Pose Detection

Logan Lundgren Data Analytics Capstone Project



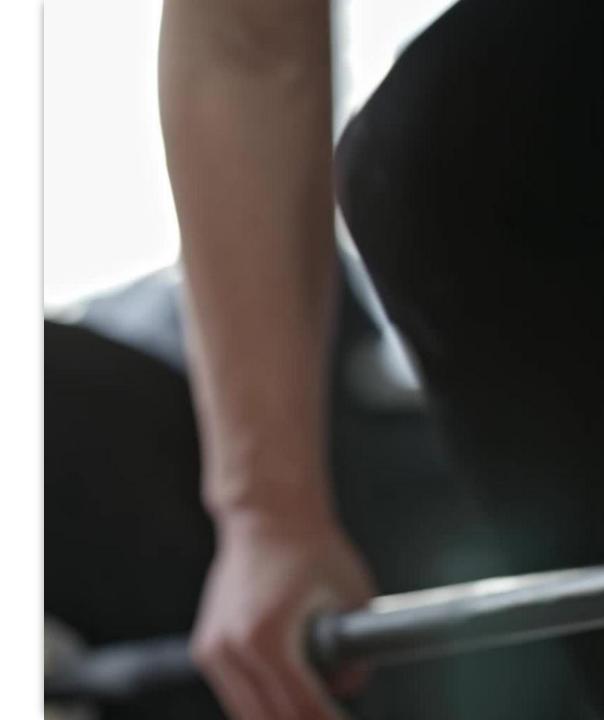
Agenda

Project Objectives

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Create a ML system that:

- Detect exercises in real-time
- Classify exercise: squat, bench, deadlift, curls, rows (potentially more if time)



Roadmap

Phase	Task	Outcome
1. Data Modeling	Gather exercise videos	Create a clean labeled dataset
2. Pose Detection	Use MediaPipe to detect body movements	Adds labeling to the videos
3. Calculated Variables	joint angles rules	Analyze form
4. Classification Model	Train a LSTM + random forest model to classify exercise	Provide real-time feedback

Dataset: "Workout/Exercises Video"

- Video data for the following exercises:
- Bicep Curl
- Barbell Bench
- Squat
- Deadlift
- TBar Rows
- ...

What is MediaPipe

MediaPipe is an open-source library by Google for real-time pose estimation.

Pre-trained movement
detection Model: No need for
large-scale training
Tracks 33 Keypoints: More
detailed than other open
source models



Data Modeling

Data Preprocessing

- Sources:
 - 22 folders, 250+ videos
 - Extracted keypoint data from videos using MediaPipe

Key Data Processing Steps:

- Store Data in Google Drive
- Extracted pose keypoints (X, Y, Z, visibility) from each frame Stored pose data in CSV format for model training
- Setup / Design Supabase Database
- Import data into Supabase



Challenges in Data Processing

- Challenges
 - Processing power (90 + minutes to run the script)
 - Troubleshooting the script to iterate through the folders correctly
 - Uploading the data to supabase

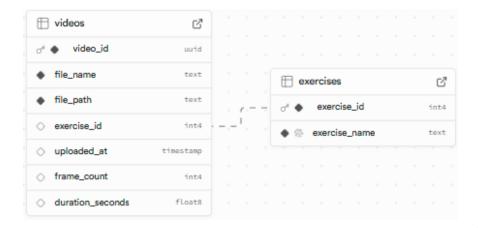


Data Features Extracted (for Model 2)

- Pose Landmark Data (CSV Format):
 - 33 landmarks per frame (X, Y, Z coordinates + visibility)
 - Key joint movements tracked: shoulders, elbows, knees, wrists
- Feature Engineering Plan: Compute joint angles

Database (Supabase)

