Real-time Arm Skeleton Tracking and Gesture Inference Tolerant to Missing Wearable Sensors

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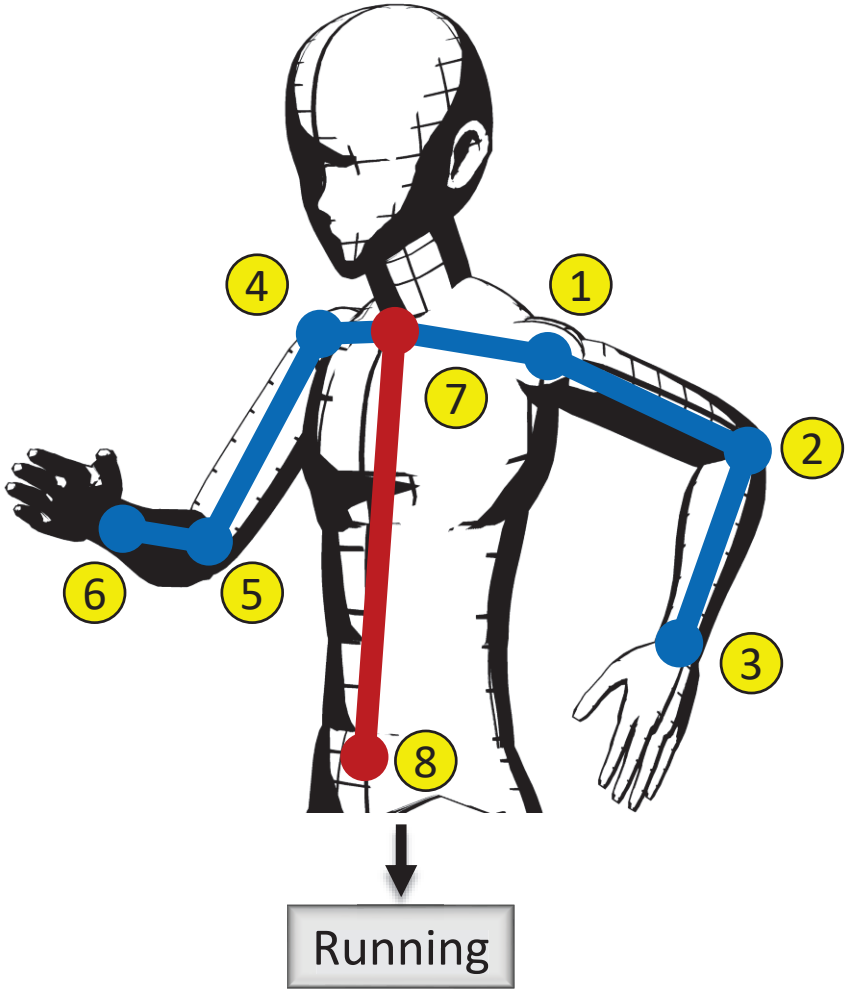
Understanding Human Arm Motions



- **How is the arm moving?**

  Skeleton tracking

- **What is the meaning of this arm motion?**

  Motion inference

Elderly Care

Elderly diseases

- Parkinson
- Alzheimer

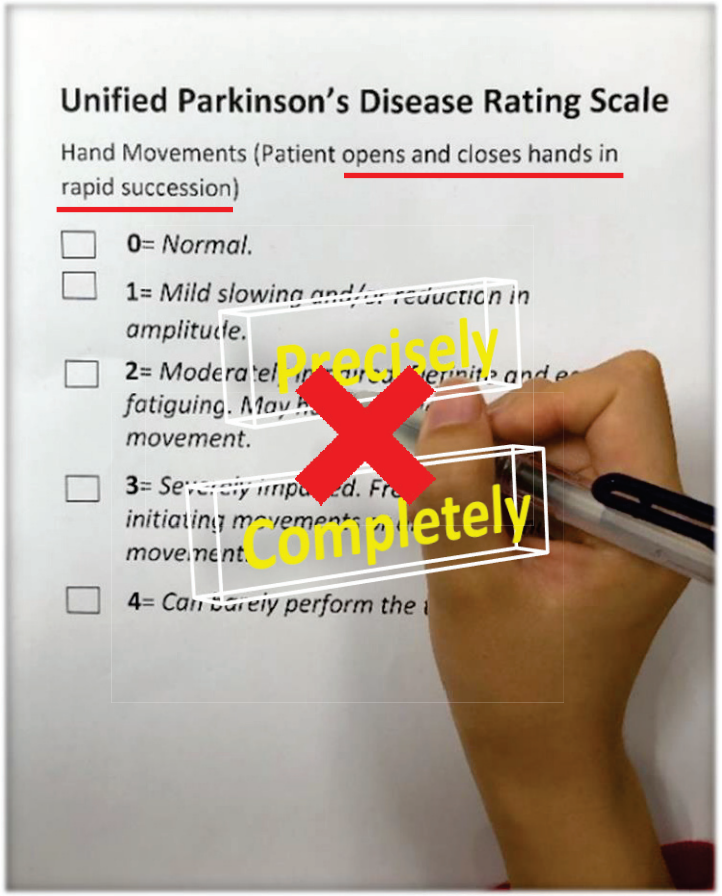
Problems with arm

- Slow motion
- Repeated motion
- Instability
- …

Last treatment

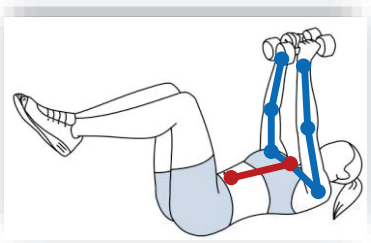
Several weeks

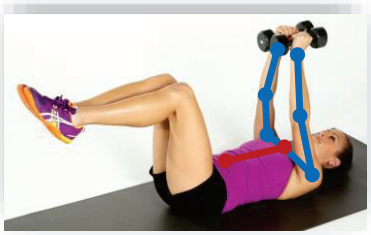
Next treatment



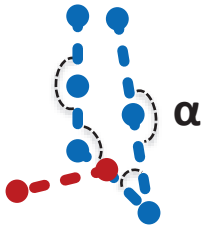
Other Applications

80USD/hour



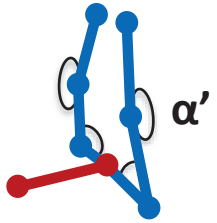


Template



Compare

User's skeleton



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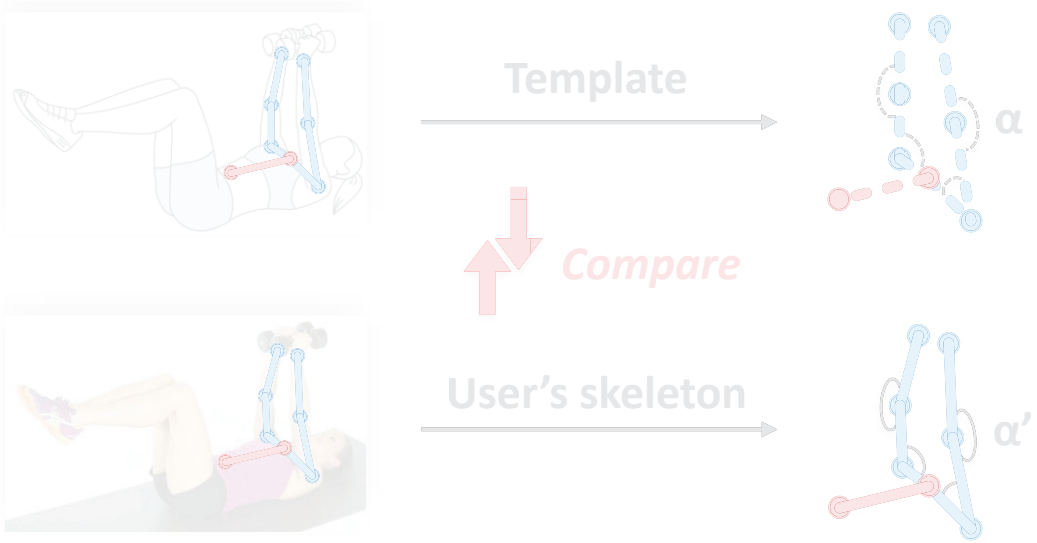
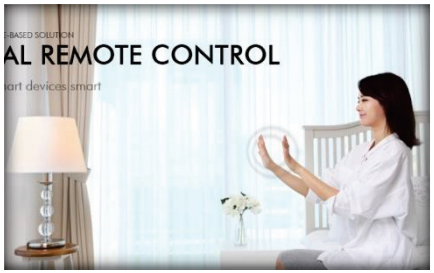
Other Applications



