



# PATIENT FEEDBACK ANALYSIS AND REMINDER SYSTEM



HIGH5 - CODE2CARE

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## TRACK 1 -GENERALITY

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The healthcare system in Cameroon, and more specifically at the Douala General Hospital (DGH), faces major challenges related to patient feedback management and user communication. The lack of a centralized, multilingual platform to collect feedback and send automated reminders results in:

- A poor ability to measure patient satisfaction.
- A high rate of missed appointments and poor treatment adherence.
- Delays in identifying and resolving recurring problems (waiting times, quality of care, etc.).

In a context marked by linguistic diversity (**French, English, local languages**) and technological constraints (**low bandwidth, limited access to digital tools**), it is crucial to develop a suitable, inexpensive solution that can be easily integrated into existing systems.

This track aims to design and prototype an integrated system enabling:

- **Multilingual collection of patient feedback** via an accessible interface (text, voice, emojis).
- **Automated feedback analysis** (sentiment, recurring themes) for informed decision-making.
- **Sending personalized reminders** (SMS, voice call) for appointments and medications.
- **A real-time dashboard** for monitoring hospital performance.

The challenges of this project include:

- **Improving the quality of care:** Better listening to patients allows services to be adjusted according to their needs.
- **Resource optimization:** Reducing missed appointments and waste due to poor scheduling.
- **Inclusion and accessibility:** The solution must be usable by patients with low literacy skills or who do not speak official languages.
- **Sustainability:** The system must be easy to maintain and scalable to adapt to the hospital's future needs.

The actual constraints of this project:

- **Technological:**

- Compatibility with existing infrastructure (poor connectivity, legacy information systems).
- Need for a lightweight and functional offline solution if necessary.

**- Linguistic and cultural:**

- Support for French, English, and local languages (Douala, Bassa, Ewondo).
- Adaptation to varying literacy levels (use of icons, synthetic voice).

**- Regulatory:**

- Compliance with health data privacy and security standards.

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## *Challenge*

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How can we design a digital, multilingual, and accessible system capable of improving patient engagement and operational efficiency at Douala General Hospital, while taking into account local technological and socio-cultural constraints?

This track aims to address this challenge by developing a modular solution, combining:

- An intuitive user interface (web/mobile).
- Automated analysis tools (NLP, machine learning).
- Automated communication mechanisms (SMS, voice calls).
- A decision-making dashboard for managers.

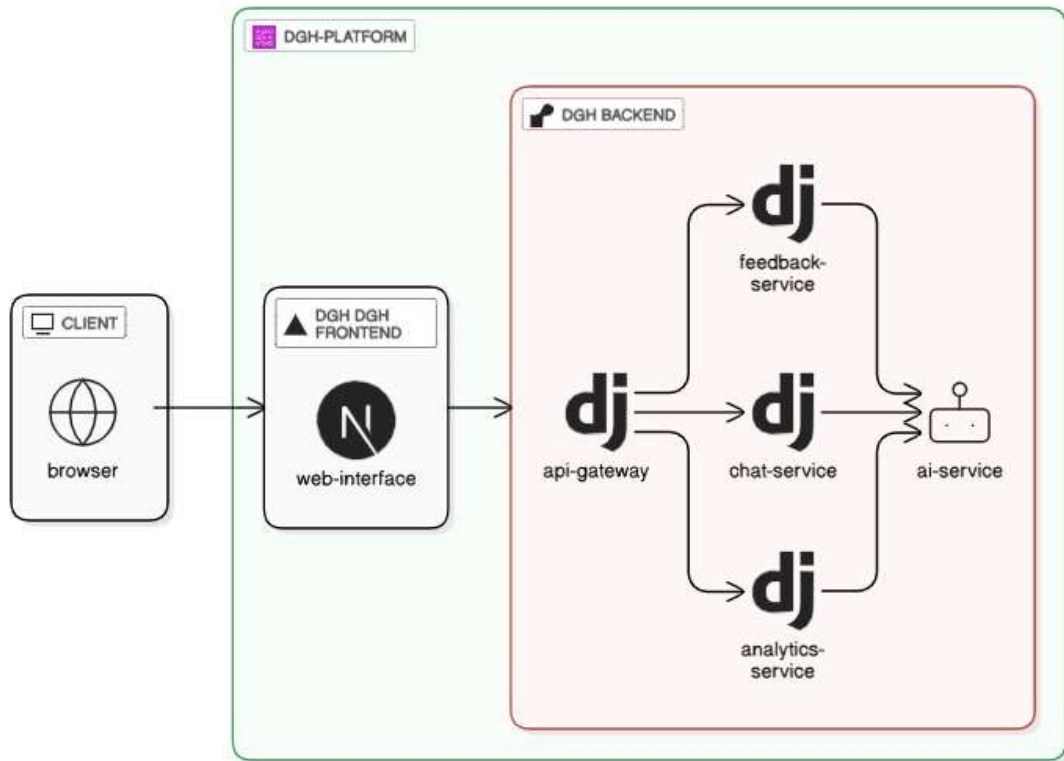
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## *CONCEPTION & ANALYSIS*

