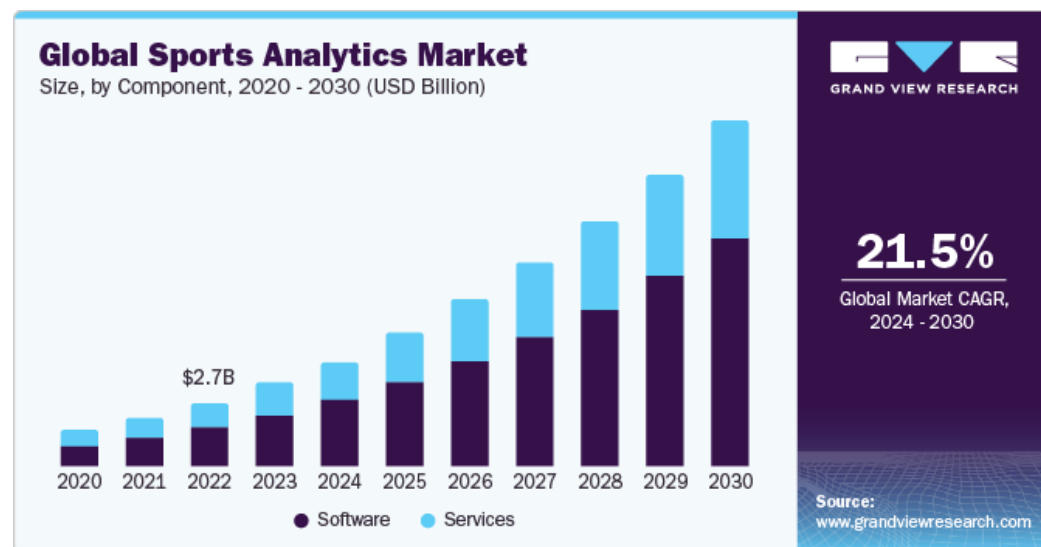


Sporty-UP: A Mobile Application for Real-Time Posture Assessment and Feedback in Indoor Sports

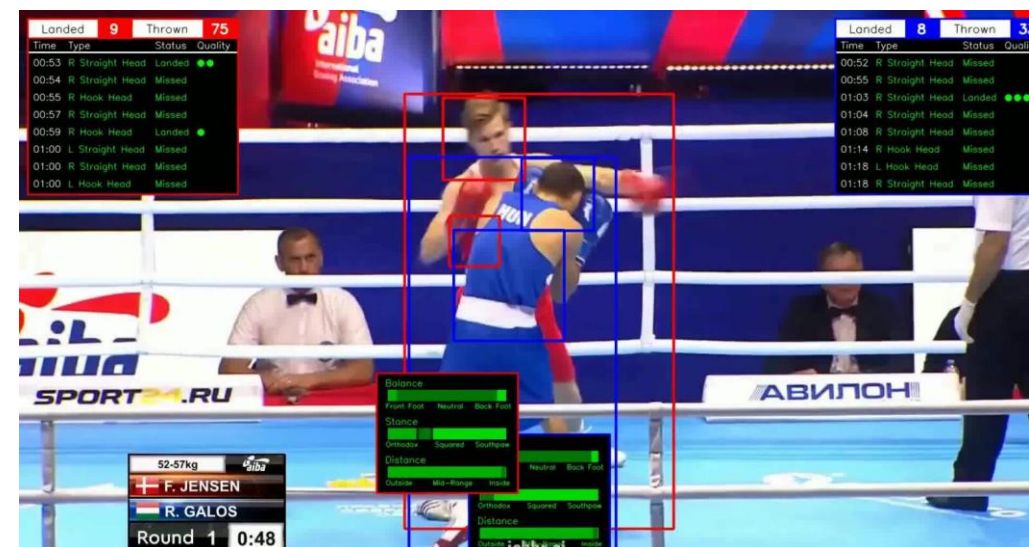
Seungtaek Lim, Juwan Son, Junhyeon Hwang, Woojin Jang
Hansung University

Growth of the Sports Industry and the Rise of Advanced Technologies

- The global **AI in sports** market is projected to grow from **USD 1.2 billion in 2024** to **USD 4.7 billion by 2034**, with a CAGR of **14.7%**.
- AI is increasingly being used for performance analysis, injury prevention, strategy optimization, and enhancing fan engagement.



