



# Robust Marker Trajectory Repair for MOCAP using Kinematic Reference

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# Motivation

## Getting Animation Data for Games

- Keyframing
- MOCAP
  - Markerless
  - Markers
    - Active
    - Passive

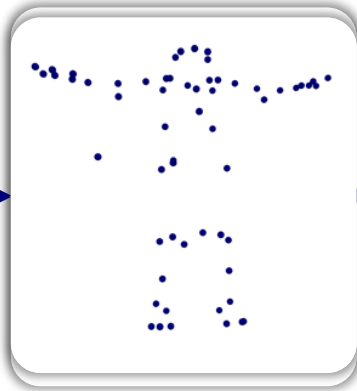


# Challenges

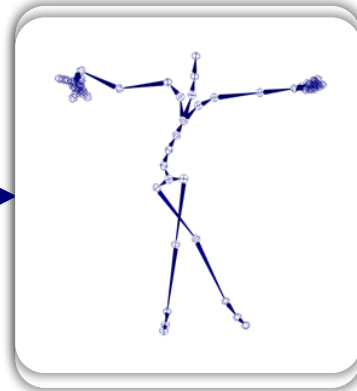
## Actual MOCAP Pipeline



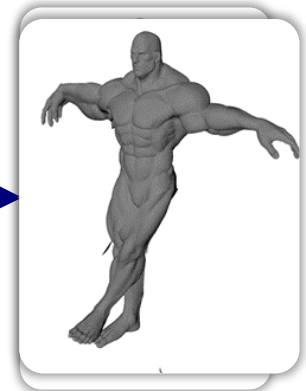
**MOCAP Studio**



**Marker Data**



**Kinematic Solution**



**Delivery**

- Markers often have erroneous gaps
  - Occlusion, Swapping, Noise, Trajectory Errors
- Erroneous gaps require MOCAP artist cleanup
  - Time and Expertise
  - Bottleneck for production

# Solutions

## Marker Level Solutions

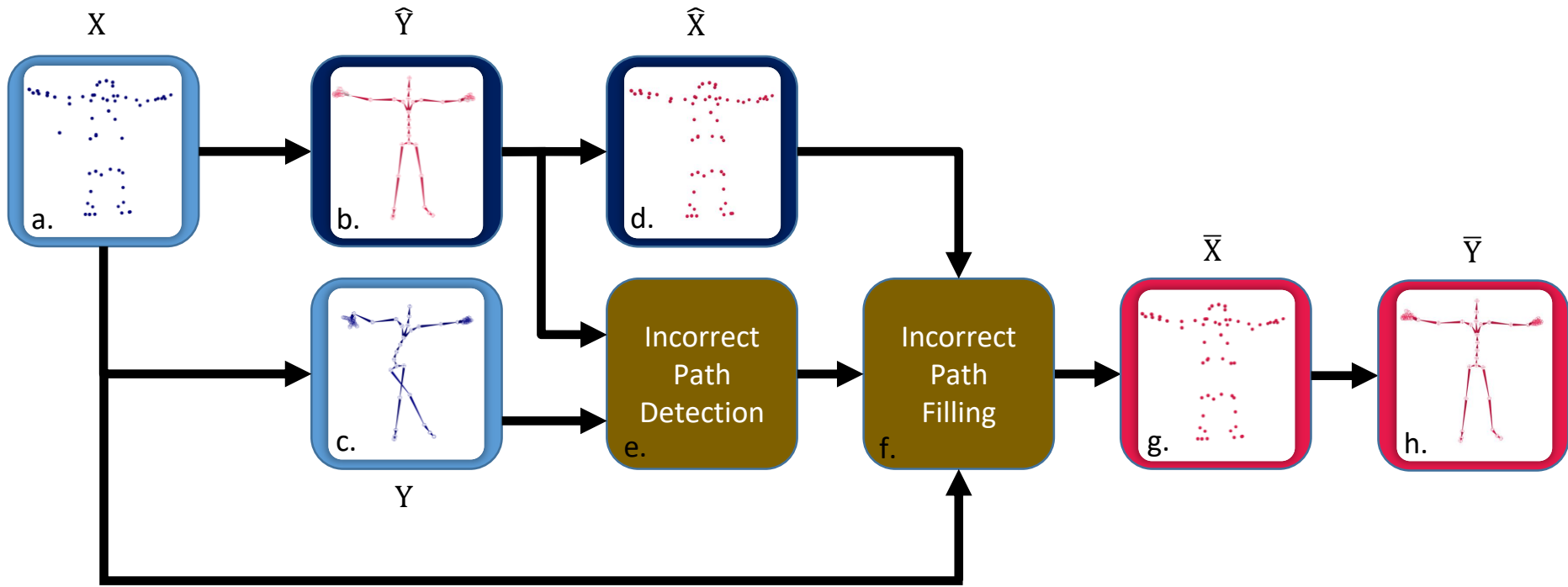
- Solve markers individually
- Marker PCA [*Liu and McMillan 2006; Federolf 2013; Gløersen and Federolf 2016*]
- Data-driven [*Baumann et al.2011; Hsu et al.2004; Wang et al.2014; Zhang and van de Panne 2018; Kucherenko et al. 2018*]
- Doesn't take into account kinematic validity