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### ARCHITECTURE

Our solution's architecture adopts a **modular, microservices approach**, perfectly suited to the needs of the Douala General Hospital (DGH).



Here is a detailed analysis of the components and their relevance:

### 1. Client (Browser)

Purpose: User interface accessible via web browser (computer, mobile).

#### - Advantage:

- No need to install a native application (reduces technical constraints for patients/staff).
- Compatible with low-performance devices.

### 2. DGH Frontend (Web Interface)

Purpose: Dynamic user interface (React.js, Flutter Web?) for:

- Capturing feedback (text/voice/emojis).
- Viewing the administrative dashboard.
- Interacting with the chatbot (if integrated).

- **Advantage:**
  - Unified across all modules (feedback, reminders, chatbot).

### 3. API Gateway

Roles:

- Routing requests to microservices.
- Authentication (JWT + OAuth2 for medical staff).
- Rate limiting (preventing abuse).
- Critical endpoints:
  - /feedback → Feedback Service
  - /predict → AI Service

### 4. DGH Backend (Microservices)

Structure: Multiple independent services communicating via an API Gateway:

- **Feedback Service:** Manages the collection and storage of patient feedback.
- **AI Service:** NLP analysis of feedback (sentiment, themes).
- **Analytics Service:** Generates data for the dashboard.
- Advantages:
  - Scalability: Each service can be updated independently.
  - Resilience: An outage on one service does not affect the others.

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### *Architecture Highlights*

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- **Modularity:** Services are decoupled, facilitating maintenance and adding features (e.g., adding an SMS reminder module).
- **Seamless Integration:** The API Gateway centralizes requests and simplifies access management (authentication, logs).
- **Adaptability to Local Constraints:** Can be deployed on a local cloud (e.g., Google Cloud with servers in Cameroon) to reduce latency.

## CHOICE OF TECHNOLOGIES

### - Backend

<b>Criteria</b>	<b>Django</b>	<b>Fast API</b>	<b>Flask</b>	<b>Spring Boot</b>	<b>Laravel</b>
<b>Language Type</b>	Python Full-stack	Python API-first, async	Python Microframework	Java Full-stack Java EE	PHP Full-stack
<b>ORM integer</b>	Django ORM	(SQLAlchemy optional)	(optional)	Hibernate	Eloquent
<b>Auto admin UI</b>	Puissant (admin site)	non	non	non	Basque
<b>Security indegree</b>	CSRF, XSS, JWT, RBAC	Basique	non	Advance (Spring Security)	Bonne (mais PHP dependent)
<b>Performance API</b>	Bonne	Très rapide (async)	Moyenne	Très bonne	Moyenne
<b>Courbe d'apprentissage</b>	Moyenne	Faible à moyenne	Faible	Complexe (Java, configs)	Faible
<b>Écosystème santé / NLP</b>	Compatible avec NLP Python, Celery	oui	oui	(Java moins NLP-ready)	(moins adapté à NLP)
<b>Microservices Friendly</b>	(via Django REST + Celery)	Oui nativement	Oui mais nécessite extensions	Excellent (native)	Moins adapté
<b>Communauté Afrique/Global Sud</b>	Large adoption	En croissance	Stable	Moins courant	Large en Afrique (mais PHP-centric)

