

# AI-Driven Personalized Cancer Treatment App using React.js, Vite, and Gemini AI

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## Abstract

This paper presents the design and implementation of a web-based, AI-driven personalized cancer treatment platform leveraging React.js, Vite, and Gemini AI models. We address the pressing need for individualized oncology care by facilitating rapid, secure, and dynamic recommendations based on patient-specific data. The proposed architecture integrates state-of-the-art generative AI for clinical decision support, delivers a fast and robust frontend with modern web technologies, and ensures user data privacy and compliance. We discuss system design, core AI workflow, frontend engineering, integration challenges, and preliminary evaluation results, demonstrating the effectiveness and promise of such technology in aiding clinicians and patients.

## 1. Introduction

## 1.1 Background

Cancer remains a leading cause of death worldwide, with treatment outcomes heavily influenced by patient biology, cancer genomics, and co-morbidities [1]. Personalized medicine aims to tailor interventions to the unique molecular profile of each patient. However, translating vast, heterogeneous clinical data into actionable treatment plans remains challenging, often overwhelming clinicians.

Recent advancements in artificial intelligence (AI), especially large language models (LLMs) like Google Gemini, can parse multidimensional clinical records, infer treatment strategies, and summarize evidence at unprecedented speed. When combined with modern web development frameworks such as React.js and rapid bundlers like Vite, these insights can be delivered interactively and securely to both clinicians and patients.

## **1.2 Objective**

We propose "OncoGenie": an AI-driven personalized cancer treatment web application implementing React.js for the frontend, Vite for build tooling, and Gemini AI for clinical decision support. Our research details the app's development, integration methodology, data flows, and AI utility in real-world scenarios.

## 2. Related Work

Existing solutions—such as IBM Watson for Oncology or DeepMind's clinical AI projects [2][3]—demonstrate the power of AI for medical decision-making but often lack seamless user interfaces, rapid prototyping, or public

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accessibility. Open-source tools typically fail to provide AI-level clinical reasoning. Our approach contrasts by emphasizing:

- Real-time, patient-specific recommendation generation via LLMs.
- Modular, scalable, and performant web engineering.
- Emphasis on explainability, user privacy, and interactivity.

## 3. System Overview

## 3.1 Architecture

The system comprises three core components:

- 1. Frontend React.js, built with Vite for fast development and loading.
- 2. Backend API Gateway handling requests, patient data processing, and securing Gemini AI interactions.
- 3. AI Reasoning Module Gemini AI model(s) generating recommendations, treatment summaries, and risk assessments.

## 3.2 Data Flow

- 1. User logs in, submitting structured (forms) or unstructured (PDFs/notes) clinical data.
- 2. Data is pre-processed, anonymized, and sent via backend to Gemini AI.
- 3. AI returns evidence-based treatment options, prognostics, clinical trial matches, and explanations.
- 4. Recommendations are rendered asynchronously in the UI, with options for user queries, visualization, and export.

## 4. Technology Stack

## 4.1 React.js

Chose React for building modular, component-based UIs, ensuring maintainability and interactivity. React hooks and context APIs power dynamic data flows and state management [4].

## 4.2 Vite

Vite offers lightning-fast build and hot module reloading, improving developer experience and reducing timeto-production. It supports code-splitting and TypeScript integration.

## 4.3 Gemini AI

Gemini, Google's multimodal foundation model, was selected for its clinical knowledge, ability to process text, images, and tabular data, and its strong performance on medical reasoning benchmarks [5].

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# 5. Implementation

## 5.1 Frontend

- Built with React (v18), functional components, and Material-UI for accessibility.
- Patient data entered via validated forms, progressive disclosures,
- Data visualization with recharts for lab values, genomics, etc.
- Secure authentication using JWT and OAuth2.

## 5.2 Backend API

- Node.js with Express; endpoints for user authentication, patient data handling, and AI querying.
- Data is stripped of identifiers before AI processing.
- Uses RESTful APIs, sockets for push notifications.

## 5.3 Gemini AI Integration

- Prompts are engineered to translate EHR data into natural questions ("Given this patient... what treatments are most effective?")
- AI outputs are post-processed for explainability (pro/con tables, citations).
- Option to use Gemini Pro for text and Gemini Pro Vision for imaging (radiology, pathology).

Here is a summary of a cancer patient:

- Diagnosis: ER+ breast cancer, Stage II
- Age: 55
- Biomarkers: ER+, PR+, HER2-
- Prior therapies: Surgery (lumpectomy)

Based on NCCN guidelines and current evidence, summarize recommended systemic therapies, expected outcomes, and clinical trial options. Include rationale.

## 6. Evaluation and Results

A pilot study tested the platform with synthetic patient cases and oncology residents. Metrics evaluated:

- Speed: End-to-end recommendations <10s for 95% queries.
- Accuracy: Agreement with board-certified oncologist recommendations in 84% cases.
- User Experience: High satisfaction with interface intuitiveness and AI explanations.

Limitations: Model can occasionally hallucinate citations; does not replace direct physician oversight.

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# 7. Challenges and Solutions

# 7.1 Data Privacy

Anonymization and GDPR/HIPAA-compliant handling enforced at all API layers. End-to-end encryption during data transmission.

# 7.2 Hallucinations & Clinical Validity

In-app warnings indicate AI suggestions must be reviewed by medical professionals. All recommendations cite links to primary guidelines/literature.

## 7.3 Scaling & Performance

Vite's fast build and code splitting, plus Gemini's API scaling, enable snappy, robust user experience even with multiple concurrent queries.

## 8. Future Work

- Integration with patient-facing mobile apps for symptom tracking.
- Real-world validation with de-identified, large-scale datasets.
- Finer-tuned, locally hosted models for enhanced privacy.
- Multi-language support; deeper clinical trial search integration.

## 9. Conclusion

AI-driven personalized oncology apps—when engineered with modern frameworks such as React.js, Vite, and Gemini AI—can bring cutting-edge clinical reasoning into usable, interactive platforms. This bridges a crucial gap between medical data, AI insight, and practical point-of-care decision support, empowering clinicians and patients alike.

## References

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