HCI



Smart home

Smart car

Gaming

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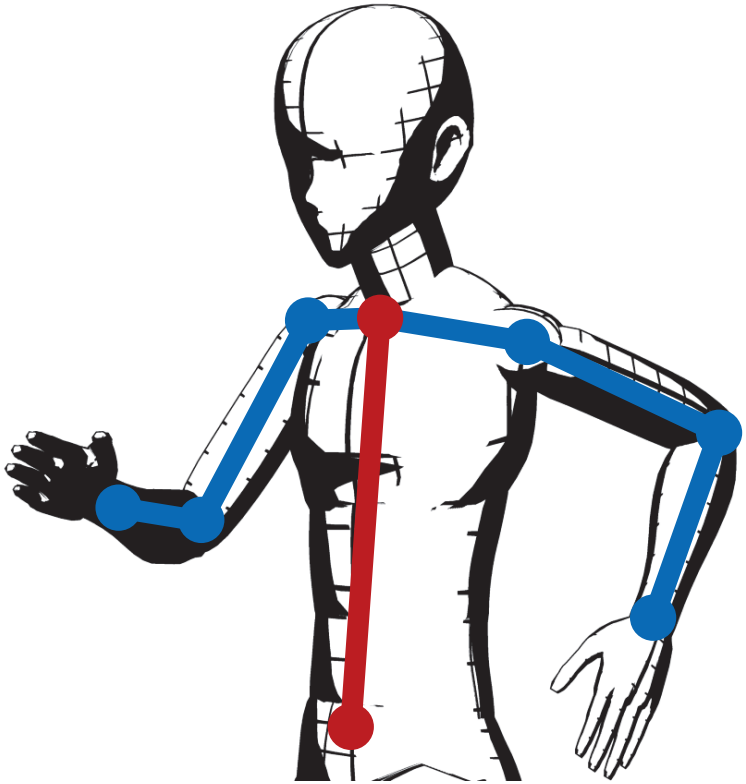
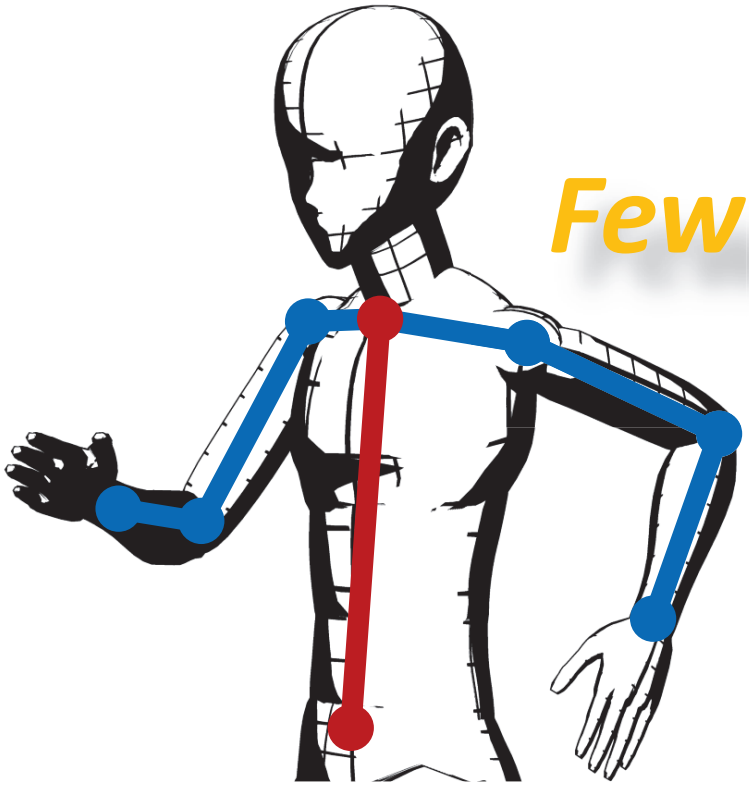
Existing Solutions



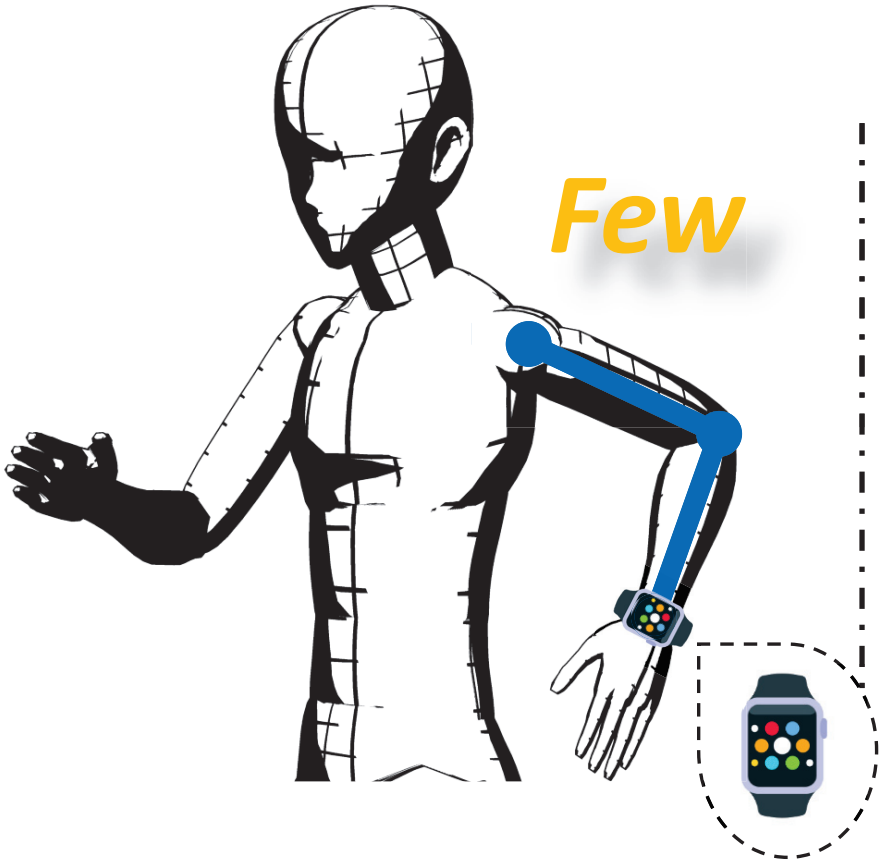
- Service coverage
- System cost
- Privacy

