

# PANDORA: Diffusion Policy Learning for Dexterous Robotic Piano Playing

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## 1. Motivation & Challenge

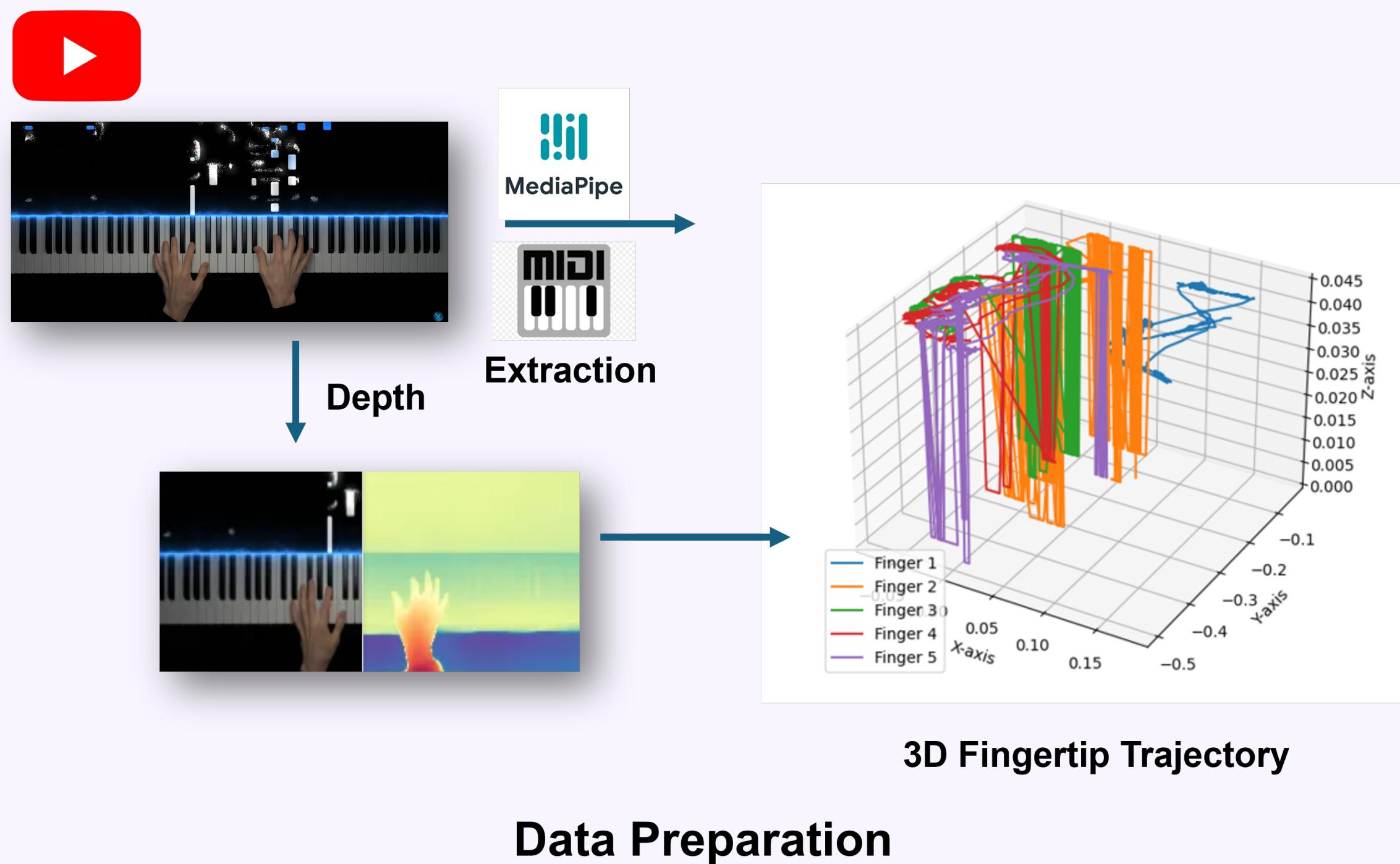
Dexterous robotic piano playing demands both **high precision** and **expressive artistry**.