# Solutions

## Kinematic Level Solutions

- Solve directly
- Kinematically valid, robust to noise
- Commercial Solvers: *[Vicon Software]*
- Data-driven: *[Ren et al.2005; Kim and Rehg 2008; Shen et al.2012; Fragkiadaki et al.2015; Mall et al. 2017; Holden 2018]*
- Hard to integrate → Marker paths are gone

# Our Solution



a. Raw Marker Data  $X$

d. Reconstructed Markers  $\hat{X}$

g. Fixed Marker Paths  $\bar{X}$

b. [Holden 2018]\* Solver  $\hat{Y}$

e. Incorrect Path Detection

h. Final Kinematic Solution  $\bar{Y}$

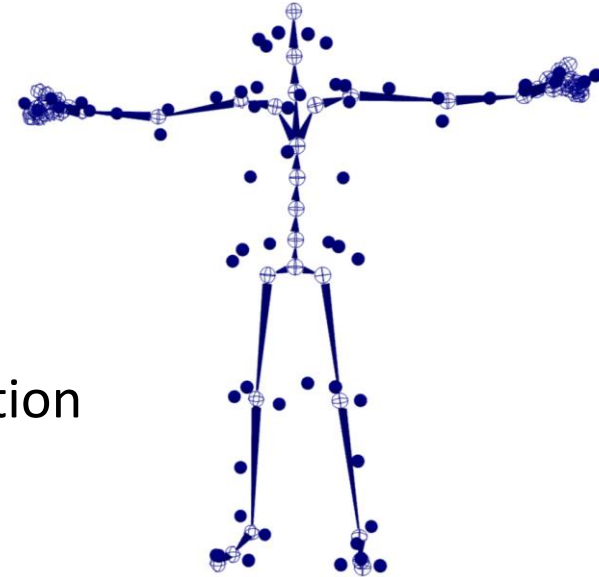
c. Commercial Kinematic Solution  $Y$

f. Incorrect Path Filling

# Solution: Marker Reconstruction

- Reconstruct markers from  $\hat{Y}$  using LBS

$$\text{LBS}(\hat{Y}, Z) = \sum_{i=0}^j w_i \odot (\hat{Y}_i \otimes Z_i)$$



- New markers coined  $\hat{X}$

- Marker representation of  $\hat{Y}$  kinematic solution
- “Clean” set of markers

