2023 embedded VISION SUMMIT

Developing an Embedded Vision AI Powered Fitness System

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The Peloton Guide

Transform any television in your home into an immersive boutique fitness studio powered by AI



Computer Vision Features







Smart Framing and Self Mode Guide digitally pans and zooms so you can see yourself on screen and compare your form to the instructor's form.

Activity Recognition and Movement Tracker Tracks members and recognizes their activity, as they follow along with instructors to complete moves in the class.

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Time Tracked Moves

Allows user to follow exercise led by instructor and receive real-time credit for completing movement.

Rep Tracker

Rep Tracking not only checks that user is performing the correct exercise, but also counts how many reps are done.

Other Machine Learning Features





Voice Control

Help members easily navigate content, find classes, control weights and manage their workouts with voice \rightarrow fast and hands free.



Body Activity

Body Activity shows the users various muscles worked out and provides personalized full body workout suggestions.

Machine Learning Flywheel





ML iteration is key to success: helps to find data gaps, data quality issues and enables focused model enhancements

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Datasets



Real-world datasets

ReferenceandVideo QAandLabel QAmaterial creationattributeattributeattribute	Reference material creation	and	Video QA	and	Label QA	
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Synthetic datasets







CV Algorithms



Deep Match

Template matching using metric learning



Deep Move

Action recognition with temporal shift buffers

Diversity Attributes for Metrics



01	Environmental attributes	 Background: walls, flooring, furniture, windows Lighting, shadows, reflections Other people, animals in the field of view Occlusion Equipment (mat, dumbbells)
02	Member attributes	 Gender, skin tone Body type Fitness level Disabilities Clothing
03	Geometric attributes	 Camera/Member relative placement Camera mounting Height Camera tilt Member orientation Member distance





<u>Equality of opportunity</u>: For a preferred label (specific exercise is being performed) and a given attribute, a classifier predicts that preferred label equally well for all values of that attribute. True Positive Rates are shown in table below.

Exercise	Median	Attribute: Body Type			Fitzpatrick Skin Tone			Gender	
		Underweight	Average	Overweight	1-2	3-4	5-6	Man	Woman
Crossbody Curl	97%	100%	100%	97%	100%	98%	100%	100%	98%
Squat	99.7%	100%	100%	100%	100%	100%	100%	100%	100%
Dumbbell Swing	96%	100%	96%	98%	98%	100%	97%	97%	100%

Robustness Driven through UI and Nudging



Edge cases nudging:

- Brightness
- Occlusion
- Camera tilt, distance
- Multiple people



Orientation nudging occurs in all time-based and rep tracking workouts to minimize inaccuracies due to occlusion.

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Conclusion



- Rapid ML iteration helps find data gaps, data quality issues and enable focused model enhancements
- Diversity attributes are important to ensure balanced training, but also key to slice and dice performance
- ML and UI work together to ensure that member has a robust experience without introducing friction
- Both objective metrics focused on worst case performance + subjective metrics measurements are necessary to drive improvements

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Peloton AI principles

Blazepose

Temporal shift module

Early stage recommender systems

Context aware recommender systems

User experience platform for connected fitness

