

# **Diabetes Progression Prediction Service**

ML service for predicting short-term diabetes progression risk to help triage nurses prioritize patient follow-ups.

Show Image



#### **Overview**

**Context:** A hospital runs a virtual diabetes clinic where nurses review hundreds of patient check-ins weekly to decide who needs follow-up calls. Manual reviews are time-consuming.

**Solution:** This ML service predicts disease progression and returns a continuous risk score, enabling nurses to prioritize high-risk patients first.

**Dataset:** Uses scikit-learn's diabetes dataset as a stand-in for de-identified EHR features (age, BMI, blood pressure, cholesterol, blood sugar, etc.).



### **Quick Start**

#### **Using Docker (Recommended)**

```
# Pull pre-built image
docker pull ghcr.io/YOUR_USERNAME/YOUR_REPO/diabetes-predictor:latest

# Run service
docker run -p 8000:8000 ghcr.io/YOUR_USERNAME/YOUR_REPO/diabetes-predictor:latest

# Test prediction
curl -X POST http://localhost:8000/predict/single \
-H "Content-Type: application/json" \
-d '{
    "age": 0.05, "sex": 0.05, "bmi": 0.06, "bp": 0.02,
    "s1": -0.04, "s2": -0.03, "s3": -0.04, "s4": 0.0,
    "s5": 0.02, "s6": -0.03
}'
```

### **Local Development**

```
bash

# Install dependencies

pip install -r requirements.txt

# Train model

python train.py --version v0.1 --model-type linear

# Run API

uvicorn app:app --reload --port 8000

# Open API docs

open http://localhost:8000/docs
```

# **API** Endpoints

### **Health Check**

bash
GET /health

#### **Response:**

```
json
{
    "status": "healthy",
    "model": "LinearRegression",
    "scaler": "StandardScaler"
}
```

### **Single Patient Prediction**

bash

POST /predict/single

#### **Request Body:**

json

```
{
    "age": 0.05,
    "sex": 0.05,
    "bmi": 0.06,
    "bp": 0.02,
    "s1": -0.04,
    "s2": -0.03,
    "s3": -0.04,
    "s4": -0.00,
    "s5": 0.02,
    "s6": -0.03
}
```

### **Response:**

```
json
{
    "patient_id": 0,
    "progression_score": 152.5,
    "risk_level": "HIGH",
    "high_risk": true
}
```

## **Batch Prediction (Triage Dashboard)**

```
bash
POST /predict
```

### **Request Body:**

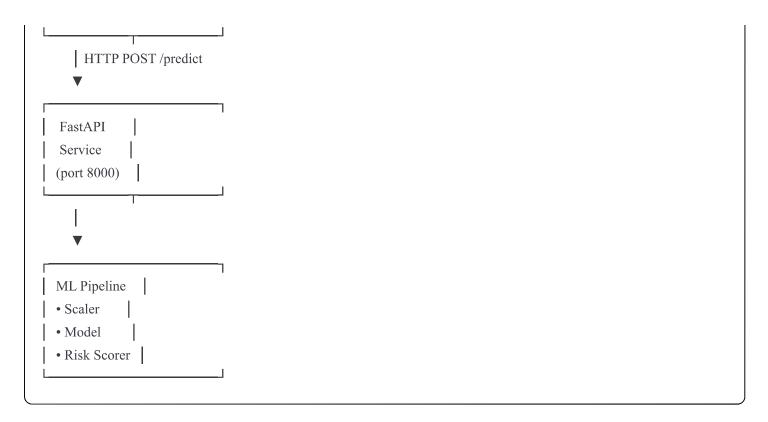
```
json
```

```
{
    "patients": [
    {
        "age": 0.05, "sex": 0.05, "bmi": 0.06, "bp": 0.02,
        "s1": -0.04, "s2": -0.03, "s3": -0.04, "s4": 0.0,
        "s5": 0.02, "s6": -0.03
    },
    {
        "age": -0.01, "sex": -0.04, "bmi": -0.03, "bp": 0.0,
        "s1": 0.01, "s2": 0.02, "s3": 0.03, "s4": 0.02,
        "s5": 0.0, "s6": 0.01
    }
}
```

#### **Response:** (sorted by risk, descending)

## **Architecture**

```
Triage Nurse |
Dashboard
```



## CI/CD Pipeline

GitHub Actions automatically:

- 1. Trains model on every push to main
- 2. **Runs unit tests**
- 3. Builds Docker image
- 4. **V** Pushes to GitHub Container Registry
- 5. Runs smoke tests
- 6. Reports metrics in Actions summary

#### **Manual Model Training:**

```
# Trigger workflow with custom parameters
gh workflow run train-and-deploy.yml \
-f version=v0.2 \
-f model_type=ridge
```

#### v0.1 (Baseline - Linear Regression)

Metric	Value
RMSE	~55-60
MAE	~45-50
R <sup>2</sup>	~0.45-0.52
<b>▲</b>	· · · · · · · · · · · · · · · · · · ·

#### Risk Classification (75th percentile threshold):

Metric	Value
Precision	~0.70-0.80
Recall	~0.65-0.75
F1 Score	~0.68-0.77
◀	•

See <u>CHANGELOG.md</u> for detailed metrics and version history.

## **Training New Models**

### Baseline (v0.1)

bash

python train.py --version v0.1 --model-type linear

### Ridge Regression (v0.2)

bash

python train.py --version v0.2 --model-type ridge --alpha 1.0

### Random Forest (v0.2)

bash

python train.py --version v0.2 --model-type random\_forest \
--n-estimators 100 --max-depth 10

#### **Training Outputs:**

- (models/model\_vX.X.pkl) Trained model + scaler
- (models/metrics vX.X.json) Performance metrics

## Testing

```
# Install test dependencies
pip install pytest httpx

# Run tests
pytest test_api.py -v

# Test with Docker
docker build -t diabetes-predictor:test .
docker run -d -p 8000:8000 --name test diabetes-predictor:test
curl http://localhost:8000/health
docker stop test && docker rm test
```

## Project Structure

```
- .github/
 — workflows/
 train-and-deploy.yml # CI/CD pipeline
- models/
                      # Trained models (gitignored)
  model v0.1.pkl
  metrics v0.1.json
                     # FastAPI service
app.py
train.py
                     # Training pipeline
- requirements.txt
                         # Python dependencies
- Dockerfile
                       # Container definition
- CHANGELOG.md
                              # Version history & metrics
                           # This file
- README.md
```

## **Security & Privacy**

- No PHI/PII: Demo uses synthetic scikit-learn dataset
- **In Production**: Would require:
  - HIPAA-compliant infrastructure
  - Data encryption at rest and in transit

- Audit logging
- Authentication/authorization
- De-identification of training data

### **Roadmap**

#### **v0.2** (Next)

- Ridge/Random Forest models
- ☐ Feature selection/engineering
- Improved risk calibration
- A/B testing framework

#### **Future**

- Model monitoring & drift detection
- Explainability (SHAP values)
- Real-time retraining
- Multi-model ensemble
- Dashboard UI



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### Contributing

- 1. Fork the repository
- 2. Create feature branch ((git checkout -b feature/amazing-feature))
- 3. Train and test your model
- 4. Commit changes ((git commit -m 'Add amazing feature'))
- 5. Push to branch (git push origin feature/amazing-feature)
- 6. Open Pull Request (CI will auto-test)

#### Contact

MLOps Team - your-email@example.com

Project Link: <a href="https://github.com/YOUR\_USERNAME/YOUR\_REPO">https://github.com/YOUR\_USERNAME/YOUR\_REPO</a>