- Convenience
- User-friendly

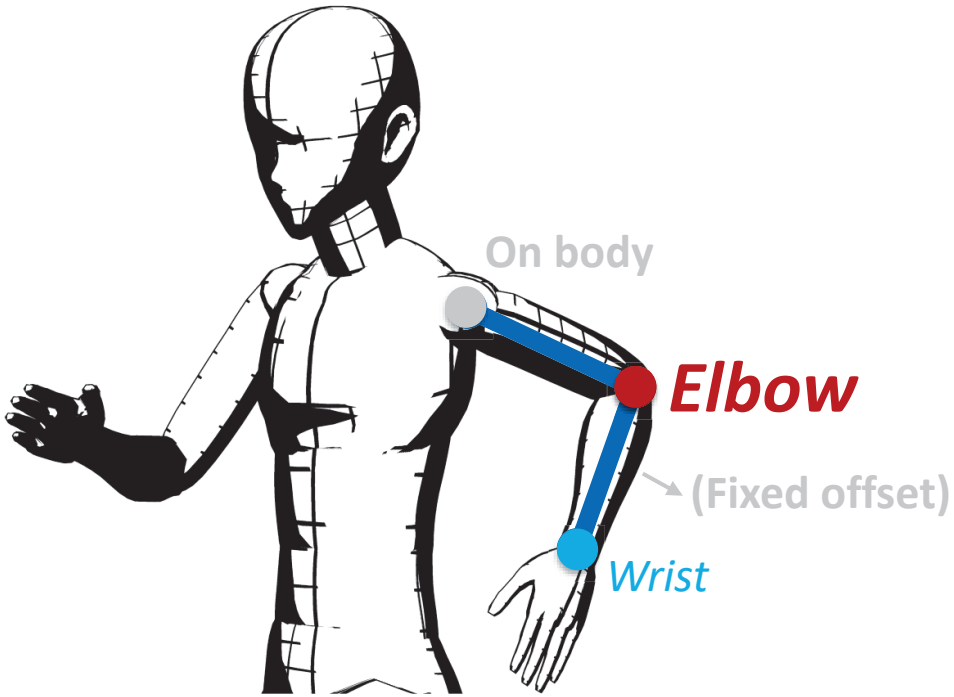
Existing Solutions



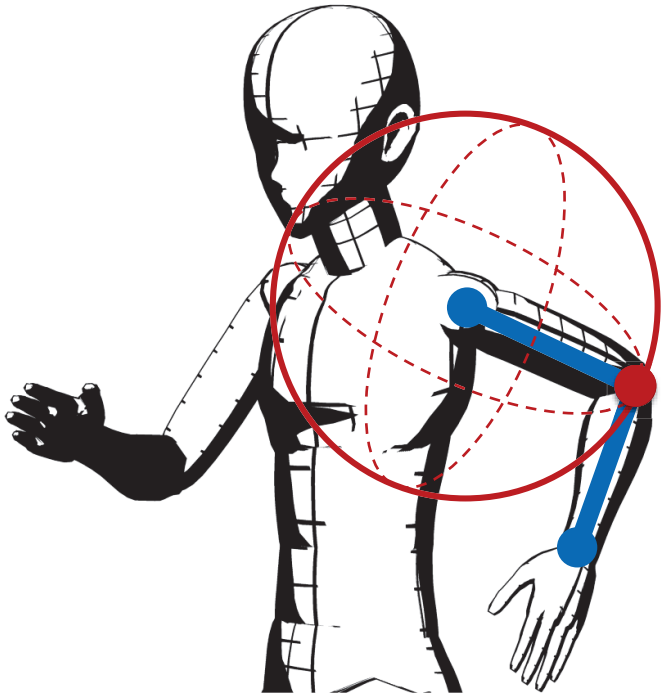
Existing Solutions



Key Problem

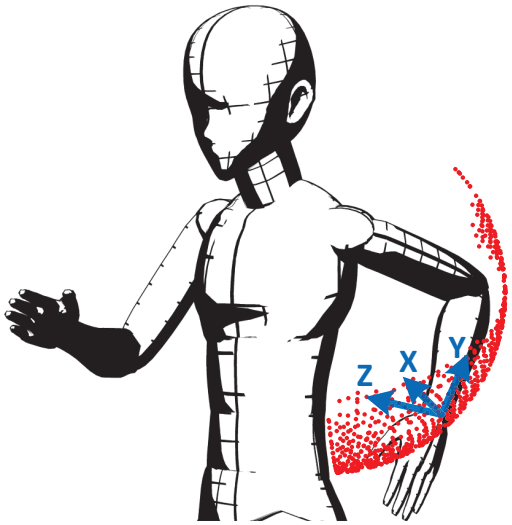


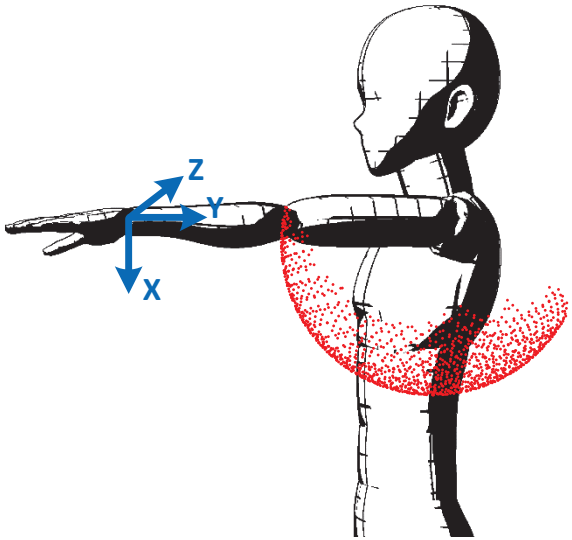
Key Problem

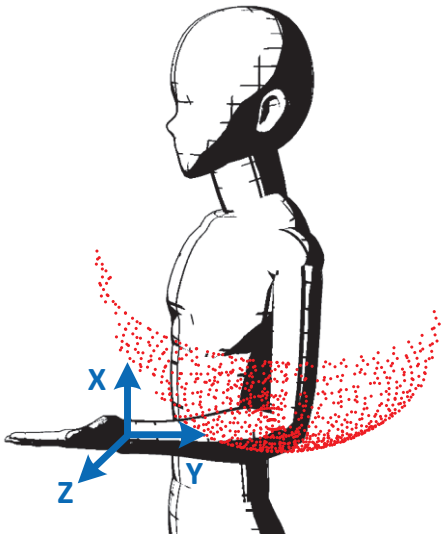


How?

Tracking Principle







For a given wrist orientation, possible elbow locations are within a limited range [1].

[1] "I am a smartwatch and I can track my user's arm", in Proc. of ACM MobiSys, 2016.

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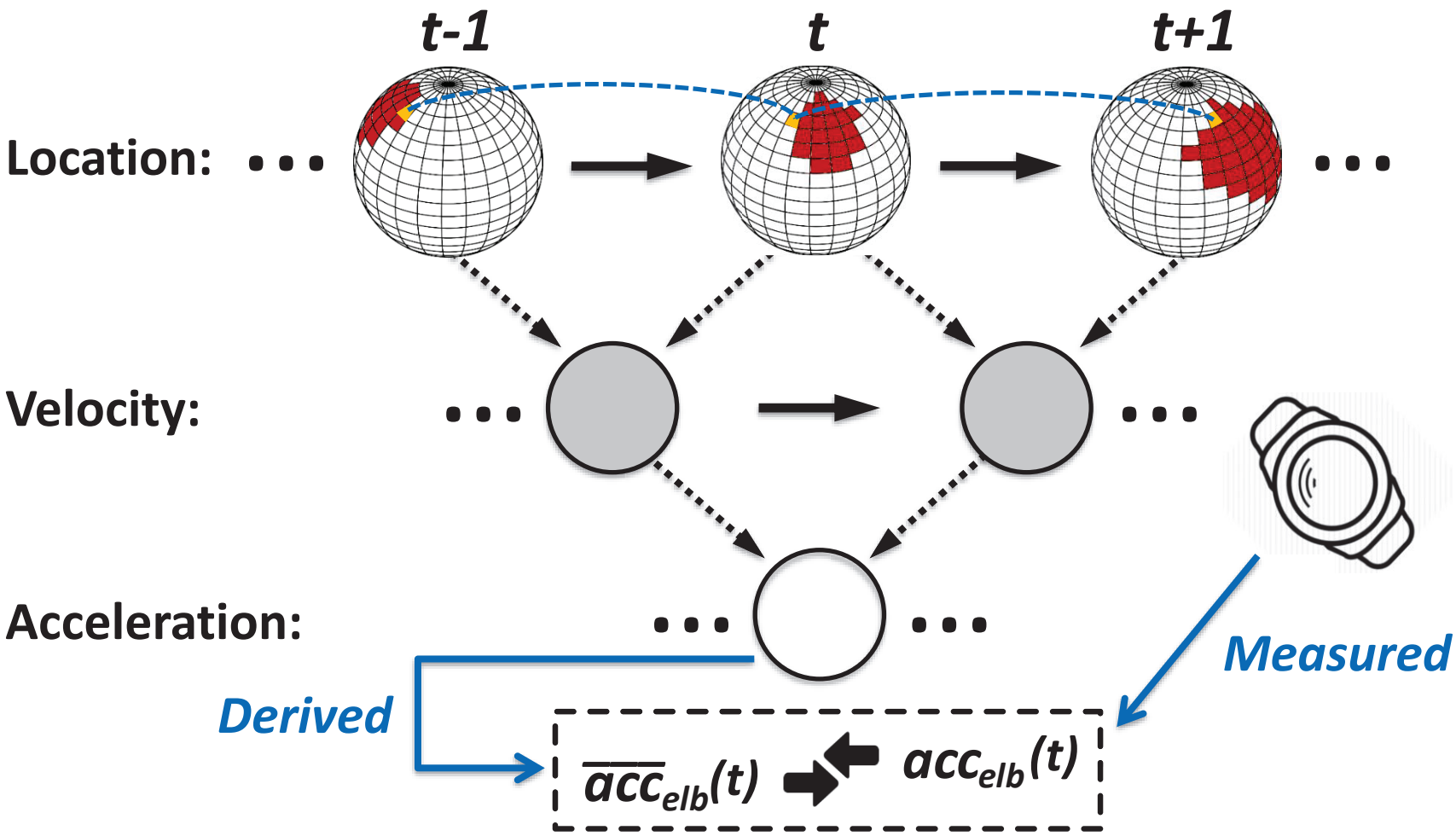
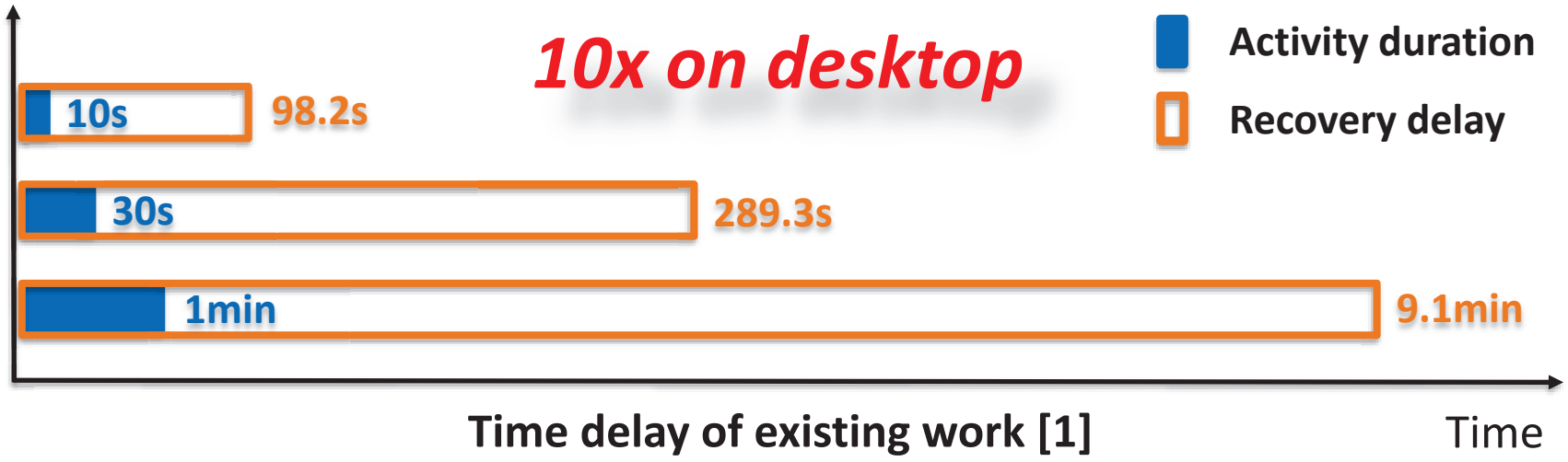
Tracking Principle



Ranges across time [1]

[1] "I am a smartwatch and I can track my user's arm", in Proc. of ACM MobiSys, 2016.

