

# Small Project F12

### Visual Reconstruction of Played Music

Computer vision solution to transcribe piano performances from silent video

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### Project Overview

### Challenge

Detect which piano keys are pressed from silent video

### Approach

Classical CV pipeline using keyboard rectification and fingertip tracking

#### Output

Transcribed key presses converted to MIDI files

# **Keyboard Calibration Process**



#### Reference Frame

Capture still image without hands



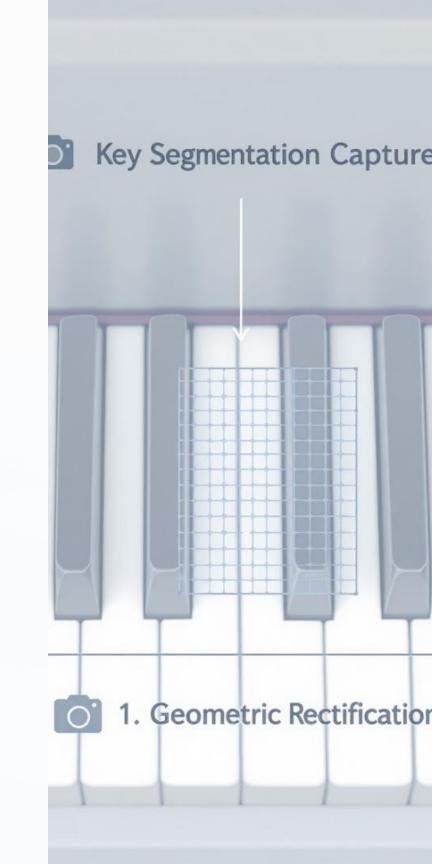
#### Geometric Rectification

Extract boundaries using Canny and Hough transforms

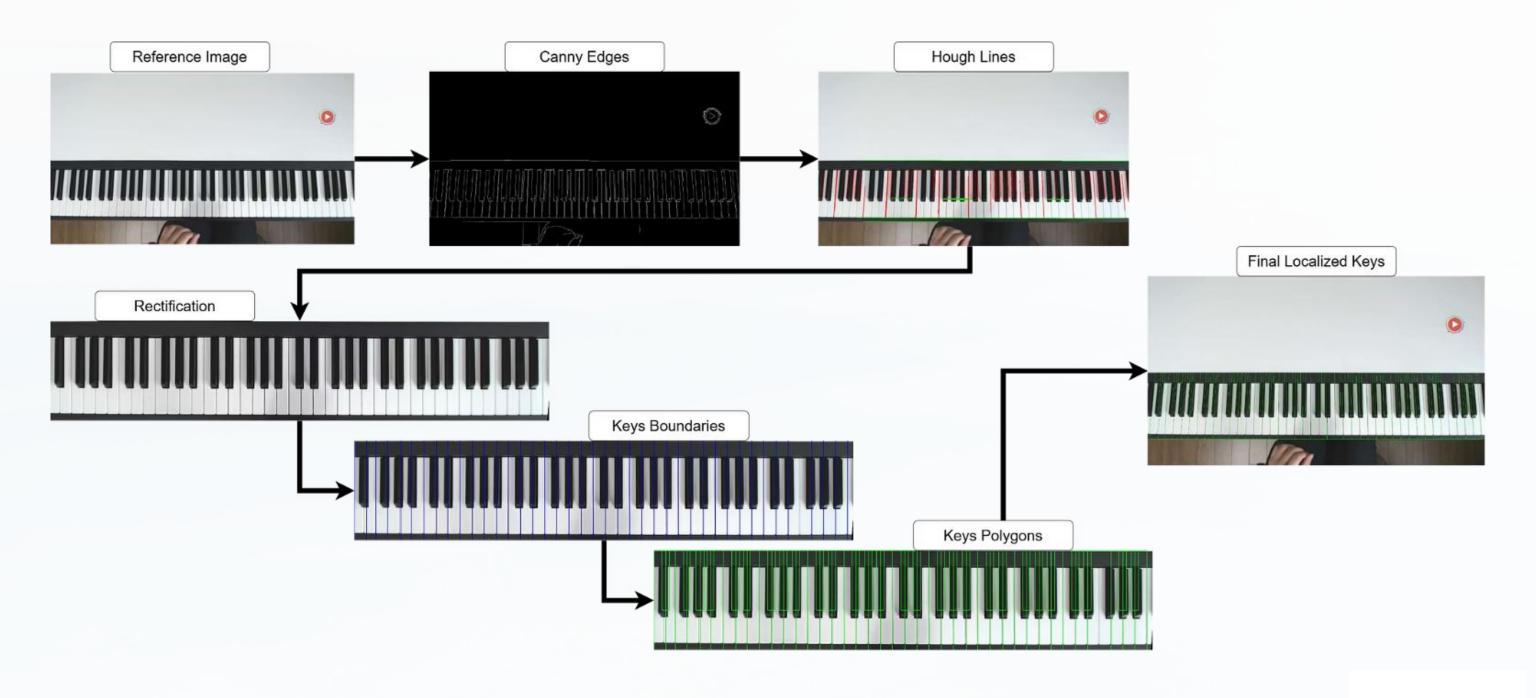


#### **Key Segmentation**

Map key polygons in rectified domain



### **Keyboard Calibration Process**





### Key-Press Detection Pipeline

#### **Hand Localization**

MediaPipe tracks fingertips in real-time

### **Candidate Mapping**

Identify keys potentially touched by fingertips

#### Photometric Validation

Compare pixel differences to confirm press

# Key-Press Detection Pipeline



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### Implementation Details

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#### Tech Stack

- Python 3.10
- OpenCV 4.11
- MediaPipe 0.10

