Badminton Al Analysis: LangGraph Orchestration Pipeline

ShreeRaj Mummidivarapu June 13, 2025

What is Agentic AI?

- Traditional ML/AI Systems (e.g., CNNs, Vision Transformers, LLMs):
 - CNNs (Convolutional Neural Networks): Excellent for image classification, object detection. Primarily pattern recognition.
 - Vision Transformers (ViTs): Leverage self-attention for image tasks, capturing global dependencies. Still largely reactive.
 - LLMs (Large Language Models): Powerful for text generation, understanding. Can exhibit emergent reasoning but lack inherent agency or persistent state.
 - Common Characteristics:
 - Monolithic & Reactive: Designed for specific tasks, respond to input without internal goals or long-term planning.
 - Limited Self-Correction: Require retraining for significant behavioral changes.
 - No Persistent State: Each interaction is often independent, lacking memory across sessions.

Agentic Al Systems

- Composed of autonomous agents with specific roles.
- Possess capabilities such as:
 - Planning
 - Memory
 - Tool use
 - Self-reflection
- Can break down complex tasks, orchestrate actions, and adapt to new information.
- Aim for more human-like problem-solving, decision-making, and continuous learning.
- **Key Capabilities:** Planning, memory, tool use, self-reflection, and dynamic adaptation.
- **Examples:** Autonomous research agents, complex task automation systems, adaptive control systems.

LangGraph Pipeline for Badminton Analysis (1/2)

• Multimodal Data Integration:

- Effective integration of video analysis (pose metrics) and audio transcription in a unified pipeline.
- Sequential processing with comprehensive error handling for reliable results.

LangGraph for Sports Analysis:

- Pioneering the use of LangGraph for structured pipeline orchestration in sports analytics.
- Foundation for future expansion to more complex agent interactions and workflows.

Actionable, Granular Feedback:

- Focus on generating highly specific, actionable feedback for players and coaches.
- Moves beyond descriptive statistics to prescriptive recommendations.

• Scalable & Extensible Architecture:

 Modular pipeline design allows for easy addition of new analysis capabilities in future iterations.

LangGraph Pipeline for Badminton Analysis (2/2)

Holistic System View:

 Our system seamlessly combines video (player movement) and audio (speech transcription) for comprehensive understanding.

• Linear Pipeline Orchestration:

- Utilizes a four-node LangGraph pipeline for video processing, audio processing, and report generation.
- Sequential processing with state management for efficient data flow between components.

• Automated, Granular Reporting:

- Generates detailed, objective reports with actionable insights.
- Identifies strengths and areas for improvement to support targeted coaching decisions.

How It Works: High-Level Overview

Input:

• Badminton match video.

Processing:

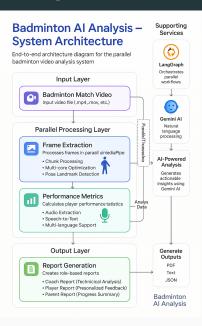
- Video frames are processed for pose estimation and audio is transcribed for speech content.
- LangGraph orchestrates a linear pipeline with four processing nodes for efficient data flow.

Output:

 Comprehensive text and PDF reports with actionable insights and strategic recommendations.

System Pipeline Overview

System Architecture Diagram



System Pipeline: High-Level Walkthrough (1/2)

• 1. Video Input:

- Raw match footage (e.g., MP4 files).
- Supported formats: MP4, AVI, MOV with H.264/H.265 encoding.
- Optimal resolution: 1080p (1920×1080) at 60fps for detailed motion capture.

• 2. Frame & Audio Extraction:

- Video is processed to extract individual frames for visual analysis.
- Audio track is separated and processed for speech transcription.
- Preprocessing includes frame resizing and color conversion.

System Pipeline: High-Level Walkthrough (2/2)

• 3. Pipeline Orchestration with LangGraph:

- Extracted data flows through a four-node linear pipeline orchestrated by LangGraph.
- Pipeline includes video processing, audio transcription, data integration, and report generation.
- State management ensures efficient data flow between processing steps.

• 4. Report Generation:

- Consolidated data is processed by Google Gemini API to generate comprehensive text and PDF reports.
- Reports include performance metrics, observations based on pose data, and actionable feedback.
- Customizable templates based on user role (player, coach, analyst).

Detailed Technical

Implementation

LangGraph Pipeline: Architecture and Integration

• Pipeline Architecture:

- Linear processing pipeline with four main nodes:
 - Video processing
 - Audio processing
 - Data integration
 - Report generation
- Simple Directed Acyclic Graph (DAG) with sequential flow and error-handling edges.
- Supports both synchronous and asynchronous execution using asyncio.