Latency



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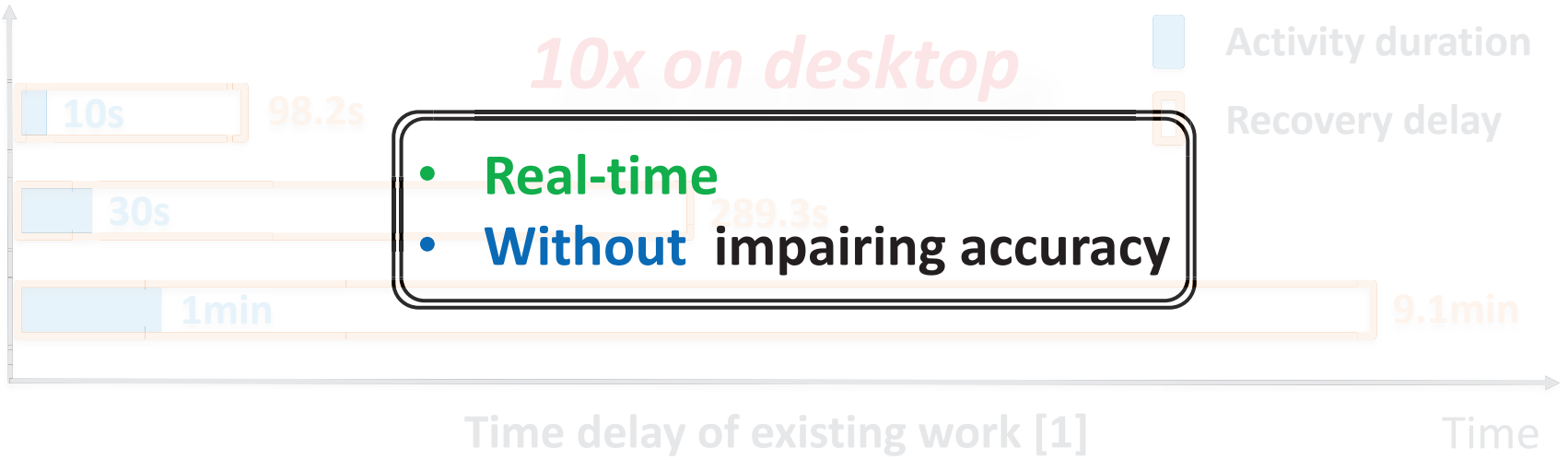
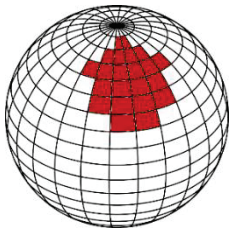
Latency



- **Real-time**
- **Without impairing accuracy**

- **Our solution [ArmTroi]:**
  - HMM state reconstruction
  - **Hierarchical search**



One search space

[1] "I am a smartwatch and I can track my user's arm", in Proc. of ACM MobiSys, 2016.

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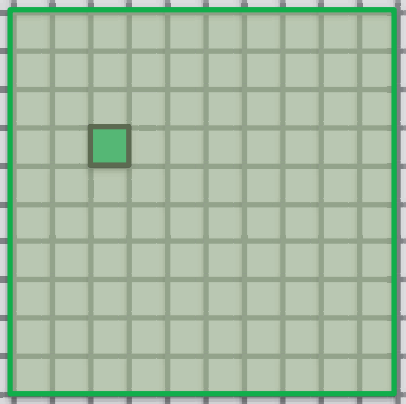
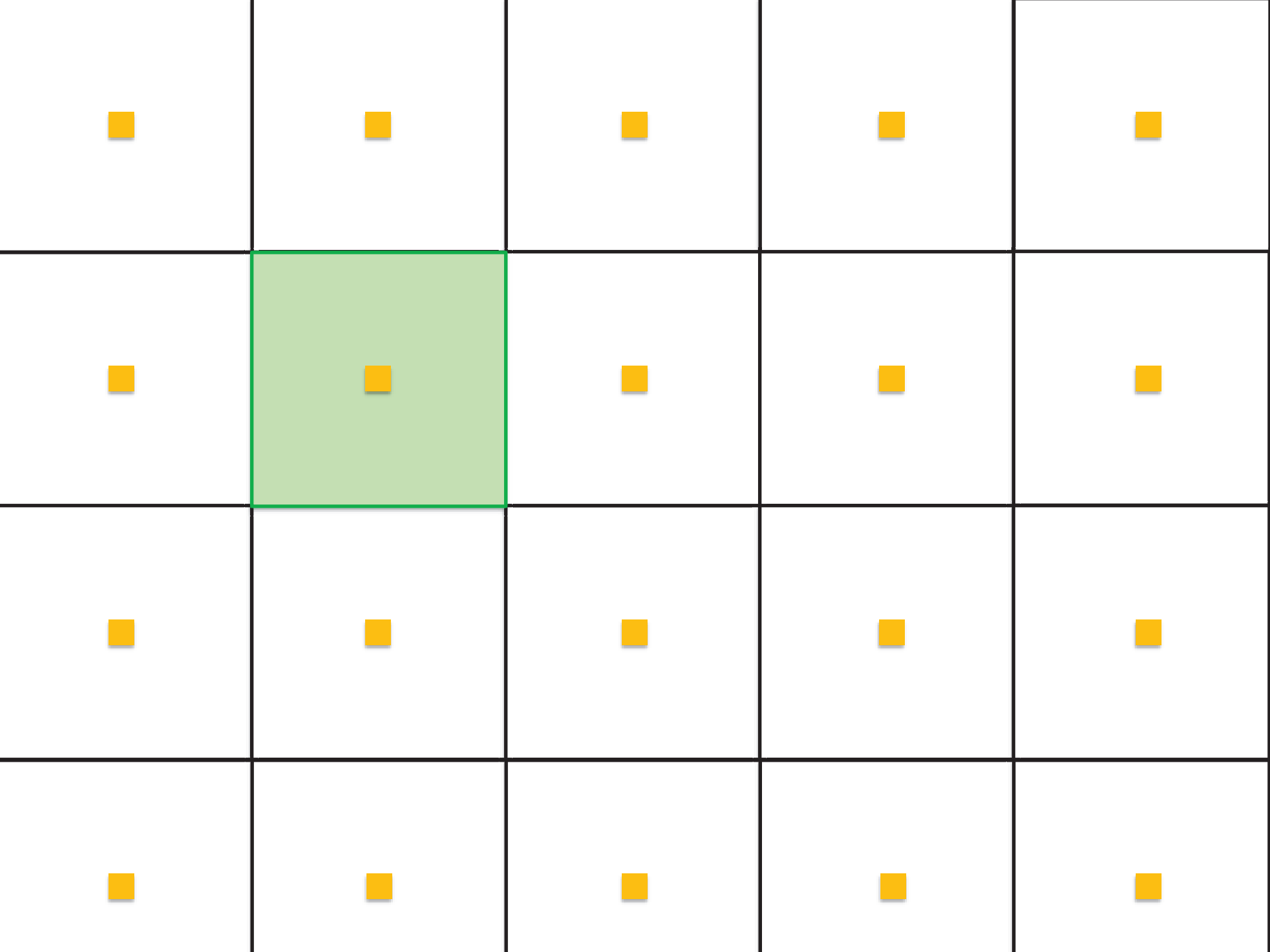
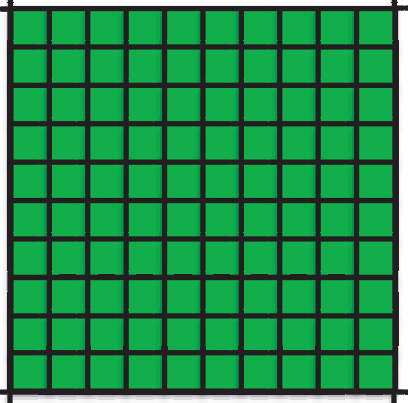


?