Global Market Insights – <https://www.gminsights.com/industry-analysis/ai-in-sports-market>



An AI-based statistical analysis program for combat sports, DeepStrike

Growing Demand for Indoor Sports

- The COVID-19 pandemic has led to a significant increase in interest in **indoor sports**.
- Expansion of Family-Oriented Entertainment Facilities



Indoor Sports Hub Tennis Courts



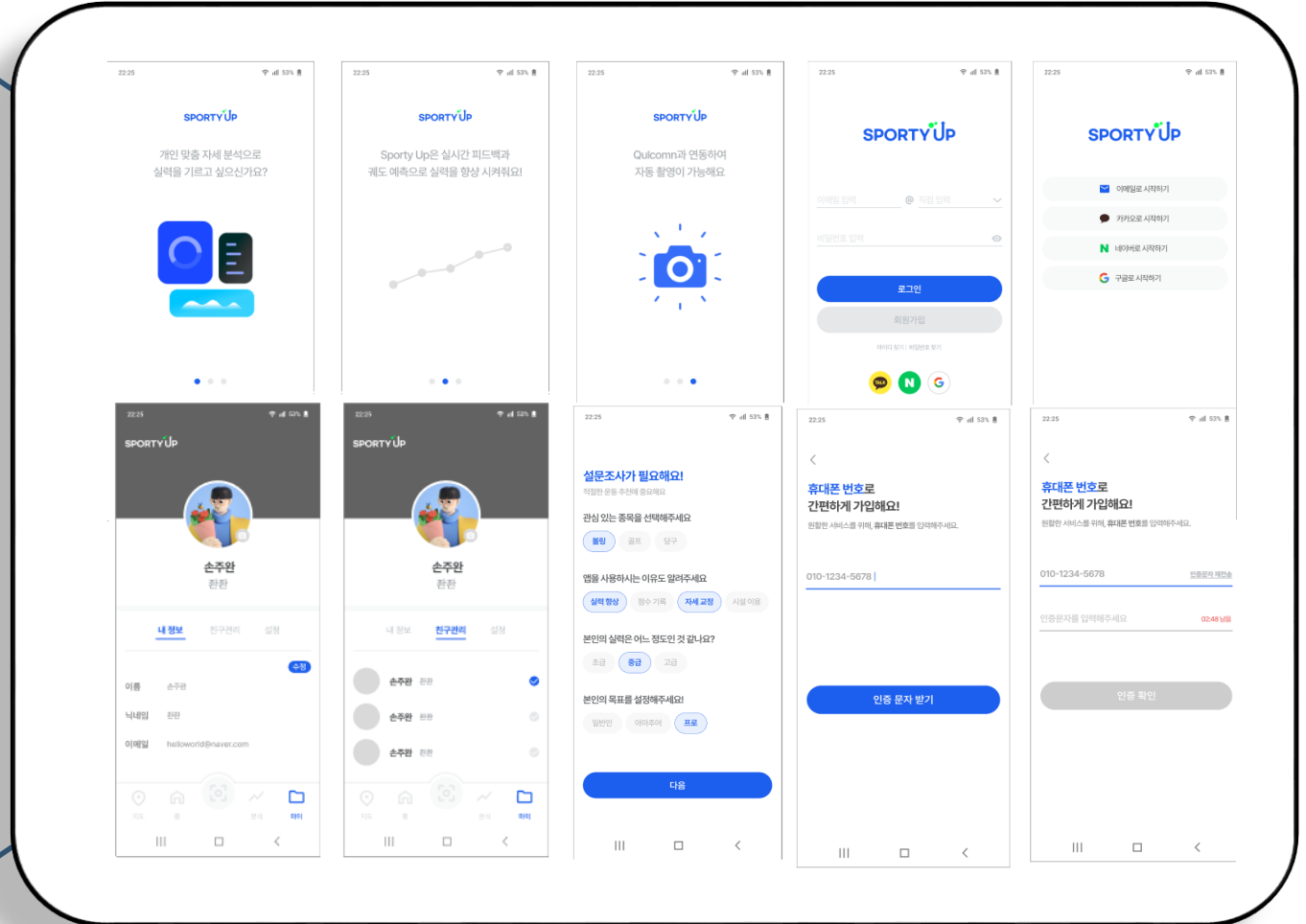
Richardson's Family Entertainment Centre