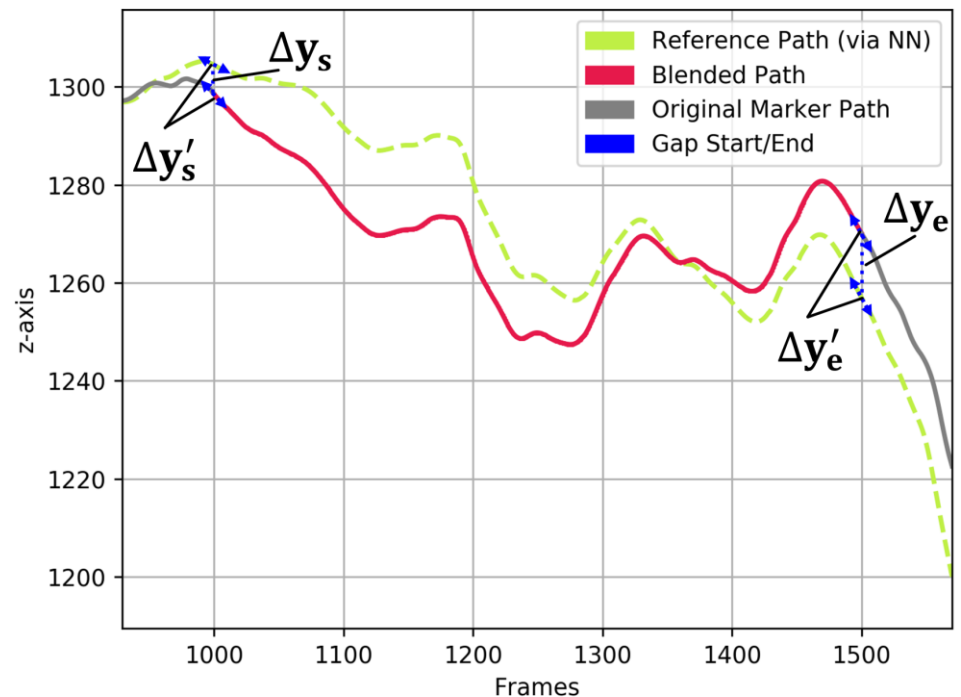
- Issues:

- Missing small detail
- Offset from original paths  $X$

# Solution: Gap Filling

Fix Erroneous Frames in  $\mathbf{X}$  using  $\hat{\mathbf{X}}$

- Get  $\Delta y_s, \Delta y'_s$  and  $\Delta y_e, \Delta y'_e$
- Fit degree 5 polynomial
- Subtract polynomial from original path

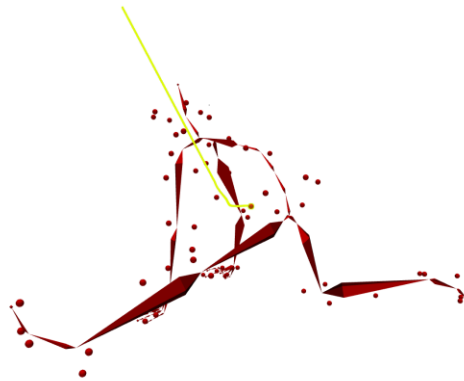
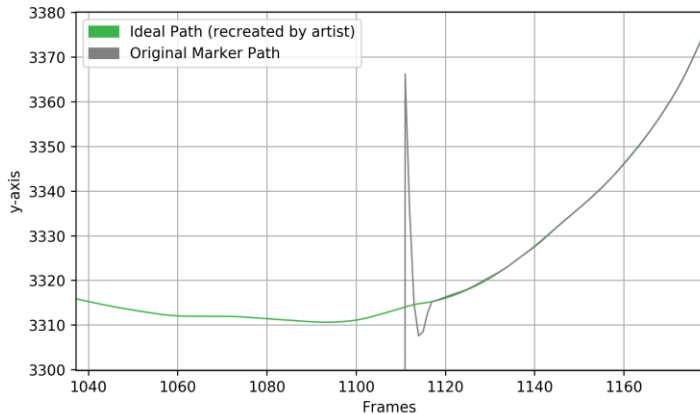