## - Front-end

<b>Criteria</b>	<b>Next.js [Recommandé]</b>	<b>React.js (pur)</b>	<b>Vue.js</b>
<b>Type</b>	Framework complet basé sur React	Librairie UI	Framework complet
<b>Rendu Serveur (SSR)</b>	Intégré nativement (SEO, rapidité)	Besoin de setup (Next, Remix...)	Possible via Nuxt.js
<b>Rendu statique (SSG)</b>	Oui, sans config	Non natif	Avec Nuxt.js
<b>Routing automatique</b>	Basé sur le système de fichiers	Nécessite React Router	Avec Vue Router
<b>SEO / Accessibilité</b>	Excellente pour hôpitaux	Faible (CSR only)	Moyenne (Nuxt recommandé)
<b>Intégration API Django (REST)</b>	Parfait via fetch / SSR API	OK	OK
<b>Multilingue &amp; i18n</b>	next-i18next, facile à intégrer	Moins structuré	Bon via vue-i18n
<b>Performance (vieux téléphones)</b>	SSR/SSG très rapide	CSR = + lent au chargement	Bonne mais dépend des bundles
<b>Déploiement sur Vercel</b>	Automatique, natif	Besoin de setup	Possible mais pas optimisé
<b>Code splitting automatique</b>	Oui, sans config	Non	Oui
<b>Auth &amp; sécurité (RBAC)</b>	Middleware intégrable	Besoin de setup JWT	Possible
<b>Courbe d'apprentissage</b>	Moyenne (React + conventions)	Moyen	Facile pour débutants
<b>Communauté Afrique / Django / Python</b>	Forte adoption	Présente	Moins courante dans écosystèmes IA

## - Sentiment Analysis

Criteria	RoBERTa Base	BERT	LSTM	Logistic Regression
Performance	9.5/10	9/10	7/10	6/10
	- SOTA on many NLP tasks- Excellent on complex classification- Handles context and nuances	- Exceptional performance- NLP benchmark reference- Good on diverse tasks	- Good performance on sequences- Effective on temporal data- Limited by architecture	- Decent performance- Good for linear data- Reliable baseline
Speed	4/10	3/10	7/10	10/10
	- Training: hours/days- Parallelization possible	- Training: hours/days- Resource intensive	- Training: minutes/hours- Sequential but fast	- Training: seconds/minutes- Very lightweight
Precision	9.5/10	9/10	7.5/10	6.5/10
Adaptability	8/10	8/10	6/10	4/10
	- Efficient fine-tuning- Excellent transfer learning- Multi-domain adaptation	- Standard fine-tuning- Versatile- Fast adaptation	- Architectural adaptation- Hyperparameter tuning- Limited to sequences	- Limited adaptation- New features required- Strict assumptions
Deployment Ease	3/10	2/10	7/10	10/10
	- Resources: GPU (8-16GB)- Complex frameworks- Optimization required	- Resources: GPU (8-16GB)- Heavy infrastructure- Maximum complexity	- Resources: Light CPU/GPU- Standard frameworks- Moderate deployment	- Resources: Minimal CPU- Simple libraries- Immediate deployment



## - Theme Extraction

Méthode	Qualité thématique	Besoin de training	Multilingue	Déploiement facile	Temps de calcul	Avantages principaux	Inconvénients
<b>LLM (LLaMA, GPT, etc.)</b>	Très élevée (compréhension profonde)	Aucun (in-context)	Oui (selon modèle)	Moyen à difficile (nécessite API ou Ollama)	Moyen à lent	Compréhension du contexte, personnalisation aisée via prompt	Coût mémoire/temps, difficile à auditer
<b>BERTopic</b>	Élevée (embedding + clustering)	Oui (sur corpus)	Oui	Moyen (besoin d'UMAP/HDBSCAN)	Moyen	Bon équilibre entre qualité thématique et interprétabilité	Moins performant sur petits jeux de données
<b>LDA</b>	Moyenne	Oui	Non (majoritairement anglophone)	Facile	Rapide	Interprétation facile, classique	Moins précis, sensible au prétraitement
<b>NMF</b>	Moyenne à bonne	Oui	Non	Facile	Rapide	Moins bruité que LDA, mieux sur petits corpus	Thèmes parfois moins cohérents
<b>Top2Vec</b>	Bonne	Non (embeddings)	Oui	Moyen (poids modèles)	Moyen	Aucun prétraitement, détecte automatiquement les sujets	Pas toujours stable sur petits corpus
<b>KeyBERT</b>	Bonne (keywords → thèmes)	Non (zero-shot)	Oui	Très facile	Très rapide	Très simple, efficace sur petits textes	Pas de regroupement automatique (juste mots-clés)

Nb: the green cell represents our choice

## TOOLS

### PATIENT FEEDBACK AND REMINDER SYSTEM TOOLS