### Key Challenges:

- ▶ **High-dimensional Control:** 20+ DoF per hand
- ▶ **Precision vs. Expressiveness:** Balancing accuracy with musical nuances
- ▶ **Hand Coordination:** Left (rhythm) vs. Right (melody)
- ▶ **Reward Design:** Traditional metrics miss artistic qualities

**Solution:** Diffusion-based policy with LLM semantic feedback for musical artistry.

## 2. Data Preparation Pipeline



### Enhanced 3D Trajectory Extraction:

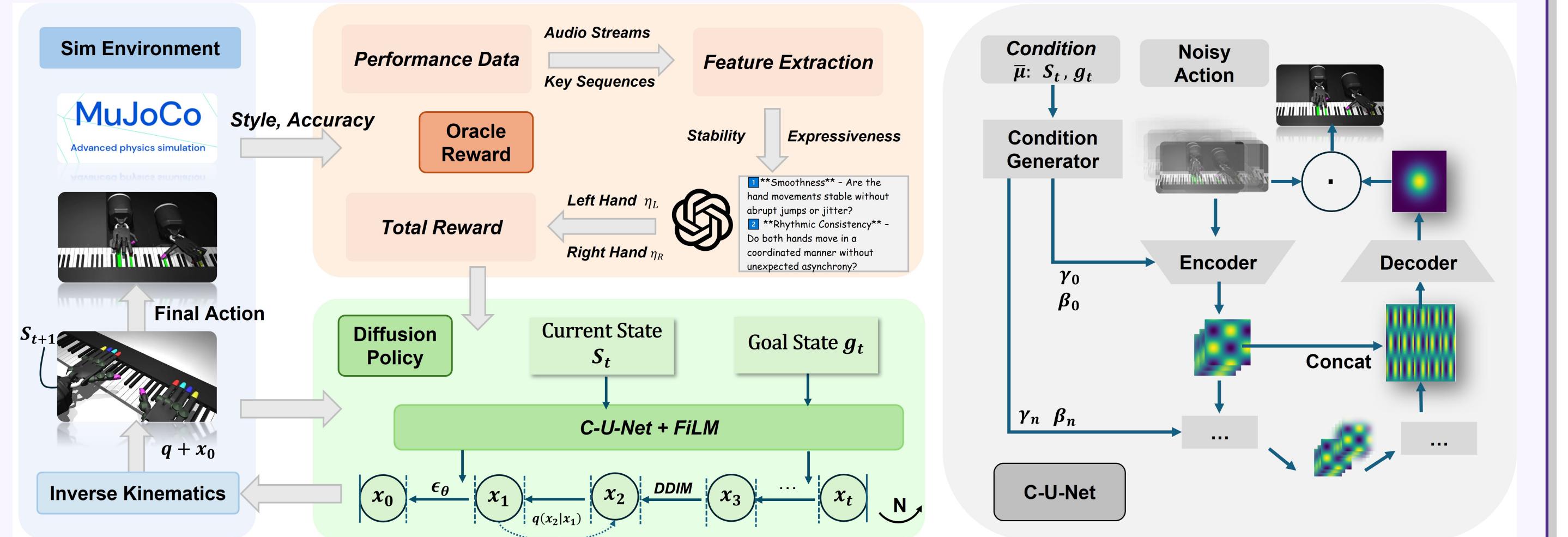
- ▶ Scrape YouTube piano performances with paired MIDI files
- ▶ Extract 2D fingertips using MediaPipe
- ▶ Augment with **DepthAnything** for depth cues
- ▶ Fuse depth + 2D to obtain robust 3D fingertip trajectories