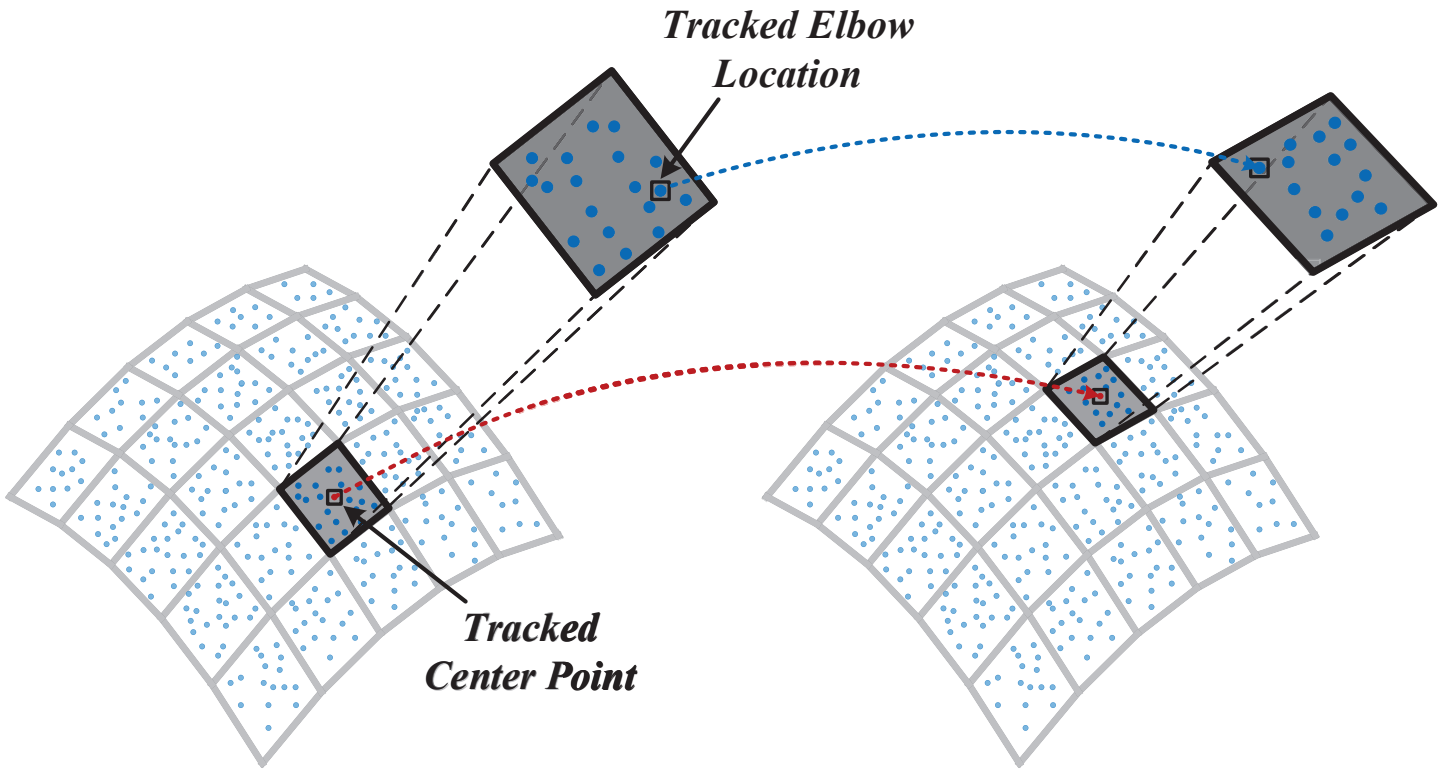
Our idea: exclude the unlikely locations using as little effort as possible





Focus on more likely candidates

Hierarchical Search



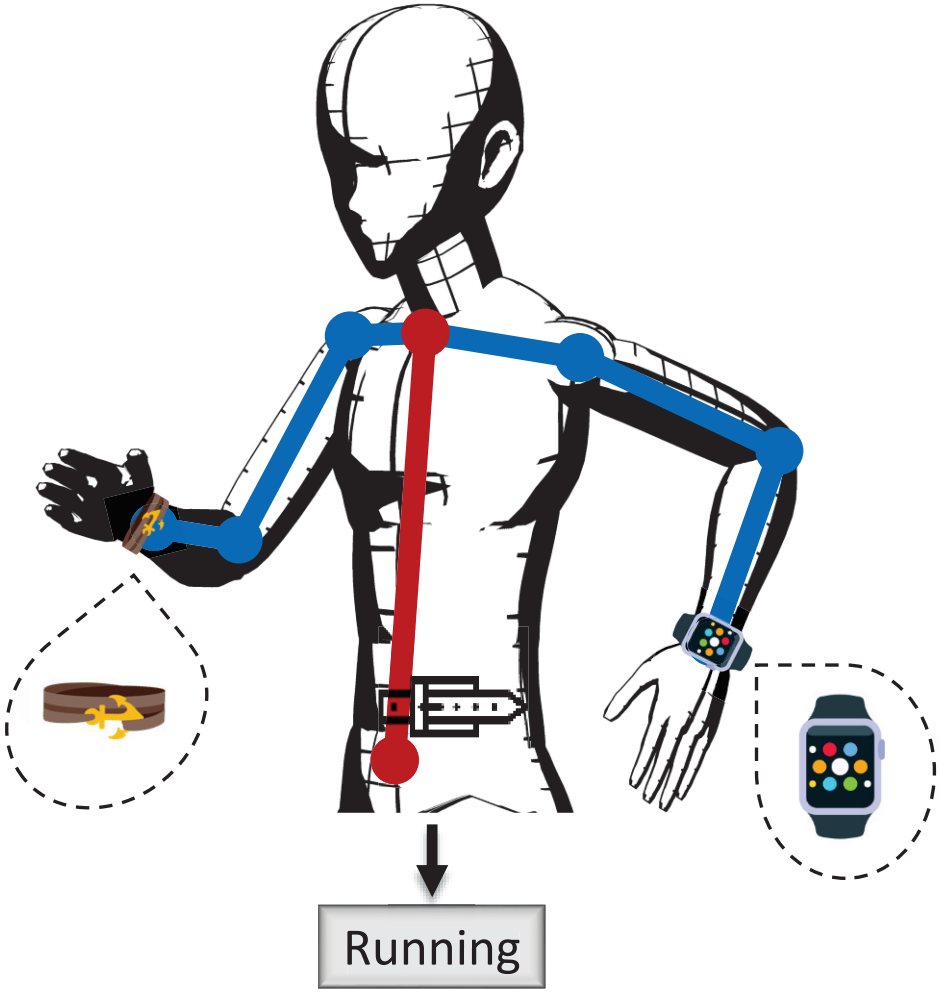
- **Time complexity:** $O\left(\left(\frac{N}{n_1}\right)^2 T + (n_1)^2 T\right)$

Original size

Size of region

- **Real-time**
- **Without impairing accuracy**

Understanding Human Arm Motions



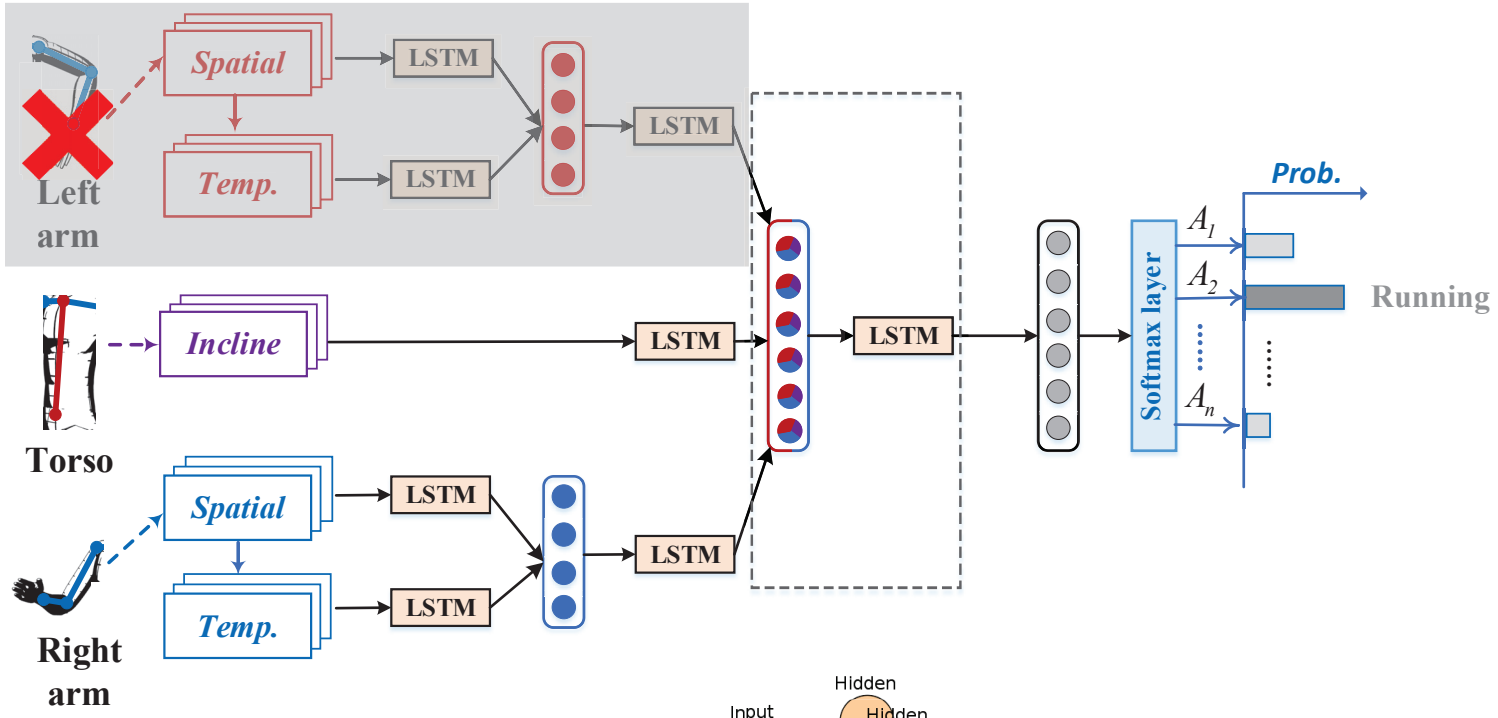
- How is the arm moving?