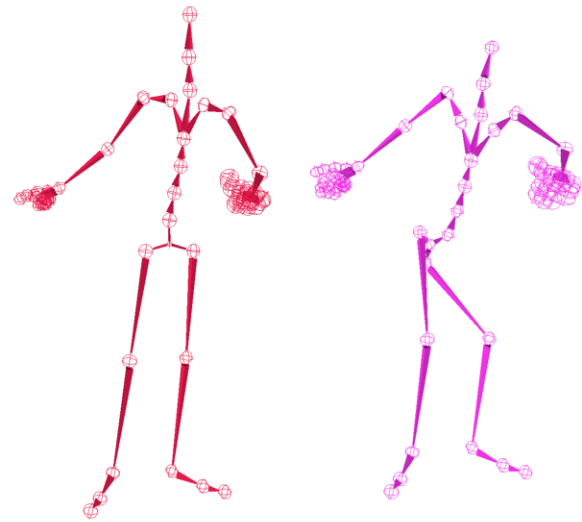
# Solution: Gap Filling

## Handle Degenerate Cases

- Pad start/end frames
- Search for best start/end frames for gaps
- Clamp velocity differences



# Results



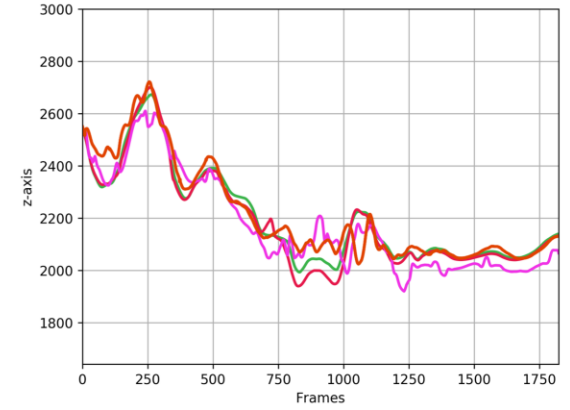
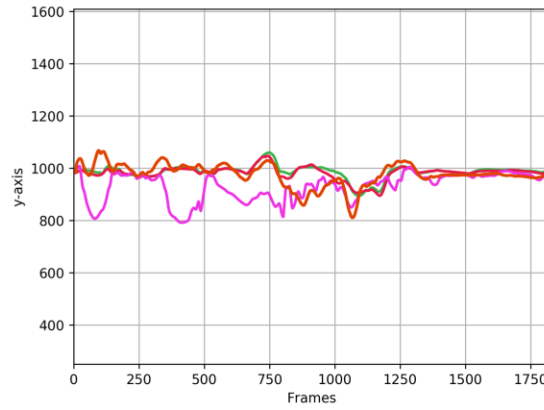
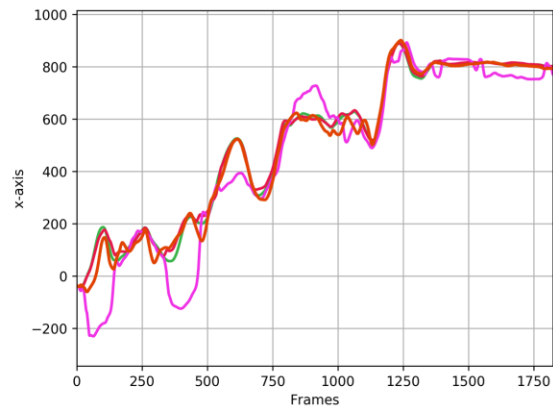
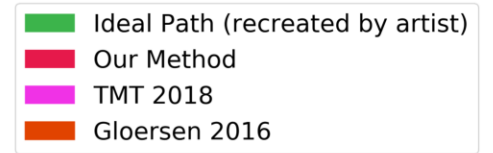
a)