Sporty-up : A Mobile Application for Real-Time Posture Assessment and Feedback in Indoor Sports



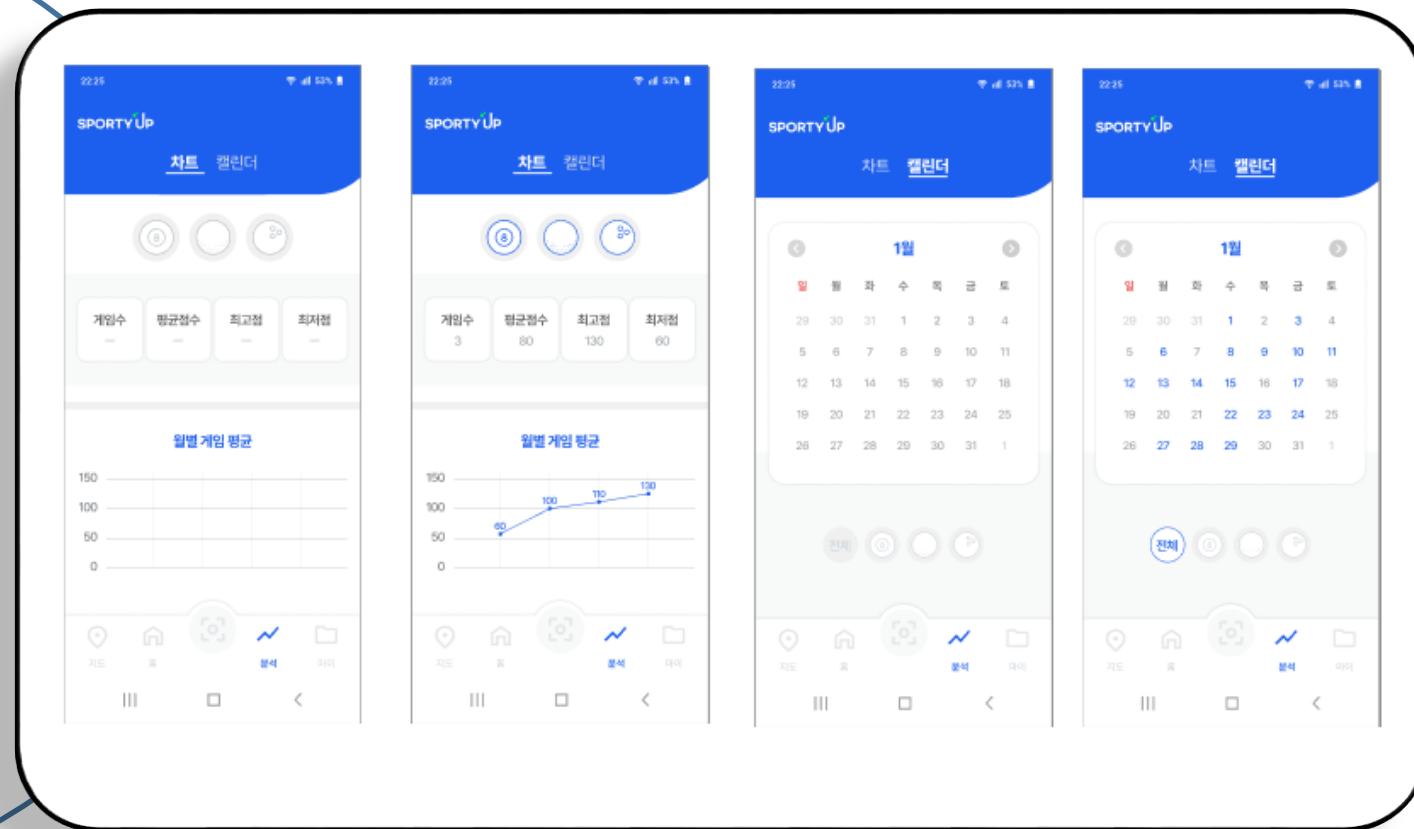


Community-driven experience sharing



