

Introduction

- ▶ Automatic Sign Language Recognition (ASLR):
 - ▷ no special hardware, inhomogeneous background, occlusions, ...
 - ▷ important features for ASLR:
 - hand-shapes
 - orientation and movement of the hands, arms or body
 - facial expressions
 - lip-patterns
 - ▷ tracking of head and hands is hard
 - ▷ problems:
 - hands are signing in front of the face
 - hands are moving very fast and abrupt
- ▶ We avoid preliminary decisions and propose to use the same techniques that are successfully applied in automatic speech recognition (ASR).

Tracking

- ▶ Tracking is necessary to extract sign language features.
(hand position relative to the body, facial expressions, ...)
- ▶ Possible features
 - ▷ motion, background subtraction
 - ▷ skin, skin-motion
 - ▷ template matching, eigenfaces
- ▶ Environment problems: features are not correct and can only be used to produce candidate regions



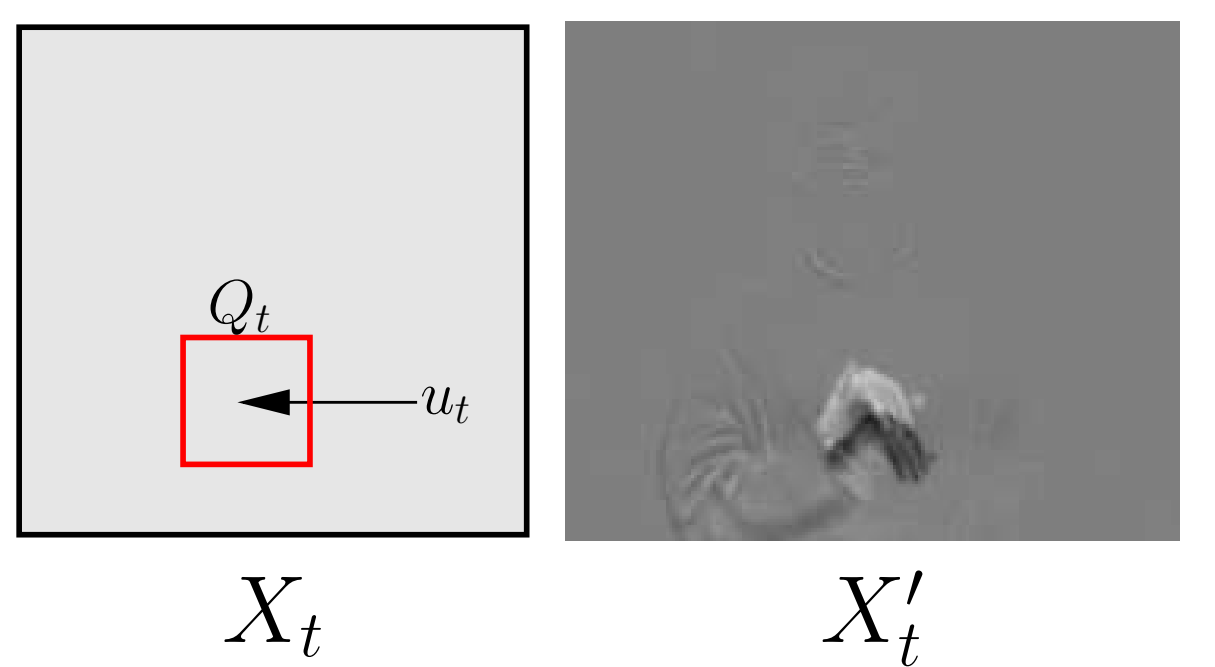
Scoring Functions

- ▶ Local scoring functions
 - ▷ example: track moving parts, i.e. maximize motion

$$u := (x, y)$$

$$Q := \left\{ (x, y) : \begin{matrix} -w \leq x \leq w, \\ -h \leq y \leq h \end{matrix} \right\}$$

$$Q_t := \{u_t + u : u \in Q\}$$

$$X'_t[u] := X_t[u] - X_{t-1}[u]$$


$$\operatorname{argmax}_{u_1^T} \left\{ \sum_{t=1}^T \left(\sum_{u \in Q_t} (X'_t[u])^2 + \sum_{u \in Q_{t-1}} (X'_{t-1}[u])^2 - \lambda \|u_t - u_{t-1}\|^2 \right) \right\}$$

- ▶ Additional scoring functions/features:
 - ▷ skin color
 - ▷ Eigenfaces
 - ▷ special distance measures like Tangent distance or IDM
- ▶ Example: combination of skin color score s_c and Eigenfaces s_f for head tracking

$$s(t, x, y) = (1 - w) \cdot s_c(t, x, y) + w \cdot s_f(t, x, y)$$



- ▶ Jump penalty functions
 - ▷ Euclidean distance: $\mathcal{T}(x', y', x, y) = \lambda \cdot \sqrt{(x - x')^2 + (y - y')^2}$
 - ▷ L_1 -norm, complex motion models, ...