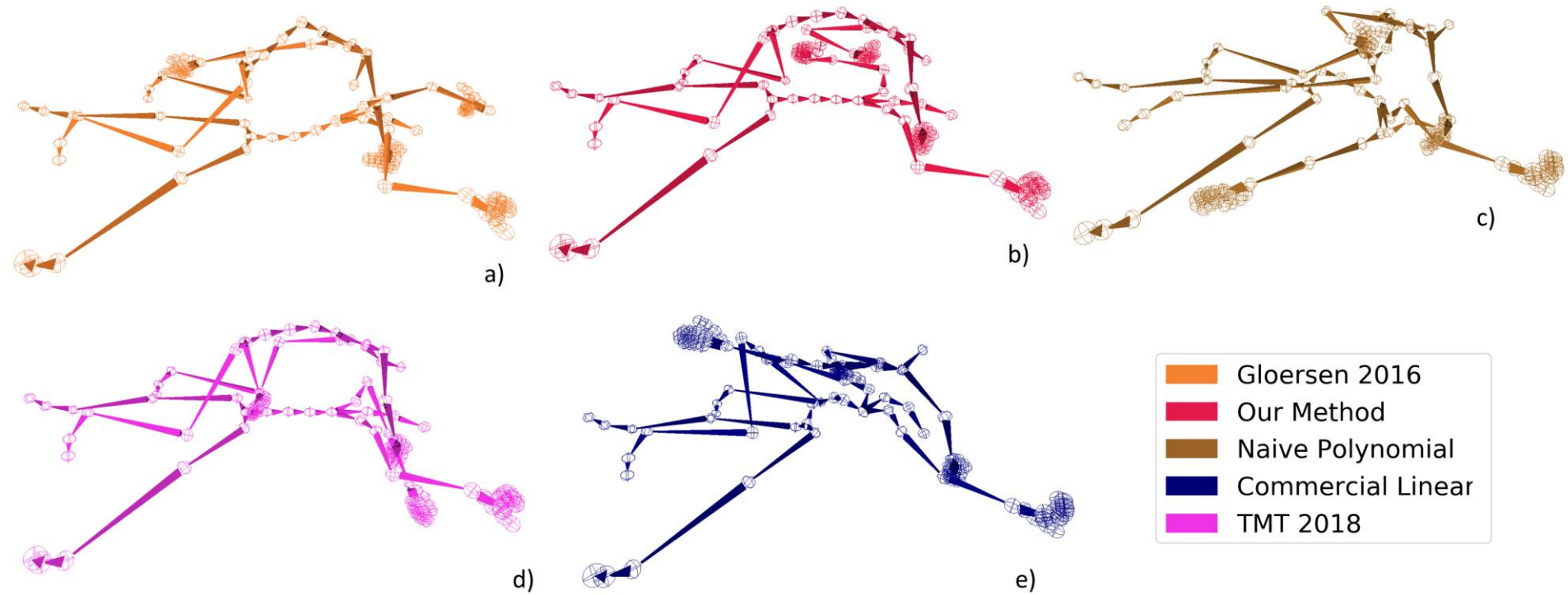
b)

c)

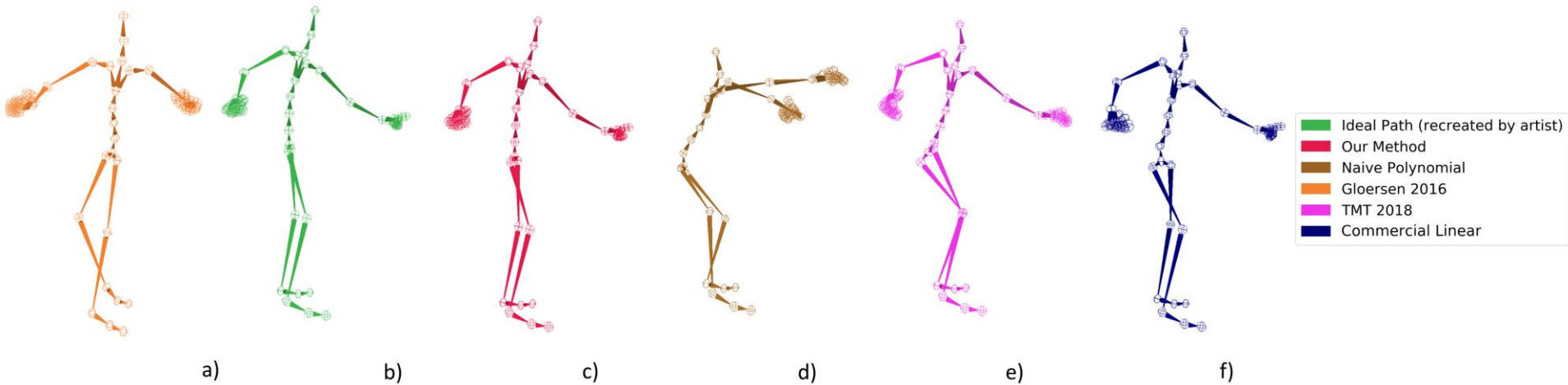
d)