  Skeleton tracking

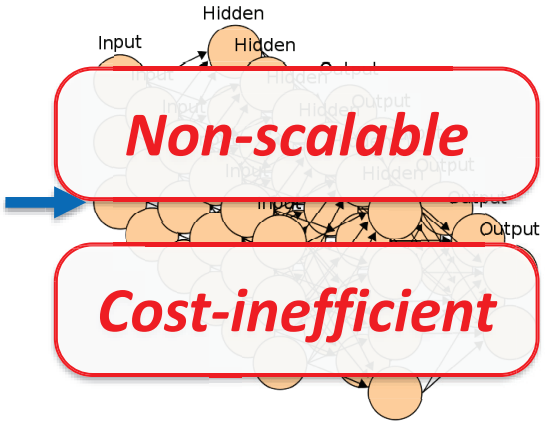
- **What is the meaning of this arm motion?**

  Motion inference

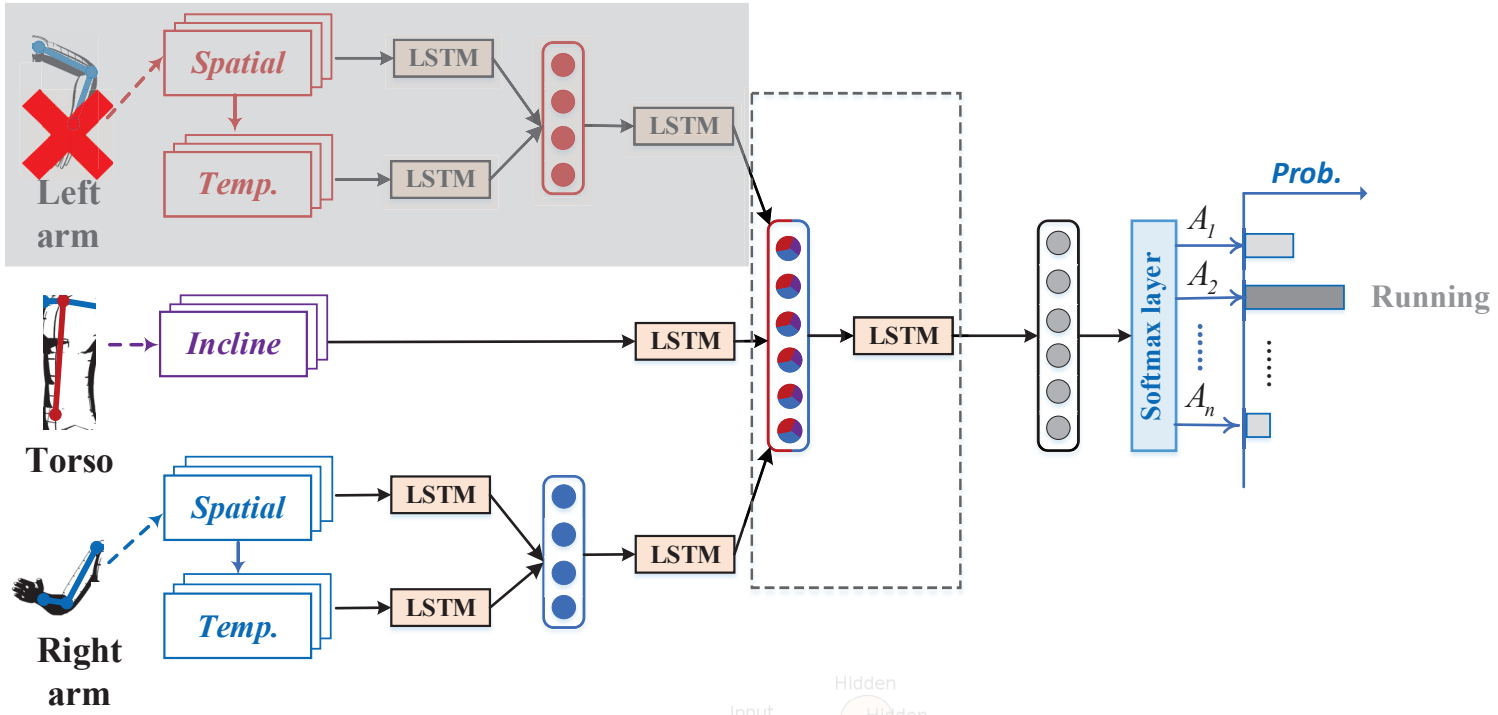
Motion Inference



6 combinations of missing inputs

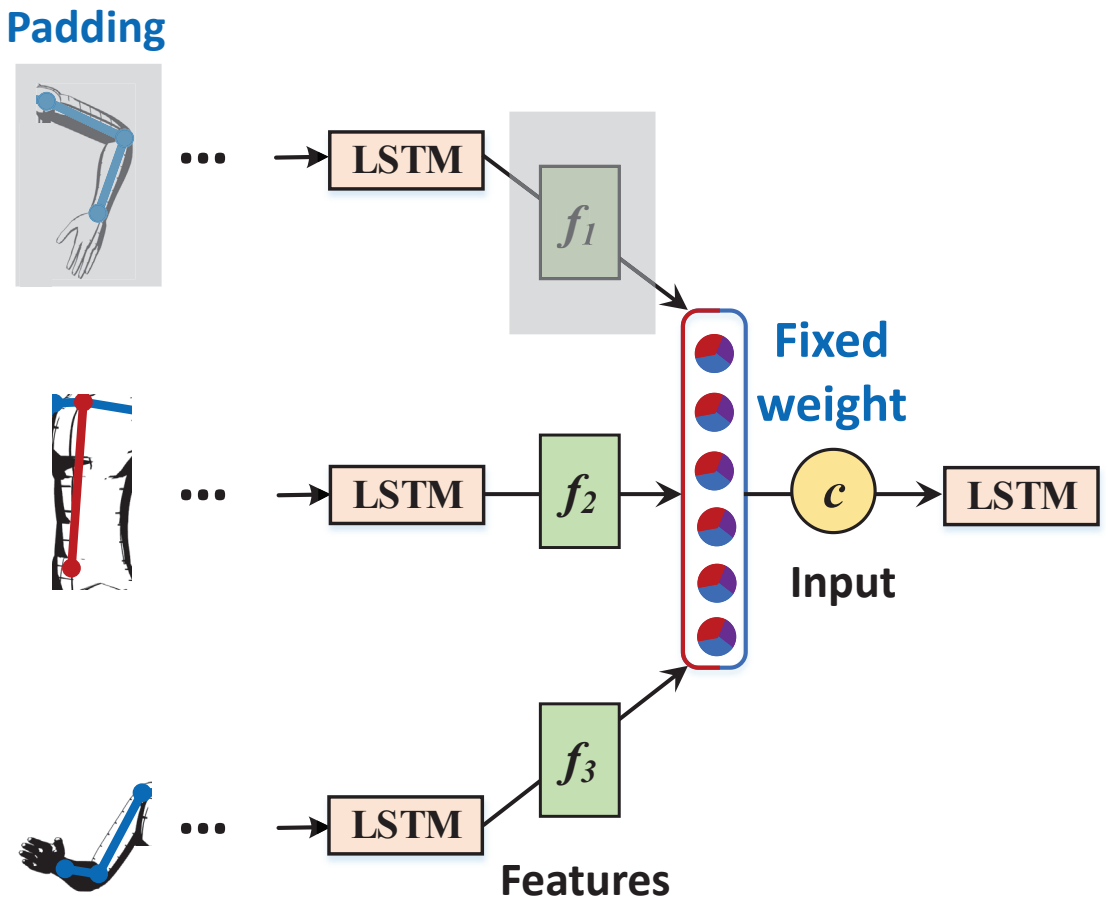


Motion Inference

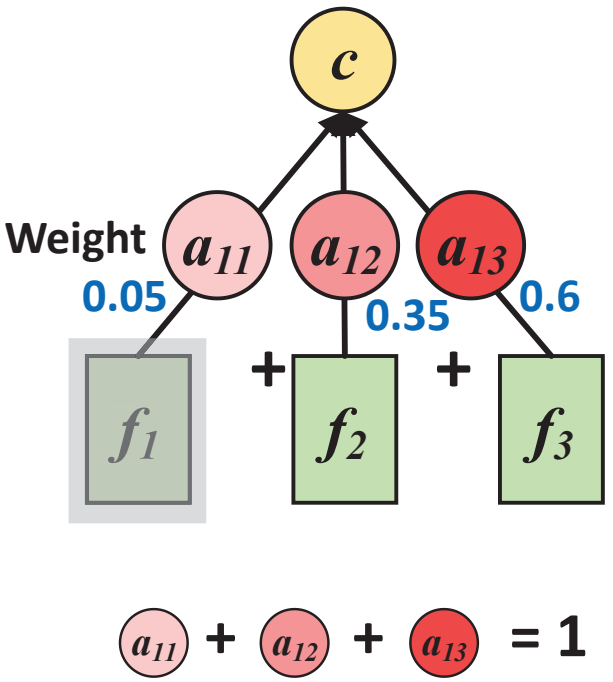


Handle **all** combinations using **one** network?

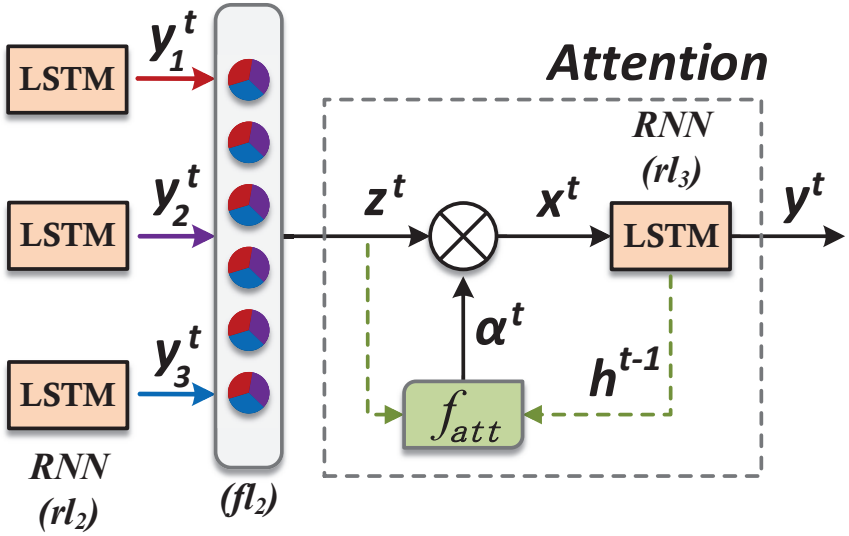
Our idea

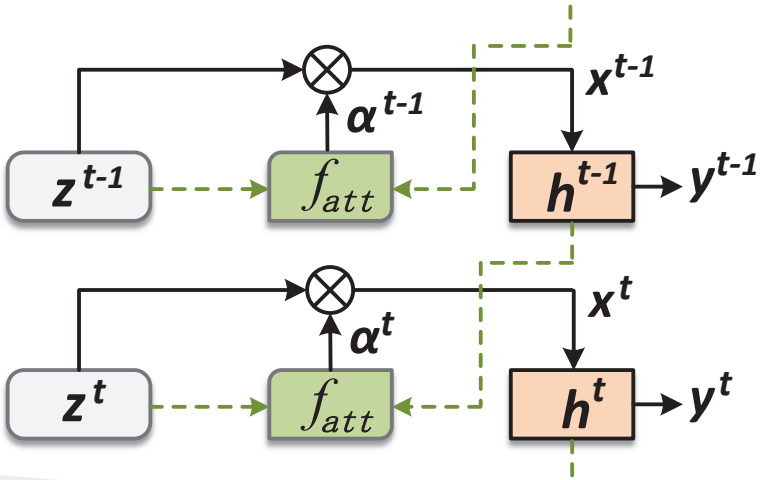


- **Adaptive design**



Attention-based network adaption