#### **Key Components**

- Homography mapping
- Background subtraction
- MIDI synthesis

#### Performance

- ~200ms per frame
- Real-time hand tracking

### **Experimental Results**

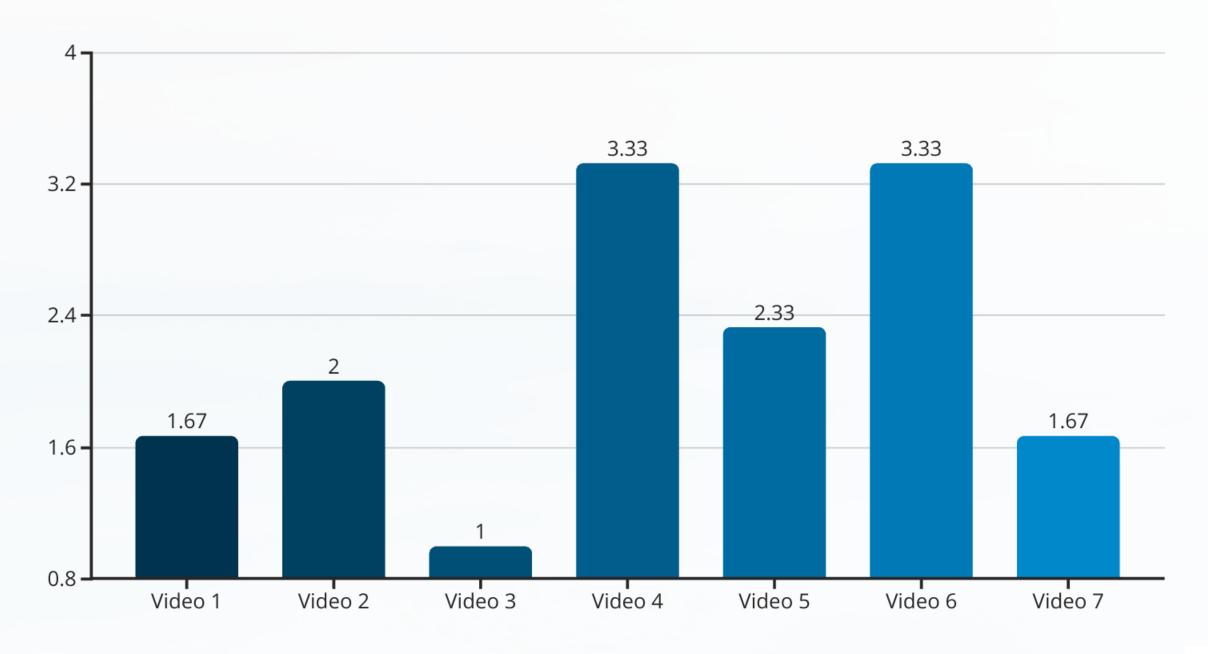
7 different videos for test cases

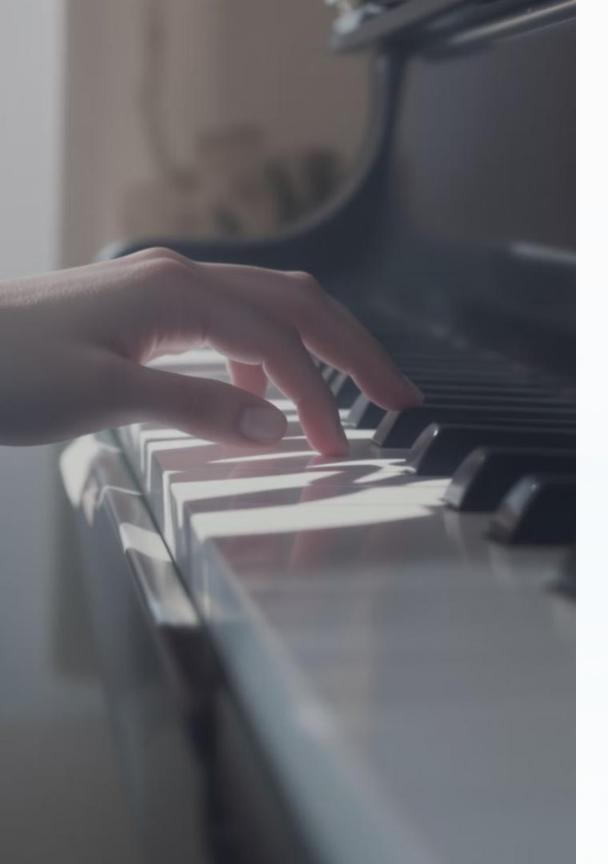
3 human testers

#### **Evaluation Scale**

- 1 Impossible to recognize
- 2 Some notes and rhythms seem to match the original song
- The song is recognizable
- The song not only is recognizable, but is pleasant to hear
- Just some mistakes were made between the original song and the simulated output

# Experimental Results (1 - 5)





### **Key Limitations**



### **Lighting Issues**

Sensitivity to reflections and shadows



### **Occlusion Problems**

Cannot resolve overlapping fingertips



### **Black Key Detection**

Poor accuracy due to minimal shadow contrast



### Parameter Sensitivity

Requires manual calibration per video

### **Future Directions**

Temporal Networks

Train on synthetic renderings

**Adaptive Calibration** 

Auto-tune parameters per video



Multi-cue Fusion

Incorporate optical flow analysis

Self-supervised Learning

Fine-tune to individual performers

### Demo: Youtube Video

Original Input



https://www.youtube.com/watch?v=P0kUMfg-dHE

### **Produced Output**