# Results



# Results



Method	$\Delta_{\text{pos}}$	$\Delta_{\text{rot}}$
Original	1.841	35.220
Naive Polynomial	10.678	42.93
Commerical	1.053	22.101
Gloersen 2016	0.716	23.827
TMT 2018	0.603	8.038
Holden 2018	1.337	19.604
Ours	<b>0.288</b>	<b>4.727</b>

# Results

- TMT 2018
- Our Method
- Naive Polynomial
- Commercial Linear
- Gloersen 2016
- Ideal Path (recreated by artist)



# Results

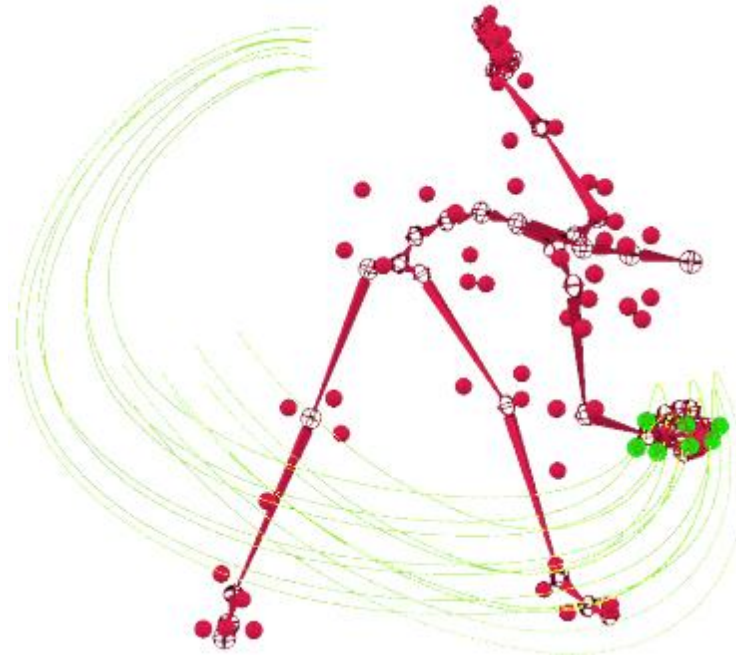
- Performance metrics on 10,920 frames of animation at 120FPS

<b>Method</b>	<b>FPS</b>	<b>Time (%)</b>
NN Preprocessing	4029	2
NN Evaluation	1339	6
Filtering	5527	1
IK Retargeting	108	74
Marker Reconstruction	2278	4
Gap Detection	10749	1
Gap Preprocessing	904	9
Marker Filling	2620	3
<b>Total</b>	<b>81</b>	<b>100</b>

# Future

- Limitations:
  - Depends on performance of robust kinematic solver
  - Degree 5 polynomial
  - Ambiguous results when large number of markers missing
- Future work:
  - Use different kinematic solver
  - Use different IK process
  - Improve on blending → interialization?
  - Integrate with markerless system

# Q&A



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# Appendix: Improved Kinematic Solver

- Based on *[Holden 2018]*
- Improve kinematic solver to remove issues such as foot sliding
- Augment dataset using local perturbations
  - Sampled on entire dataset of training data
- Increase correlations between missing markers

# Appendix: Erroneous Interval Detection

- Compare kinematic solutions: Positions and Rotation
- Check against threshold
  - Positions → 10cm max
  - Rotation → 30 degrees max
- Determine erroneous markers