Features: $z^t = \{z_1^t, z_2^t, z_3^t\}$

Input: x^t

$z^t \rightarrow x^t$: Weighted fusion

$$x^t = \emptyset(\{z_r^t\}, \{\alpha_r^t\}), r = 1,2,3$$

Updated weights

- **Weight update**
  - aligning with the activity descriptor h_{t-1}

$$\alpha^t = f_{att}(z^t, h^{t-1})$$
$$att^t = Relu(z^t + Wh^{t-1} + b)$$
$$\alpha^t = f_{soft-max}(Uatt^t)$$

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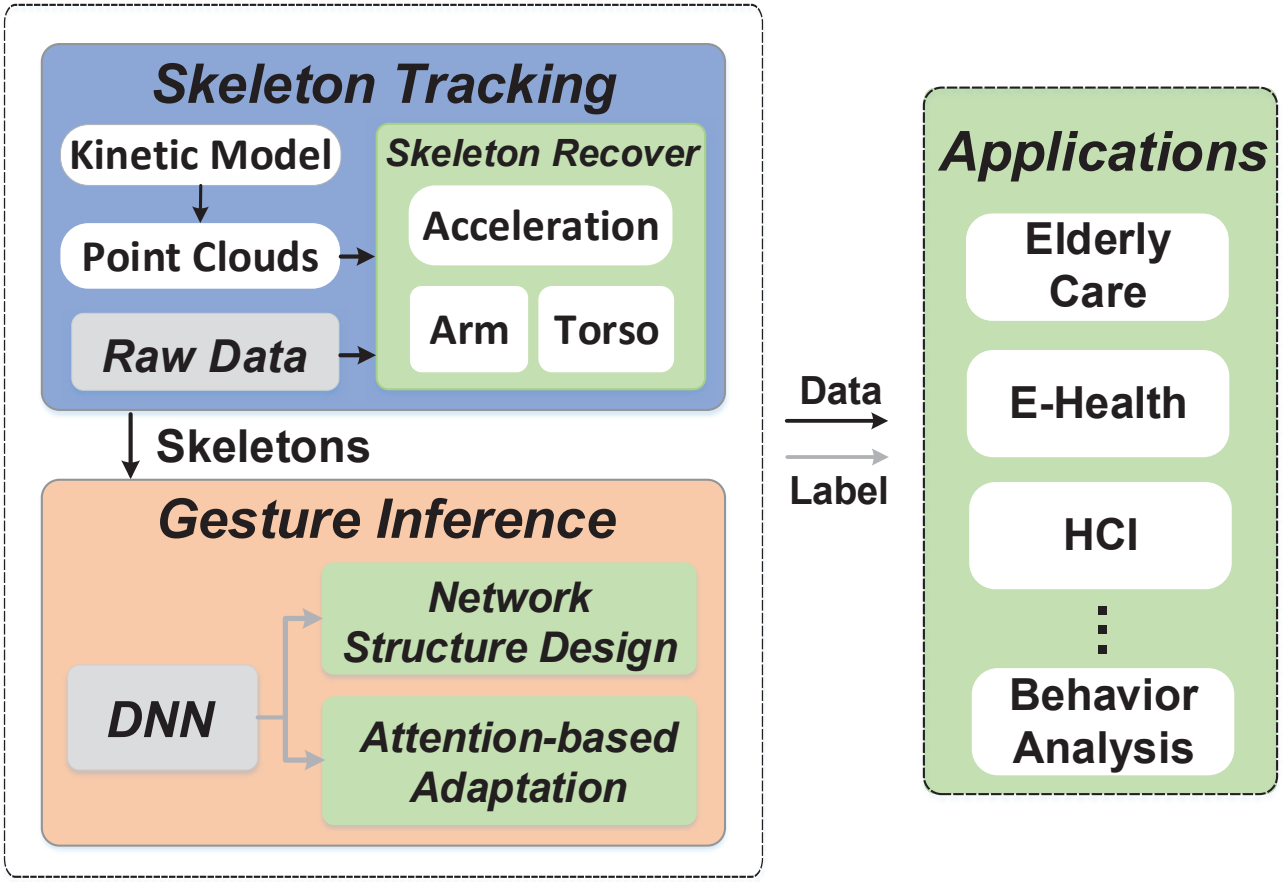
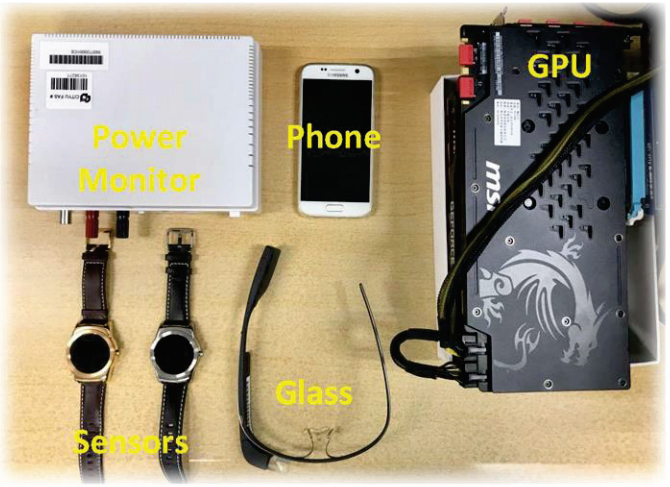
ArmTroi Implementation