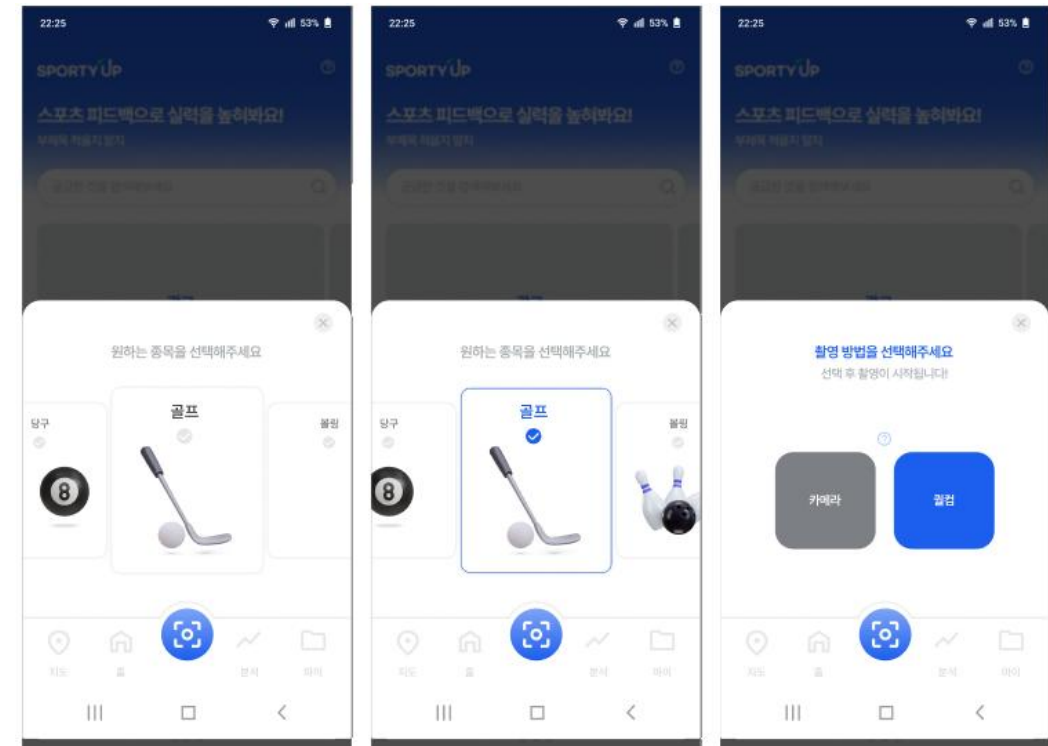


Personal data tracking and visualization



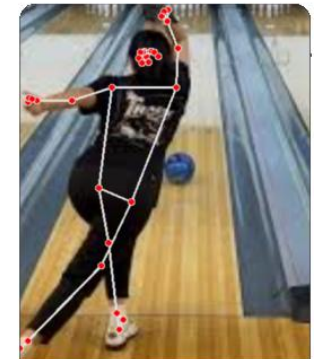
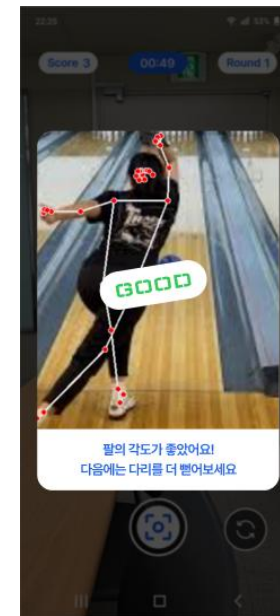


