

☐ README

README.md

Predicting Family Planning Demand and Optimizing Service Delivery in Kenya

Create README.md

This project aims to leverage data science and machine learning to predict family planning (FP) demand and optimize service delivery in Kenya. By forecasting demand for specific FP methods and predicting continuation rates, this initiative seeks to enable proactive resource allocation, reduce stock-outs, and improve overall FP outcomes. This version of the project focuses specifically on a **regression-based approach**.

2 weeks ago

1. Project Structure

The project is organized into the following directories and files:

- ./
 - README.md: This file.
 - FP_Demand_Optimization.ipynb: The main Jupyter Notebook for data analysis, modeling, and evaluation.
 - requirements.txt: Lists all necessary Python libraries and their versions.
 - src/: (Optional) Directory for reusable Python scripts (e.g., data cleaning functions, custom models).
 - data_preparation.py
 - model_training.py
 - data/: Contains raw and processed datasets.
 - ke_fp_population_data.csv : County-level demographic data.
 - ke_fp_service_data.csv : Provides historical family planning service delivery statistics by county and method.
 - external_data/ : (Optional) Directory for external data sources after acquisition.
 - models/: (Optional) Directory to save trained machine learning models.
 - notebooks/: (Optional) Additional notebooks for exploration or specific analyses.
 - o reports/: (Optional) Contains generated reports, visualizations, or presentations.

2. Setup and Installation

To set up the project environment, follow these steps:

- 1. Clone the repository:
- 2. Create a virtual environment (recommended):
- 3. Install required libraries:

3. Data

3.1 Data Sources

The project utilizes two primary datasets, located in the data/ directory:

- ke fp population data.csv : Contains county-level demographic information over time.
- ke_fp_service_data.csv : Provides historical family planning service delivery statistics by county and method.

External data sources such as DHS, Census, Guttmacher Institute data, FP Track, PMA, UN SARA, and USAID statistics will be integrated to enrich the dataset.

3.2 Data Preparation Overview

The data preparation phase involves:

- Data Cleaning: Handling missing values, addressing outliers, and standardizing data.
- **Feature Engineering**: Creating new features like *lagged time-series values*, *demographic ratios*, and *supply-side indicators*.
- Target Variable Creation: Defining the specific outcomes for prediction, focusing on regression targets:
- For Regression: new_clients_method_X (e.g., injectables, pills) and continuation_rate_method_Y.
- Data Transformation: Scaling numerical features and encoding categorical variables.

4. Methodology: Machine Learning Approach (Regression)

This project focuses on a regression-based machine learning approach:

4.1 Regression for Predicting Future Demand and Continuation Rates

- Objective: To forecast continuous numerical values
- Models Used: Linear Regression, Random Forest Regressor, Gradient Boosting Regressor.
- Evaluation: Performance will be assessed using metrics like *Mean Absolute Error (MAE)*, *Root Mean Squared Error (RMSE)*, and *R-squared (R2)*.

5. Results and Actionable Insights

The project aims to deliver:

- Model Performance: Quantified accuracy and reliability of predictions.
- Feature Importance: Identification of key drivers influencing FP demand and continuation rates.
- Impact Quantification: Calculation of Couple Years of Protection (CYP) and averted pregnancies based on predicted service delivery.
- Resource Requirements: Estimates for commodity procurement, equipment needs, and staffing allocations.
- Policy & Programmatic Recommendations: Data-driven insights for policymakers and program managers to optimize interventions.

6. Usage

To run the analysis and generate predictions:

- 1. Ensure all setup steps are completed.
- 2. Open the FP_Demand_Optimization.ipynb notebook in a Jupyter environment.
- 3. Execute the cells sequentially to load data, preprocess, train models, and evaluate results.

For more detailed information on specific data points, feature engineering, or model parameters, please refer to the comments within the FP_Demand_Optimization.ipynb notebook.

Releases

No releases published Create a new release

Packages

No packages published Publish your first package

Contributors 3



FranciscarNdanu Franciscar_Ndanu



EdgarSoudie Soudie Okwaro



RGitau-coder RGitau-RG

Languages

Jupyter Notebook 100.0%