Experiment setup

- Participants: 7 volunteers
- Dataset:



Daily activities

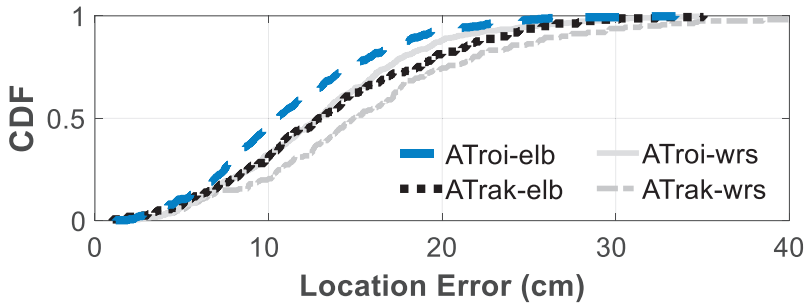
Categories	Gestures
Daily gestures (4)	*shake hands, make a call, open a door, drink water*
Free-weight (10)	*front raise (a/p), biceps curl (a/p), bent over single arm, chest fly (i/s) bench press (i/s), lateral raise*
Customized (3)	*push, pull, circle*

Table 1: Targeted gestures in ArmTroi. The *a, p, i, s* stand for alternating, in parallel, incline and sitting, respectively.

- Training: Intel i7-6700 CPU and Nvidia GTX 1080Ti GPU
- Running: SAMSUNG Galaxy S7

Evaluation

- Skeleton tracking



- **ArmTrak [1]**
  - Elbow: 12.94cm
  - Wrist: 14.91cm
- **ArmTroi**
  - Elbow: 10.53cm
  - Wrist: 12.94cm

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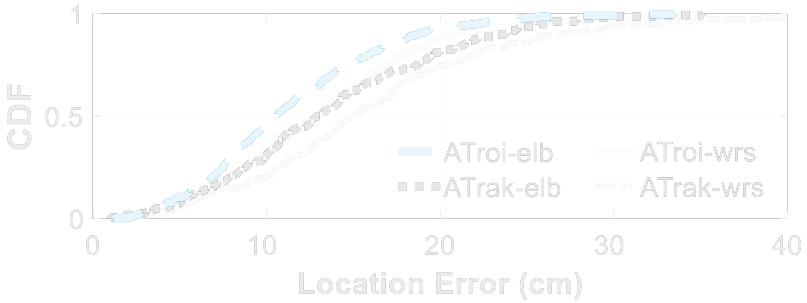
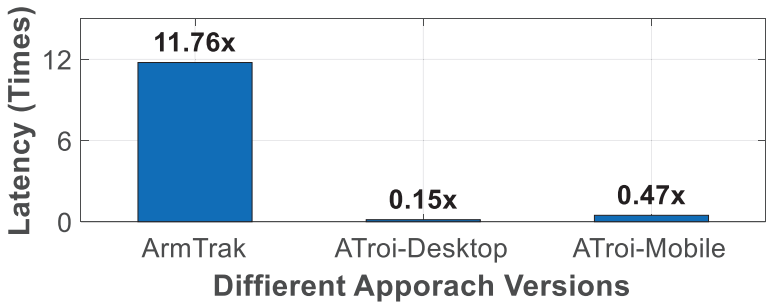
Evaluation

- Skeleton tracking





- ArmTrak [1]
  - Elbow: 12.94cm
  - Wrist: 14.91cm
- ArmTroi
  - Elbow: 10.53cm
  - Wrist: 12.94cm
- **Our latency**
  - Desktop: 0.15x
  - Phone: 0.47x

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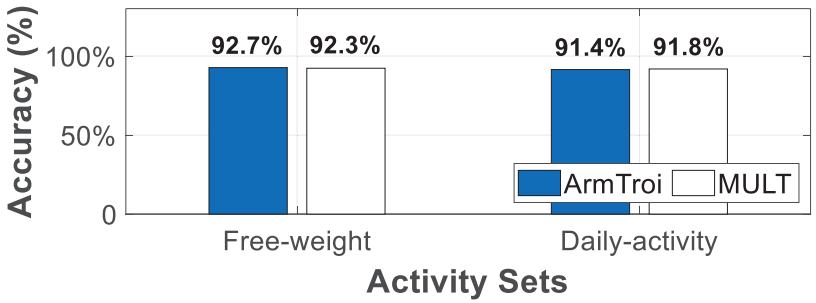
Evaluation

- Motion inference



- **Baseline: MULT**
  - Each combination of missing input
- **Accuracy with full set**
  - FW: 92.7% vs 92.3%
  - DA: 91.4% vs 91.8%

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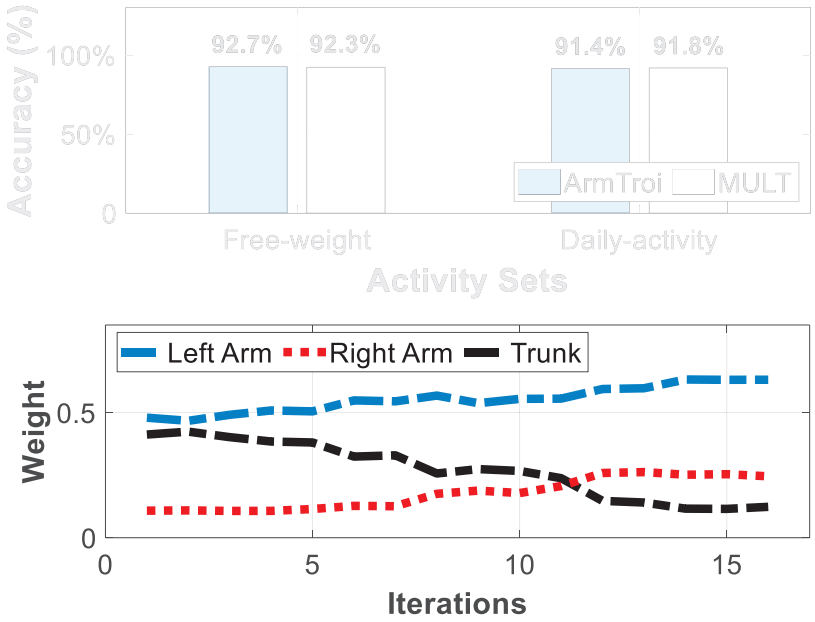
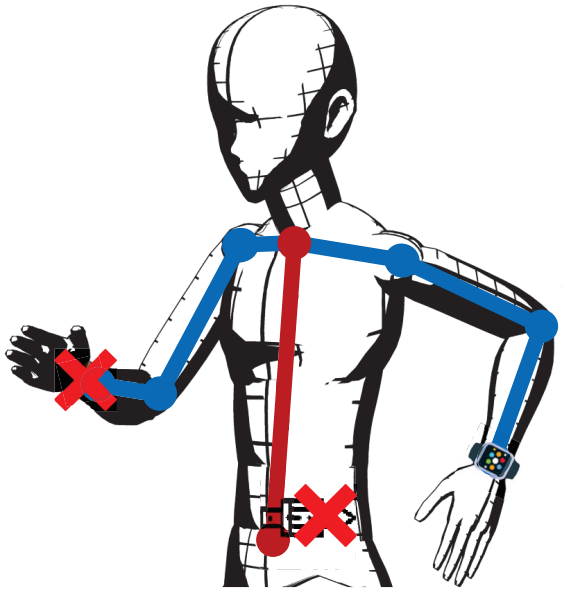
Evaluation

- Motion inference





- **Weight updating**
  - Available input: Left Arm
  - LA's weight increases

Conclusion 1, 2, 3

1. **One goal:**
   - Understanding **human arm motions**
2. **Two aspects:**
   - **Real-time** tracking
   - Motion inference **tolerant to missing inputs**
3. **Three techniques:**
   - HMM state **reorganization**
   - **Hierarchical** search
   - Attention-based network **adaption**