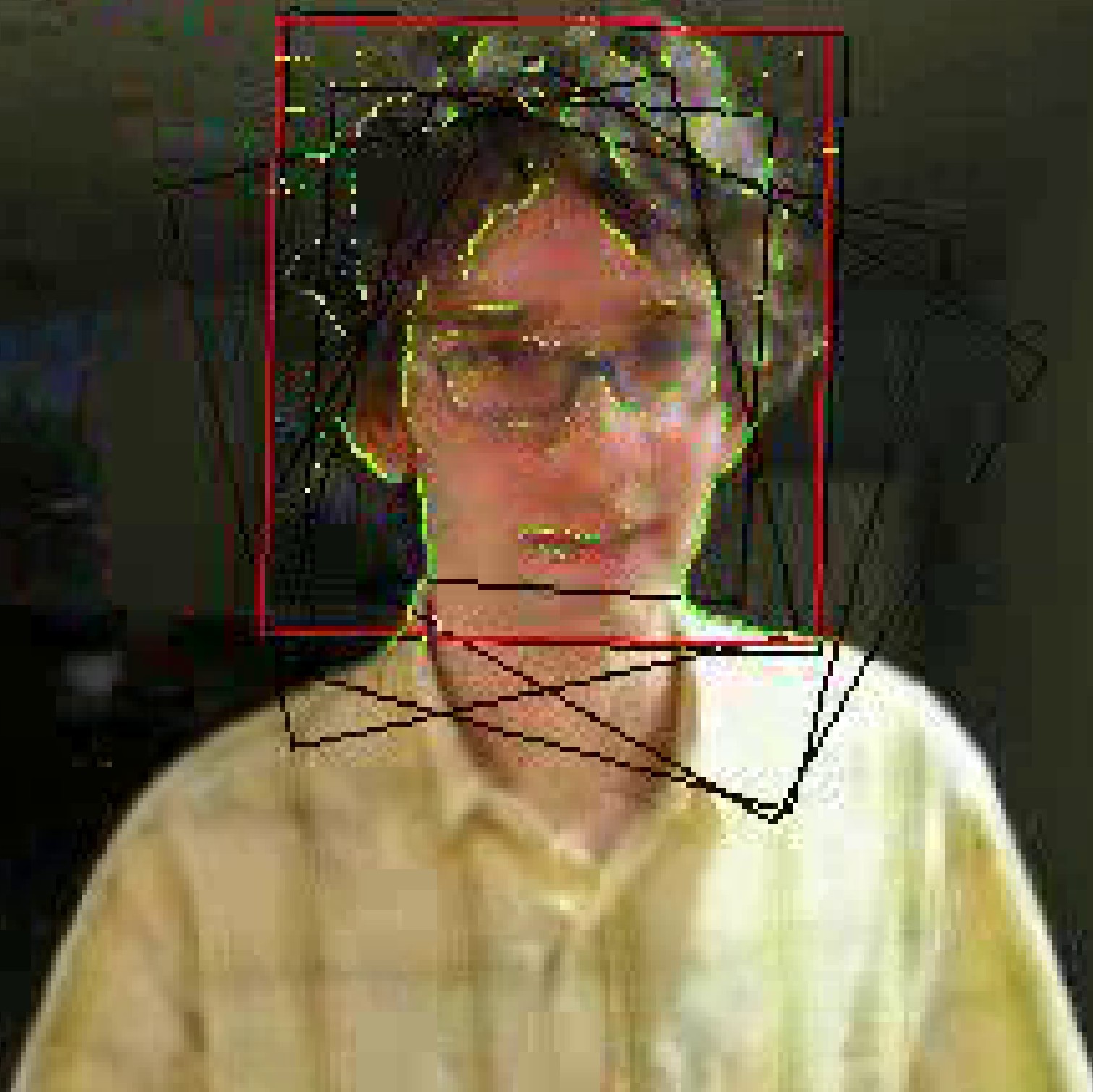
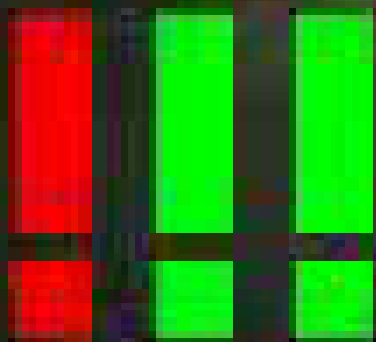
TRACK

TRACK & LEARN



Current  
model

QUALITY OF  
CORRECTION



POSITION  
DIFFERENCE

TRACK

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TRACK & LEARN

Current  
model

QUALITY OF  
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