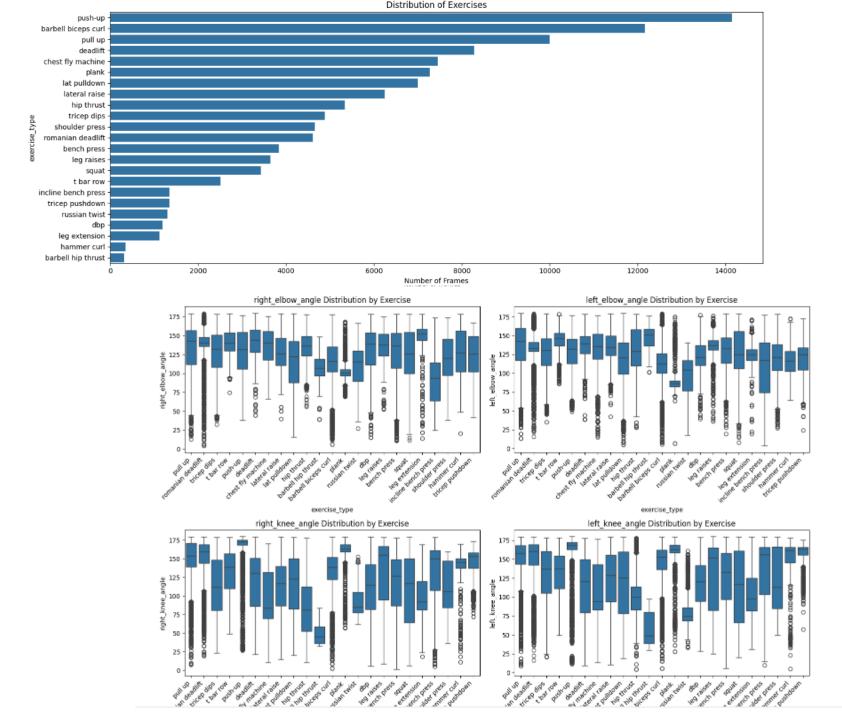
- CREATE TABLE
- video_frames
- (frame_id SERIAL PRIMARY KEY,
- video_id UUID REFERENCES videos(video_id),
- frame_number INT NOT NULL,
- timestamp_seconds FLOAT NOT NULL,
- x0 FLOAT, y0 FLOAT, z0 FLOAT, visibility0 FLOAT,
- •
- x32 FLOAT, y32 FLOAT, z32 FLOAT, visibility32 FLOAT,



Exploratory Data Analysis (By Frame)

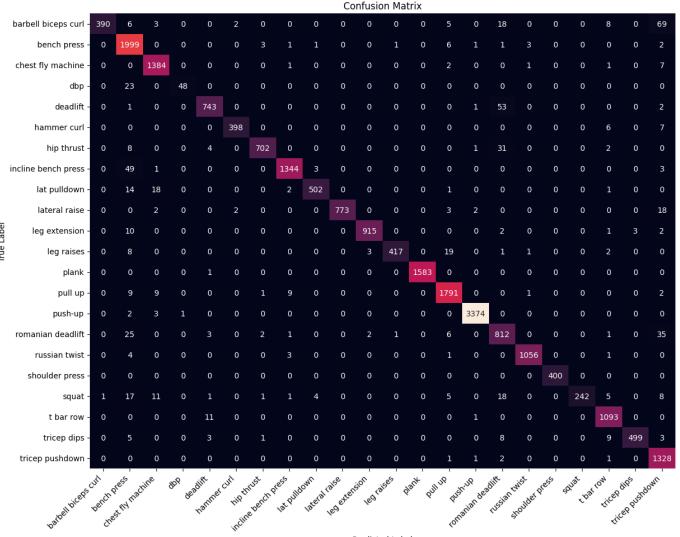
Exercise types found: exercise_type push-up 14141 barbell biceps curl 12165 pull up 10000 deadlift 8276 chest fly machine 7447 plank 7267 lat pulldown 7000 lateral raise 6245 hip thrust 5337 tricep dips 4885 shoulder press 4654 romanian deadlift 4609 bench press 3835 leg raises 3643 sauat 3428 t bar row 2506 incline bench press 1350 tricep pushdown 1346 russian twist 1305 1189 leg extension 1129 hammer curl 357 barbell hip thrust 313 Name: count, dtype: int64

Total frames: 112427 Number of exercises: 23 Calculating joint angles...



Model: Random Forest (Frames)

Classification Report:						
	precision	recall	f1-score	support		
barbell biceps curl			0.87	501		
bench press	0.92	0.99	0.95	2018		
chest fly machine	0.97	0.99	0.98	1396		
dbp	0.98	0.68	0.80	71		
deadlift	0.97	0.93	0.95	800		
hammer curl	0.99	0.97	0.98	411		
hip thrust	0.99	0.94	0.96	748		
incline bench press	0.99	.99 0.96 0.97		1400		
lat pulldown	0.98	0.93	0.96	538		
lateral raise	1.00	0.97	0.98	800		
leg extension	0.99	0.98	0.99	933		
leg raises	1.00	0.92	0.96	451		
plank	1.00	1.00	1.00	1584		
pull up	0.97	0.98	0.98	1822		
push-up	1.00	1.00	1.00	3380		
romanian deadlift	0.86	0.91	0.89	888		
russian twist	0.99	0.99	0.99	1065		
shoulder press	1.00	1.00	1.00	400		
squat	1.00	0.77	0.87	314		
t bar row	0.97	0.99	0.98	1105		
tricep dips	0.99	0.95	0.97	528		
tricep pushdown	0.89	1.00	0.94	1333		
accuracy			0.97	22486		
macro avg	0.97	0.94	0.95	22486		
weighted avg	0.97	0.97	0.97	22486		