## 3. Key Contributions

### Novel Technical Innovations:

- ▶ **Diffusion-Based Policy:** Conditional U-Net with FiLM for smooth action generation via DDIM
- ▶ **Composite Reward:** Task accuracy + audio fidelity + style mimicry + LLM semantic evaluation
- ▶ **Hand-Specific Modulation:** Left (stability) vs. Right (expressiveness) dynamic rewards
- ▶ **Residual IK:** Combines IK solver with learned residuals for precise control
- ▶ **Enhanced Data:** DepthAnything for robust 3D fingertip trajectory extraction

## 4. Framework Overview



PANDORA uses a FiLM-conditioned U-Net to iteratively denoise noisy action sequences into smooth trajectories, combines them with a residual IK head, and optimizes a composite reward that includes task accuracy, audio fidelity, style mimicry, and LLM-based semantic feedback.

## 5. Diffusion-Based Policy Learning

**DDIM Denoising:** Starting from  $x_T \sim \mathcal{N}(0, I)$ , iteratively refine:  

$$x_{t-1} = \sqrt{\bar{\alpha}_{t-1}} \left( \frac{x_t - \sqrt{1 - \bar{\alpha}_t} \epsilon_\theta(x_t, t)}{\sqrt{\bar{\alpha}_t}} \right) + \sqrt{1 - \bar{\alpha}_{t-1}} \epsilon_\theta(x_t, t)$$

### Architecture:

- ▶ U-Net with 4 blocks (64, 128, 256, 512 channels)
- ▶ FiLM layers inject state  $s$  and goal  $g_t$
- ▶  $T = 100$  steps with cosine schedule

**Training:** Cosine LR (1e-4 to 1e-6), EMA (0.9999),  $\approx 1$  hr on RTX 4090

## 6. Composite Reward Function

### Multi-Component Design:

$$(1) \text{ Task } R_{\text{task}} = w_{\text{press}}(1 - \text{error}) - w_{\text{fp}} \cdot \text{FP}$$

$$(2) \text{ Audio } R_{\text{audio}} = \frac{X_{\text{target}} \cdot X_{\text{robot}}}{\|X_{\text{target}}\| \|X_{\text{robot}}\|}$$

$$(3) \text{ Style } R_{\text{style}} = -\|\tau_{\text{robot}} - \tau_{\text{human}}\|_2^2$$

### (4) LLM Oracle with Hand-Specific Modulation:

$$R_{\text{LLM}}^L = S_{\text{LLM}} \times \eta_L, \quad R_{\text{LLM}}^R = S_{\text{LLM}} \times \eta_R$$

where  $\eta_L$  emphasizes stability,  $\eta_R$  emphasizes expressiveness.

### Final Reward:

$$R = \alpha R_{\text{task}} + \beta R_{\text{audio}} + \gamma R_{\text{style}} + \delta (R_{\text{LLM}}^L + R_{\text{LLM}}^R)$$

## 7. Quantitative Results

13-song evaluation using Precision, Recall, and F1 metrics.

Method	Precision	Recall	F1
PianoMime (Two-stage)	0.68	0.54	0.57
PianoMime (Residual)	0.70	0.56	0.58
<b>PANDORA</b>	<b>0.78</b>	<b>0.60</b>	<b>0.68</b>

+10% F1 improvement with  $\sim 3\times$  faster training vs. baselines.

