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# Patient-Specific 3D Models for Adaptive Physical Therapy Plans

Tailoring Rehabilitation Exercises Based on Post-Surgical Anatomical Data

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#### **Abstract:**

Recent innovations in 3D anatomical modeling have significantly advanced pre-operative surgical planning; however, their application to post-operative rehabilitation remains underutilized. This paper examines a comprehensive framework for employing patient-specific 3D anatomical reconstructions to design adaptive physical therapy regimens. By leveraging individualized post-surgical imaging data, we detail an approach that enables clinicians to prescribe biomechanically optimized exercises tailored to each patient's recovery trajectory. The architecture integrates edge AI, wearable sensors, and kinematic simulations to provide real-time feedback and facilitate dynamic therapy adjustments. This system promotes safer, more effective, and engaging rehabilitation, with reproducibility and clinical scalability as core design objectives.

Keywords: Patient-specific modeling, adaptive rehabilitation, edge AI, biomechanical simulation, post-operative care.

#### 1. INTRODUCTION

Conventional rehabilitation paradigms following surgical interventions often rely on generalized protocols that insufficiently account for patient-specific biomechanical conditions. This lack of personalization can hinder recovery, potentially leading to suboptimal outcomes, re-injury, or patient noncompliance. The advent of 3D anatomical modeling, particularly from volumetric medical imaging data, presents an opportunity to tailor therapy plans based on individualized musculoskeletal constraints and capacities.

Technological convergence in medical imaging, biomechanical simulation, and edge computing has enabled the translation of surgical 3D models into actionable assets for physical therapy. This paper delineates a system-level methodology for generating patient-specific 3D models from CT/MRI data, simulating physical constraints, and integrating wearable sensor feedback via edge AI. The proposed approach facilitates adaptive therapy that evolves in parallel with the patient's recovery.

# 2. SYSTEM ARCHITECTURE AND WORKFLOW

# 2.1 Data Acquisition and 3D Model Generation

The pipeline commences with the acquisition of high-resolution DICOM-format imaging data—either CT or MRI—which is subsequently processed through semi-automated or machine learning-enhanced segmentation algorithms. Anatomical structures of clinical interest (e.g., osseous components, tendons, musculature, and implants) are segmented and rendered into topologically accurate, watertight surface meshes. Software tools such as 3D Slicer and Mimics are employed for segmentation, while mesh optimization is executed using platforms like Meshlab and Blender.

### 2.2 Workflow Diagram

Flowchart 1.1 - Adaptive Therapy Model Pipeline

[Radiological Imaging Acquisition]  $\rightarrow$  [Volumetric Segmentation Pipeline]  $\rightarrow$  [Surface Mesh Reconstruction]  $\rightarrow$  [Computational Kinematic Modeling]  $\rightarrow$  [Clinical Interface with AI-Augmented Feedback]  $\rightarrow$  [Patient-Specific Rehabilitation Execution]



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#### 2.3 Biomechanical Simulation

With a validated 3D anatomical model, computational simulations are conducted to evaluate range of motion (ROM), joint kinematics, and stress distribution. These simulations leverage platforms like OpenSim and AnyBody Technology to facilitate forward dynamics modeling. By simulating movement scenarios, therapists can assess the safety and biomechanical appropriateness of prospective exercises before clinical implementation.

# 2.4 Integration with Edge AI for Feedback

Real-time kinematic data is captured using inertial measurement units (IMUs) or camera-based motion capture systems. This data is processed locally using lightweight AI models deployed on edge devices to maintain latency constraints and data privacy. Edge inference algorithms detect deviations from expected motion patterns, perform load symmetry analysis, and generate adaptive feedback for the therapist and patient alike.

# 2.5 Therapist Dashboard and Personalization Interface

Clinicians interact with a secure, HIPAA-compliant web interface that integrates 3D anatomical visualizations, motion capture overlays, and real-time performance analytics. This platform enables:

- Dynamic visualization of patient-specific musculoskeletal models.
- Access to AI-generated adherence and compliance metrics.
- Programmatic adjustment of rehabilitation protocols in response to live data.

#### 3. IMPLEMENTATION CONSIDERATIONS

# 3.1 Data Security and Compliance

The architecture is designed with rigorous adherence to healthcare data protection standards such as HIPAA and GDPR. All sensitive patient data is encrypted both in transit and at rest. Wherever feasible, computations involving identifiable data are performed on-device or within institutionally governed edge environments, mitigating cloud exposure risks.

# 3.2 Scalability and Performance

System performance is sustained through containerized microservices and modular deployment strategies. AI models such as MobileNetV3 and TinyPose are optimized for edge inference on consumer-grade hardware, ensuring accessibility for both clinical and at-home use cases.

# 3.3 Interoperability

The platform adheres to HL7 FHIR standards to ensure seamless interoperability with electronic medical record (EMR) systems. It supports bidirectional data exchange with surgical planning software, radiology information systems, and physiotherapy documentation interfaces.

## 4. CASE STUDY: ACL RECONSTRUCTION REHABILITATION

A clinical application involved a patient recovering from anterior cruciate ligament (ACL) reconstruction. Post-operative MRI scans were used to generate a high-fidelity 3D model of the femur-tibia complex. Using this model, joint dynamics were simulated to determine safe parameters for knee flexion and extension. Early rehabilitation excluded deep flexion exercises based on modeled strain profiles. Wearable sensor data tracked patient compliance, while AI-driven analysis enabled the progressive introduction of more demanding movements. The approach reduced the overall rehabilitation period by three weeks compared to standard protocols and increased adherence due to interactive feedback features embedded in the patient's mobile application.

# 5. CONCLUSION

The use of patient-specific 3D anatomical models in adaptive physical therapy represents a paradigm shift in post-operative rehabilitation. This methodology enables a biomechanically informed, data-driven approach to exercise planning that aligns precisely with a patient's evolving physiological state.



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By integrating edge AI, motion analytics, and personalized visualization tools, the system empowers clinicians to optimize therapy intensity, prevent overexertion, and enhance patient engagement. Its applicability spans multiple domains, including orthopedics, neurology, and cardiopulmonary rehabilitation, and offers transformative potential in both inpatient and outpatient contexts.

Future research will focus on expanding the system's capabilities to include multiscale musculoskeletal load prediction, increasing accessibility through camera-only motion tracking, and validating outcomes through longitudinal clinical trials. These developments aim to solidify the role of precision modeling in enhancing recovery trajectories and overall therapeutic efficacy.

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