# Appendix: Erroneous Interval Detection

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**Algorithm 1** Given two kinematic solutions, an erroneous one and a clean one, determine the set of missing markers paths

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**Function** *DetectBad* ( $Y \in \mathbb{R}^{n \times j \times 3 \times 4}$ ,  $\hat{Y} \in \mathbb{R}^{n \times j \times 3 \times 4}$ )

// Get differences between two kinematic solutions

$$\Delta P \in \mathbb{R}^{n \times j} \leftarrow \|\hat{Y} - Y\|$$

$$\Delta R \in \mathbb{R}^{n \times j} \leftarrow \text{Angle}(\hat{Y}Y^T)$$

// Get joints surpassing allowed threshold

$$J \in (0, 1)^{n \times j} \leftarrow \begin{cases} 1, & \text{if } \Delta P - \Delta P_{\max} \geq 0 \\ & \text{or } \Delta R - \Delta R_{\max} \geq 0 \\ 0, & \text{otherwise} \end{cases}$$

// Get markers associated with joints surpassing threshold

$$\hat{w} \in (0, 1)^{m \times j} \leftarrow \begin{cases} 1, & \text{if } w > 0 \\ 0, & \text{otherwise} \end{cases}$$

$$M \in (0, 1)^{n \times m} \leftarrow J \hat{w}$$

**return** M

**End**

# Appendix: Gap Filling

**Algorithm 2** Given a gap in marker data, fill the gap by referring to a reference marker path

**Function** *GapFill* ( $X \in \mathbb{R}^{n \times m \times 3}$ ,  $\hat{X} \in \mathbb{R}^{n \times m \times 3}$ ,  $M \in \mathbb{N}^{g \times 3}$ )

// Copy over marker paths

$\bar{X} \in \mathbb{R}^{n \times m \times 3} \leftarrow X$

// Loop over all gaps in a clip

**for**  $i \dots g$  **do**

// Get start, end and marker indices

$s, e, m \leftarrow M_{i,1}, M_{i,2}, M_{i,3}$

// Get differences in positions

$$\Delta y_s \in \mathbb{R}^3 \leftarrow \begin{cases} \hat{X}_{s-1}^m - X_{s-1}^m, & \text{if } s > 1 \\ \hat{X}_{e+1}^m - X_{e+1}^m, & \text{if } e < n \\ 0 & \text{otherwise} \end{cases}$$

$$\Delta y_e \in \mathbb{R}^3 \leftarrow \begin{cases} \hat{X}_{e+1}^m - X_{e+1}^m, & \text{if } e < n \\ \hat{X}_{s-1}^m - X_{s-1}^m, & \text{if } s > 1 \\ 0 & \text{otherwise} \end{cases}$$

// Get differences in velocities

$$\Delta y'_s \in \mathbb{R}^3 \leftarrow \begin{cases} \frac{(\hat{X}_{s-1} - \hat{X}_{s-3})}{2} - \frac{(X_{s-1} - X_{s-3})}{2}, & \text{if } s > 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\Delta y'_e \in \mathbb{R}^3 \leftarrow \begin{cases} \frac{(\hat{X}_{e+3} - \hat{X}_{e+1})}{2} - \frac{(X_{e+3} - X_{e+1})}{2}, & \text{if } e < n \\ 0 & \text{otherwise} \end{cases}$$

// Clamp velocity differences

$\Delta y'_s \leftarrow \text{AbsClamp}(\Delta y'_s, \Delta_{\max})$

$\Delta y'_e \leftarrow \text{AbsClamp}(\Delta y'_e, \Delta_{\max})$

// Fit cubic polynomial using constraints

$\mathcal{P} \leftarrow \text{HermiteSpline}(\Delta y_s, \Delta y_e, \Delta y'_s, \Delta y'_e)$

// Subtract polynomial from reference

$\bar{X}_{s\dots e}^m \leftarrow \hat{X}_{s\dots e}^m - \mathcal{P}_{s\dots e}$

**end for**

**return**  $\bar{X}$

**End**