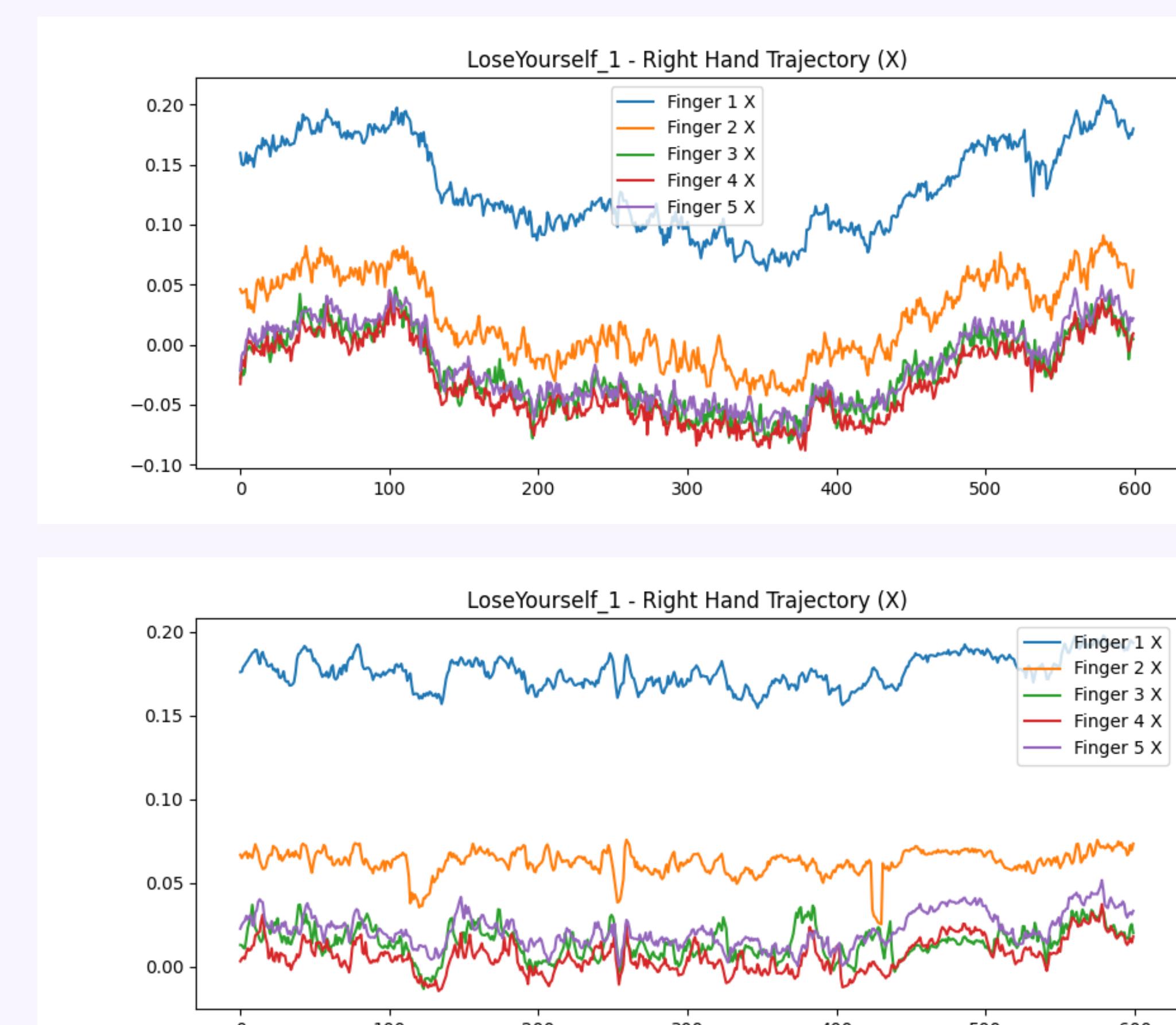
## 8. Ablation Study

Configuration	Mean F1 Score
LLM + Residual (Full)	<b>0.90</b>
LLM - no Residual	0.73
no LLM + Residual	0.68
no LLM - no Residual	0.62

### Insights:

- ▶ **Residual policy** is essential for precise, stable key strikes.
- ▶ **LLM feedback** is critical for expressive phrasing and dynamics.
- ▶ Their **synergy** yields the best accuracy and artistry.

## 9. Trajectory Analysis



## Conclusion & Resources

**Summary:** PANDORA bridges precision and artistry via diffusion, residual IK, and LLM-guided rewards.

**Future:** Faster sampling, multi-instrument extension, real-world deployment.

**Project:** <https://taco-group.github.io/PANDORA>

**Paper:** <https://arxiv.org/abs/2503.14545>

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