Real-time feedback system & High-quality feedback through mobile devices



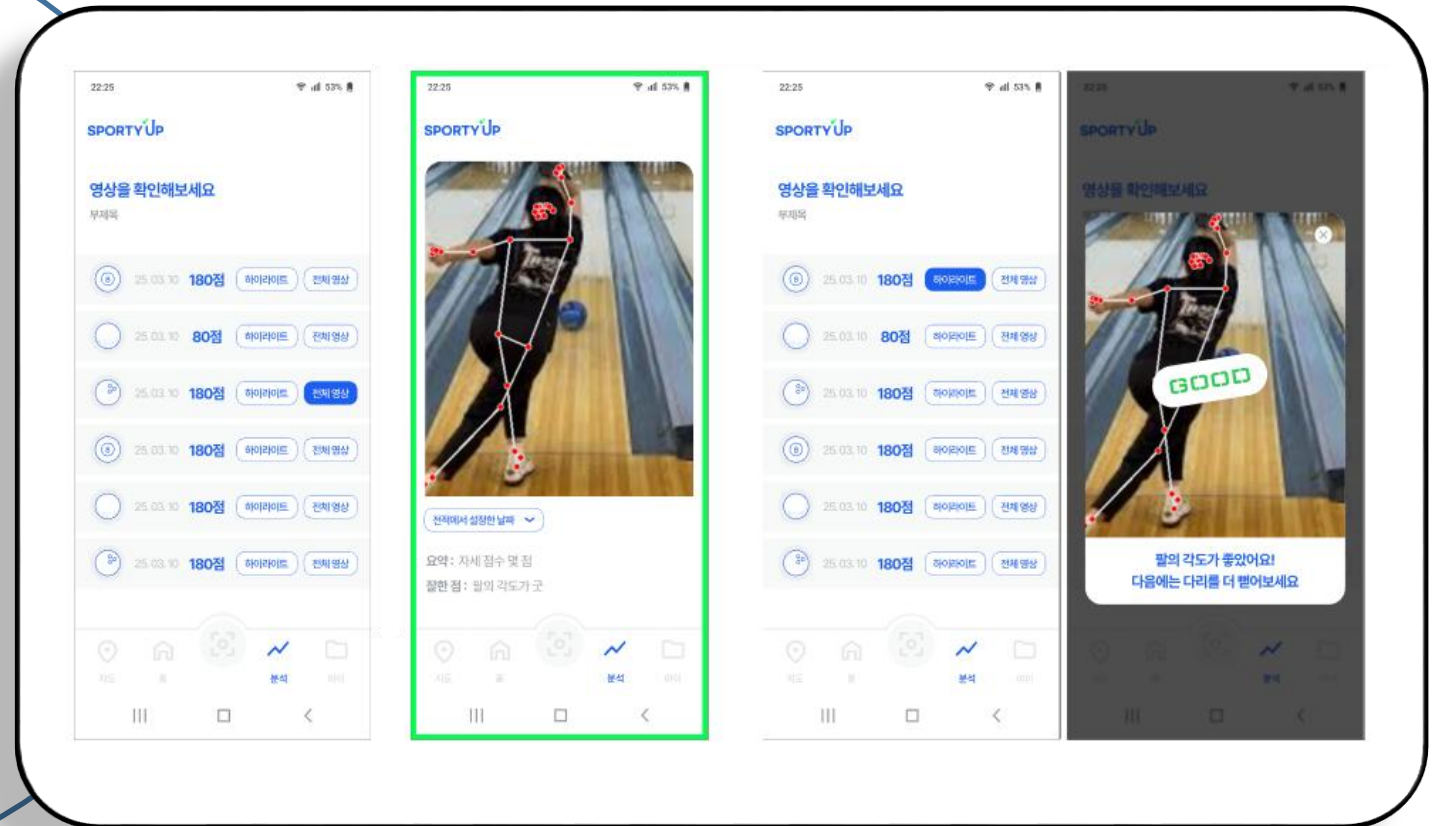


Real-time feedback system & High-quality feedback through mobile devices





Real-time feedback system & High-quality feedback through mobile devices



Technical Challenges in Mobile-Based Real-Time Posture Feedback

1

- **Limited computational resources and memory**

- Mobile devices have limited processing power and memory, making it difficult to run complex AI models effectively.

2

- **Technical burden of real-time processing**

- Real-time feedback requires both high speed and accuracy, placing significant demands on system performance.