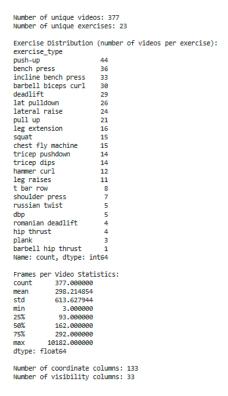


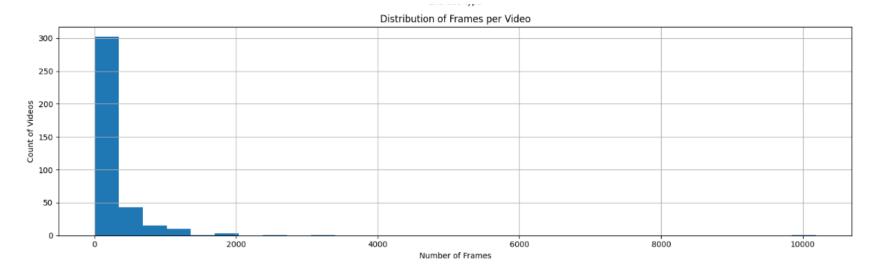
- 2500

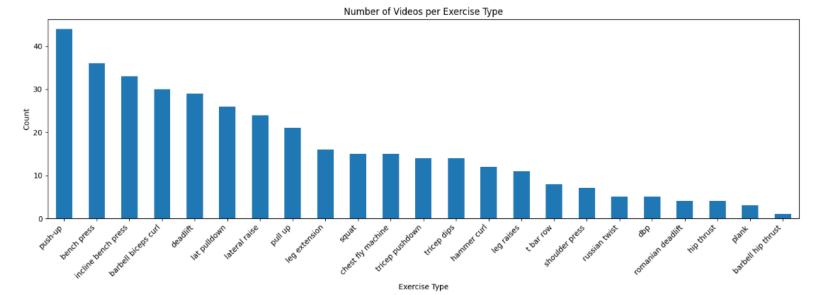
- 2000

Predicted Label

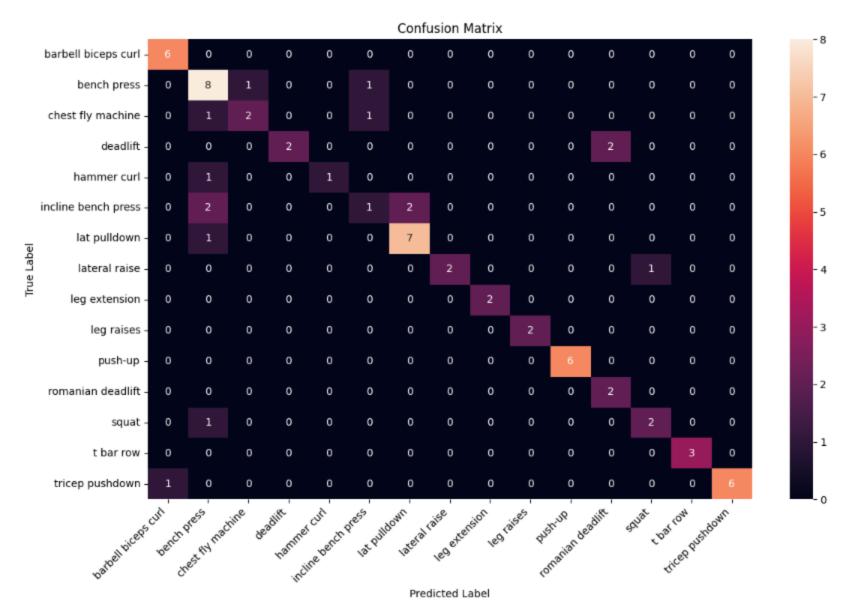
Exploratory Data Analysis (By Video)







Random Forest (Full Video)



CNN Model Output (Fail)

- Not good
- This would require more data
- Could work frame by frame, but random forest already worked

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