

## PLP Academy Community Article: Understanding the AI Development Workflow

**Course:** AI for Software Engineering\ **Assignment Title:** Understanding the AI Development Workflow\

**Duration:** 7 Days\ **Author:** Grateful Juma

---

### Overview

This article explores the end-to-end process of applying the AI Development Workflow to a real-world case: predicting hospital patient readmission within 30 days. It includes theoretical insights, technical choices, ethical considerations, and deployment strategies, aligned with the CRISP-DM framework.

---

### Case Study: Predicting Patient Readmission

**Problem:** Build an AI model that predicts whether a discharged patient will be readmitted within 30 days.

#### Objectives:

- Lower readmission rates.
- Assist clinicians with early warnings.
- Improve healthcare outcomes.

#### Stakeholders:

- Hospital administrators
- Healthcare providers

#### Data Strategy:

- **Sources:** EHRs, demographics
- **Ethics:** Privacy & consent
- **Preprocessing:** Imputation, encoding, feature engineering

#### Modeling:

- **Choice:** Gradient Boosting (XGBoost)
- **Metrics:** Precision (0.78), Recall (0.70)
- **Confusion Matrix:** [TP=70, FN=30, FP=20, TN=80]

#### Deployment:

- Packaged API (Flask)
- Integrated into EHR system
- Encryption + access controls

### Optimization:

- Regularization + early stopping to mitigate overfitting
- 

## Short Answer Insights

### 1. Problem Definition:

- **Example:** Predict employee attrition
- **KPI:** Attrition rate post-model deployment

### 2. Data Preprocessing:

- Normalize, impute, encode
- Watch for survey bias

### 3. Model Development:

- Use Random Forest for structured HR data
- Hyperparameters: `max_depth`, `n_estimators`

### 4. Evaluation:

- AUC-ROC & F1-score
  - Handle concept drift with retraining triggers
- 



## Critical Thinking

### Ethics:

- Biased data can worsen patient outcomes
- Mitigation: Fairness-aware algorithms (e.g., reweighing)

### Trade-offs:

- High-accuracy black-box vs. interpretable models
  - Limited hardware favors lightweight models
- 



## Reflection & Diagram

**Challenge:** Aligning technical AI solutions with clinical reality

**Improvement:** Add explainability (e.g., SHAP) and stakeholder interviews

### Workflow Diagram:

1. Define Problem
  2. Collect & Clean Data
  3. Develop Model
  4. Evaluate
  5. Deploy + Monitor
- 

### GitHub Repo





#### Structure:

```
ai-readmission-predictor/  
├── src/  
│   ├── preprocessing.py  
│   ├── model_training.py  
│   └── deploy_api.py  
├── data/  
│   └── sample_ehr.csv  
├── notebooks/  
│   └── exploratory_analysis.ipynb  
├── README.md  
├── requirements.txt  
└── LICENSE
```

#### Includes:

- Well-commented code
  - Requirements file
  - Instructions to run & deploy model
- 

### Final Deliverables

-  PDF Report
-  GitHub Repo
-  This Article (PLP Submission)
-  Slide Deck (available on request)

**Let's build AI that works for people—not just the system.**