3

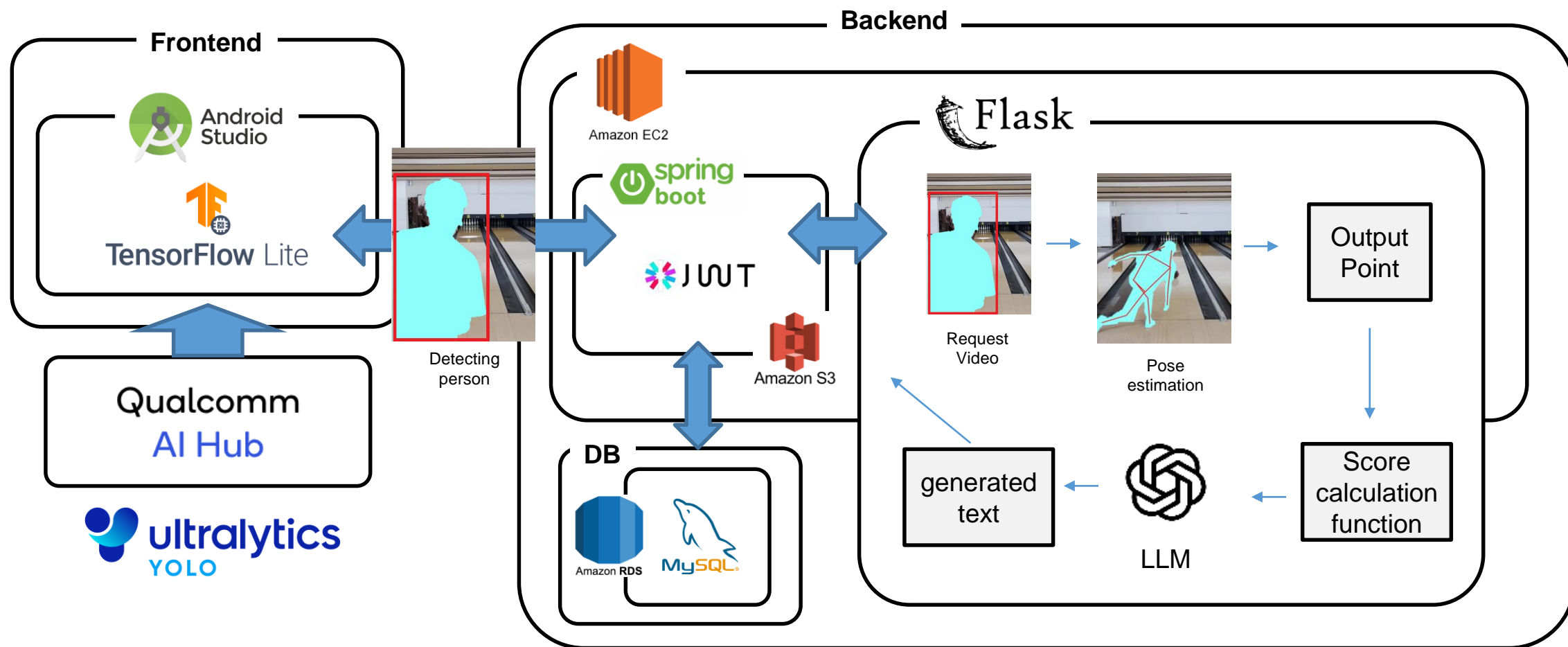
- **Inability to handle complex analysis**

- On-device processing alone has limitations when it comes to performing advanced or high-level posture analysis.

System Design Strategy ① – On-Device AI with Server Integration

- **On-device AI utilization (Qualcomm AI Hub-based)**
 - Automatic determination of real-time capture timing
 - Avoidance of unnecessary computation → performance optimization
- **Server-based API integration**
 - High-precision analysis performed on the server.
 - Ensuring high accuracy in a mobile environment
- **Hybrid structure of on-device AI and server AI**
 - Achieves a balance between fast responsiveness and high-precision analysis

System Design Strategy ① – On-Device AI with Server Integration



- **Device Fit Optimization**
 - Performance adjustment tailored to the target device
 - Real-time execution on the device

System Design Strategy ③ - Active Utilization of Qualcomm AI Hub

Model
Target: TensorFlow Lite mnl3jz3n

Show tensor shapes

Model overview

Model ID mnl3jz3n

Type TensorFlow Lite

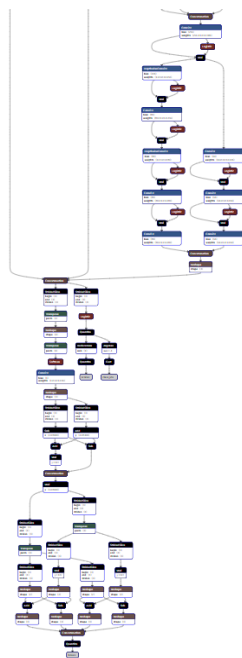
Model attributes

description "Created by Qualcomm AI Hub."

version 3

Showing 377 / 377 Filter

Name	Type	Unit
image	Input	
[Quantize]	Quantize	
[Pad]	Pad	
[Conv2D]	Conv2D	
[Logistic]	Logistic	
[Mul]	Mul	
[Pad]	Pad	
[Conv2D]	Conv2D	