Dynamic Programming Tracking

- ▶ Idea: prevent taking possibly wrong local decisions
- ▶ How: tracking is done at the end of a sequence by tracking back the decisions to reconstruct the best path
- ▶ the best path is the path with the highest score wrt. a given scoring function
- ▶ 2 Steps:
 1. **Score calculation:** calculate a global score $S(t, x, y)$ and a back-pointer table B for the best tracking until time step t which ends in position (x, y)
 2. **Traceback:** reconstruct the best path $t \rightarrow (x, y)$ using S and B

1. Step: Score calculation

- ▷ local score: For each pixel (x, y) of a frame X_t at time step $t = 1, \dots, T$ a score $s(t, x, y)$ is calculated
- ▷ bookkeeping at time t of the best predecessors at time $t - 1$ within a neighborhood of position (x, y)

2. Step: Traceback

- ▷ full traceback: starts from last frame of the sequence at time T using $c_T = \operatorname{argmax}_{x, y} S(T, x, y)$
- ▷ best tracking center c at time $t - 1$: $c_{t-1} = B(t, c_t = (x', y'))$

Recursive equation for DPT:

$$S(t, x, y) = \max_{x', y' \in M(x, y)} \{ (S(t-1, x', y') - \mathcal{T}(x', y', x, y)) + s(t, x, y) \}$$

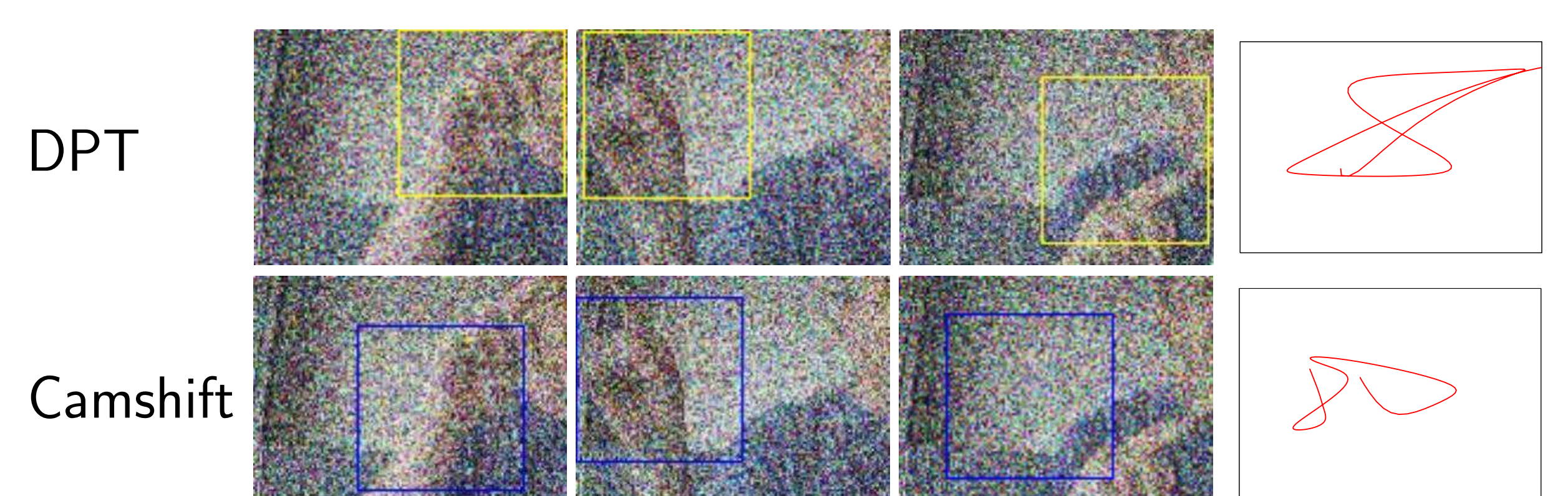
$$B(t, x, y) = \operatorname{argmax}_{x', y' \in M(x, y)} \{ (S(t-1, x', y') - \mathcal{T}(x', y', x, y)) \}$$

$M(x, y)$ is the set of possible predecessors of point (x, y)

$\mathcal{T}(x', y', x, y)$ is the jump-penalty from point (x', y') in the predecessor image to point (x, y) in the current image

Experimental Results

- ▶ Tracking under noisy circumstances



- ▶ Successful tracking of head and dominant hand for ASLR



Conclusion

- ▶ if no pruning is used, the optimal path is guaranteed to be found
- ▶ the proposed tracking algorithm enables to track a target disregarding information gaps (e.g. due to occlusions)
- ▶ advantages under noisy circumstances

Outlook

- ▶ integrate further scoring functions which don't require any priori knowledge of the target
- ▶ combine tracking and recognition
- ▶ extend tracking framework:
 - ▷ multiple objects
 - ▷ automatic traceback
- ▶ apply tracking in different areas, e.g. PETS