• Integration with Python Ecosystem:

- Integrates with asyncio for non-blocking execution.
- Compatible with:
 - MediaPipe (vision)
 - Google Web Speech API (audio)
 - Gemini API (report generation)

Key Components: Vision and Audio Nodes

Video Processing Node:

- Primary Function: Analyzes video frames for player pose
- Technical Implementation: Media Pipe Pose model for human pose detection and tracking.
- Key Capabilities: Pose estimation, elbow angle calculation, wrist distance measurement.
- Output: Structured JSON with timestamped keypoints and performance metrics.

• Audio Processing Node:

- Primary Function: Extracts and transcribes speech from the video's audio track.
- Technical Implementation: PyDub for audio extraction; Google Web Speech API for transcription.
- Key Capabilities: Multi-language support, silence-based segmentation.
- Output: Transcribed text of spoken content.

Report Generation: Technical Implementation

• Data Aggregation:

- Input Sources: Combines pose metrics from video analysis and speech transcription from audio.
- Data Sampling: Processes first 100 pose metrics to manage context size for LLM processing.
- JSON Formatting: Structures data in standardized format for AI model consumption.

• Natural Language Generation:

- Al Model: Google's Gemini 1.5 Flash model generates contextual, role-specific reports.
- Role-Based Prompting: Custom system prompts tailored to coach, student, or parent perspectives.
- Personalization: Adapts language, technical depth, and focus areas based on target audience.

• Multilingual Support:

• Language Options: Reports available in multiple languages (English, Hindi, Tamil, Telugu, Kannada).

Novelty and Innovation

Multimodal Data Integration:

- Effective integration of video analysis (pose metrics) and audio transcription in a unified pipeline.
- Sequential processing with comprehensive error handling for reliable results.

LangGraph for Sports Analysis:

- Pioneering the use of LangGraph for structured pipeline orchestration in sports analytics.
- Foundation for future expansion to more complex agent interactions and workflows.

Actionable, Granular Feedback:

- Focus on generating highly specific, actionable feedback for players and coaches.
- Moves beyond descriptive statistics to prescriptive recommendations.

Scalable & Extensible Architecture:

 Modular pipeline design allows for easy addition of new analysis capabilities in future iterations.

System Implementation: Technical Stack

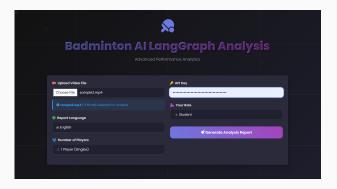
Frontend Technologies:

- Web Interface: HTML5, CSS3, JavaScript with responsive design.
- Video Upload: Custom file uploader with format validation and progress tracking.
- Report Viewer: Interactive PDF viewer with annotation capabilities.

Backend Technologies:

- Server: Flask for web application serving and file handling.
- Video Processing: OpenCV, MediaPipe for pose estimation and tracking.
- Audio Processing: Google Web Speech API for audio transcription.
- Pipeline Orchestration: LangGraph for linear pipeline definition and state management.
- **LLM Integration:** Google Gemini API for report generation with custom prompt templates.
- Report Generation: Custom templates for role-based and multilingual reports.

Demo Step 1: Upload Video



- User uploads a badminton match video and all necessary through the web interface.
- Supported formats include MP4, AVI, MOV.
- The system initiates pre-processing after recieving all information

Demo Step 2: Processing Status



- System displays processing stages with real-time progress indicators.
- Separate modules handle video frame extraction, pose detection, and audio transcription.
- Users are informed of each module's completion status.

Demo Step 3: Interactive Report



- An interactive, browser-based report is generated after processing.
- Key metrics like elbow angles, wrist distances, and speech insights are displayed.
- Visual overlays and summaries make interpretation intuitive.

Demo Step 4: Download PDF Report



- Users can explore specific sections in detail through the interface.
- A downloadable PDF report summarizes key findings for offline review.
- Report layout adapts to role coach, player, or parent.

Demo Walkthrough: User Experience Summary

- Complete Journey Overview:
 - **Step 1: Upload** Upload badminton match video via a simple interface.
 - Step 2: Process Monitor real-time progress with clear indicators.
 - Step 3: Report View See insights in a rich, interactive web view.
 - **Step 4: Export** Download professional reports for long-term use.
- Focus on Usability: Designed for non-technical users (coaches, athletes).
- Future Enhancement: Mobile-friendly UI and real-time streaming support.

Thank You!

Questions?

Demo Video

► Click here to watch on YouTube

GitHub Repository

Click here to view the code

Contact