YOLOv11 감지

Snapdragon® 8 Elite 모바일
스냅드래곤 8 엘리트 QRD

토치스크립트 → 티플라이트

1.11 미시시피
추론 시간

0 - 27 엠비
메모리 사용량

372 NPU
레이어

- **Pre-performance verification through profiling**
 - Prediction of performance (speed, memory usage, etc.) on the target device

Stage	Time	Memory
Compilation ⓘ	0.0 ms	0.0 MB
First App Load ⓘ	697.6 ms	56 - 63 MB
Subsequent App Load ⓘ	694.2 ms	8 - 51 MB
Inference ⓘ	(min) 4.3 ms (median) 4.5 ms	0 - 31 MB



- **Utilization of model visualization tools**
 - Intuitive understanding of the model architecture
- **Ensuring stability through pre-deployment testing**
 - Final verification tailored to the device environment
 - Improved processing performance and predictability



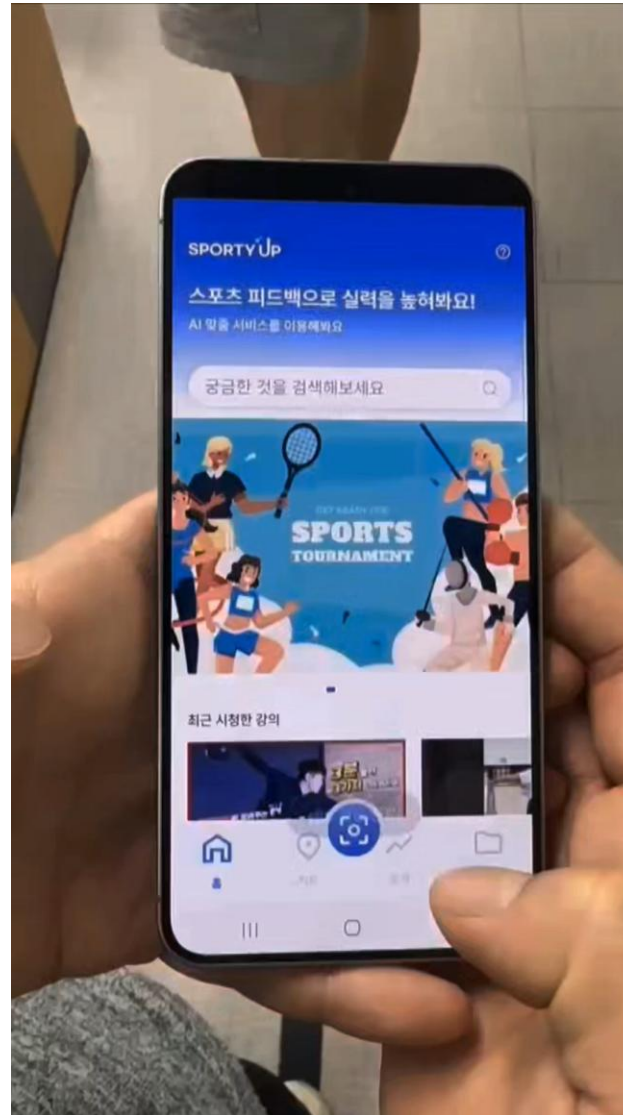
Conclusion

- Performance Comparison: Before and After Optimization

Platform: Snapdragon 8 Gen 1 Elite QRD
Runtime: TensorFlow Lite (TFLite)
Quantization: W8A8 (Weight & Activation 8-bit)

Metric	Full Precision (Before)	Optimized (W8A8 + TFLite + NPU)	Improvement
Min Inference Time	4.3 ms	1.1 ms	74% faster
Median Inference Time	4.5 ms	1.1 ms	4x speed up
Peak Memory Usage	31 MB	26 MB	16% less memory
Compute Operations	359 ops	373 ops (optimized)	Better NPU utilization
First App Load Time	697.6 ms (56~63MB)	396.4 ms (27~35MB)	Faster & lighter
Subsequent Load Time	694.2 ms (8~51MB)	394.8 ms (0~36MB)	Optimized loading
Accuracy (PSNR)	N/A	Boxes: 37.02 dB / Scores: 37.7 dB	Accuracy preserv

Demo



Thank you!