

# PORTFOLIO

## DATA SCIENCE

### NASSER CHAOUCHI



# SUMMARY

**1.WHO AM I?**

**2.MY WORK EXPERIENCE**

**3.MY PROJECTS**

a.THE MOVIE RECOMMENDER SYSTEM

b.MULTICLASS CLASSIFICATION FOR DIABETES

c.THE CKD AND DYALYSIS PREDICTION

# SUMMARY

## My academic journey

I'm Nasser, a **French computer science engineer** passionate about **artificial intelligence, data, and innovation**. I **graduated from UTC** (Université de Technologie de Compiègne) in 2025 with a **major in AI and Data Science**.

I completed:

- A **dual-focused internship** at **Numberly** as a **Data Engineer and Project Manager**, combining technical and management responsibilities.
- An **exchange semester** at the **Escuela de Ingeniería y Arquitectura in Zaragoza** (Spain), as part of a **Data Science Master's program**.
- A **final-year internship** at **Ubisoft** as a **Data Scientist**, working on **real-world game data** and **predictive models**.



## Strengths

Curious

Rigorous

Positive

Patient

## Interests

Artificial  
Intelligence

Sports

Literature

Chess

# MY WORK EXPERIENCE

An **internship** at **Ubisoft** from **October 2024** to **March 2025** as a **Data Scientist**, with the following missions:

**Audiences Understanding**

**Segmentation Based on  
Players' Profiles**

**Player Behavior Prediction**

I worked on the game **Avatar: Frontiers of Pandora**. My role was **to understand the game's underperformance** and to **identify and target potential players** within the **Ubisoft ecosystem** who would most likely acquire the game.

The project was divided into **three main phases**:

**Ad-hoc Analyses**

**Clustering Development**

**Classifier with Prediction**

You can contact the team manager for a reference:

- Nicolas Tatin, Associate Director, Data & Analytics

Data Analysis

Data Science

# MY PROJECTS – THE MOVIE RECOMMENDER SYSTEM

## CONTEXT

- **Dataset:** [MovieLens 32M](#)
- **Goal:** Recommend movies users might like, based on behavior and content
- **Type:** Hybrid Recommendation System
  - Collaborative Filtering (ratings)
  - Content-Based Filtering (genres, titles)
- **Size:** 32M+ ratings, ~270k users, 62k movies

## BUILT WITH

- Scikit-learn
- Pandas, NumPy
- Seaborn
- Matplotlib
- HuggingFace datasets
- Streamlit

[Open the repository](#)

[Open the interface](#)  
(with [MovieLens 1M](#))

## APPROACH

- **Data Cleaning:** Merged movies.csv and ratings.csv, extracted year, processed genres
- **Collaborative Filtering:** Built user-item matrix, applied cosine similarity
- **Content-Based Filtering:** Used TF-IDF/CountVectorizer on genres and titles
- **Hybrid Strategy:** Combined top recommendations from both approaches
- **Implemented multiple strategies:** Most rated movies, Top-rated by genre, Top-rated by year, User-user collaborative hybrid, Item-item collaborative hybrid
- **Profile-Based Recommendation:** Built a user profile from favorite movies to generate personalized suggestions

## WHAT I LEARNED

- **Designing** and **comparing recommender strategies**
- Using **similarity metrics** (cosine) on **sparse data**
- **Evaluating trade-offs** between **relevance** and **diversity**

# MY PROJECTS – THE MOVIE RECOMMENDER SYSTEM

### Recommend Similar Movies

Start typing a movie title:

Select a movie:

Toy Story (1995)

Balance (ratings vs. genres) 0.00 0.50 1.00

Recommend

#### Recommendations:

- Toy Story 2  
1999  
Animation|Children's|Comedy
- Bug's Life, A  
1998  
Animation|Children's|Comedy
- Aladdin  
1992  
Animation|Children's|Comedy|Musical
- Chicken Run  
2000  
Animation|Children's|Comedy
- American Tail, An  
1986  
Animation|Children's|Comedy

### ★ Popular Picks

Choose a recommendation method:

☒ Top Rated Movies

☐ By Genre

☐ By Year

☐ New User

Show top movies

#### Recommendations:

- Usual Suspects, The  
1995  
Crime|Thriller
- Shawshank Redemption, The  
1994  
Drama

### ★ Popular Picks

Choose a recommendation method:

☐ Top Rated Movies

☐ By Genre

☒ By Year

☐ New User

Select a year:

1983

Recommend by Year

#### Recommendations:

- Star Wars: Episode VI - Return of the Jedi  
1983  
Action|Adventure|Romance|Sci-Fi|War
- Right Stuff, The  
1983  
Drama

### ★ Popular Picks

Choose a recommendation method:

☐ Top Rated Movies

☒ By Genre

☐ By Year

☐ New User

Select a genre:

Action

Recommend by Genre

#### Recommendations:

- Star Wars: Episode IV - A New Hope  
1977  
Action|Adventure|Fantasy|Sci-Fi

### ★ Popular Picks

Choose a recommendation method:

☐ Top Rated Movies

☐ By Genre

☐ By Year

☒ New User

Choose your favorite genres:

Animation × Crime × Documentary ×

Recommend for New User

#### Recommendations:

- Usual Suspects, The  
1995  
Crime|Thriller

### MovieLens Hybrid Recommendation System

★ Popular Picks By Movie Item-Item Hybrid By User Item-Item Hybrid By Manual Selection User

#### Personalized Recommendations

Select a user ID:

6

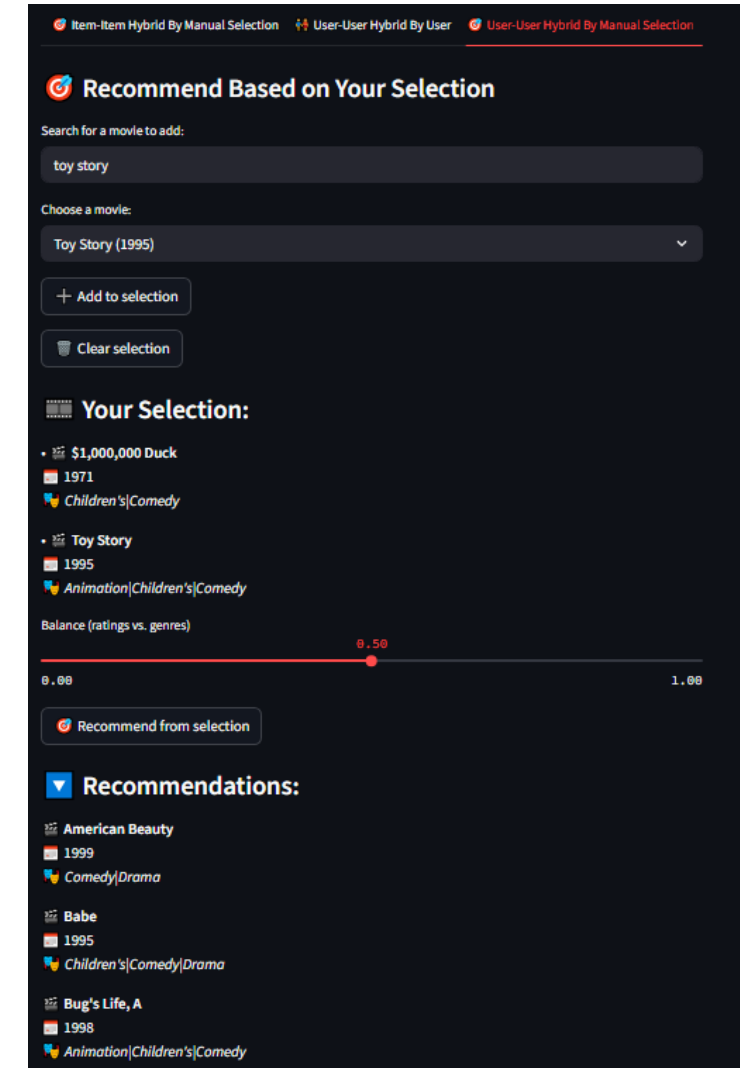
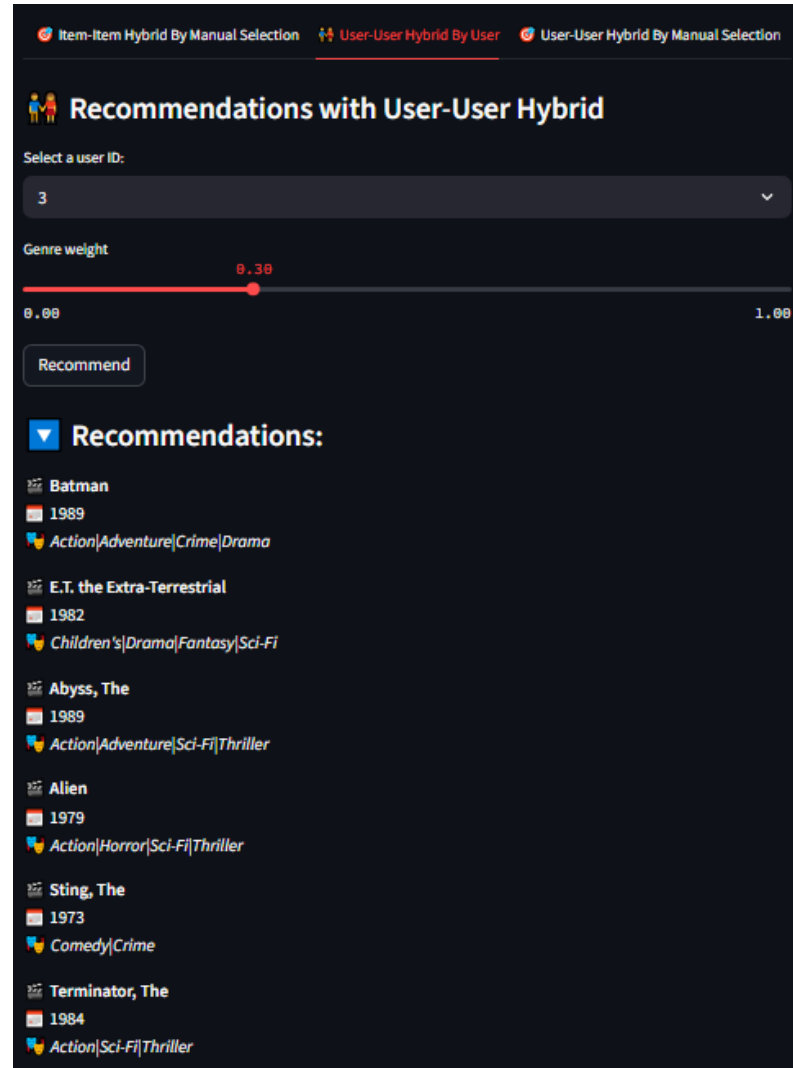
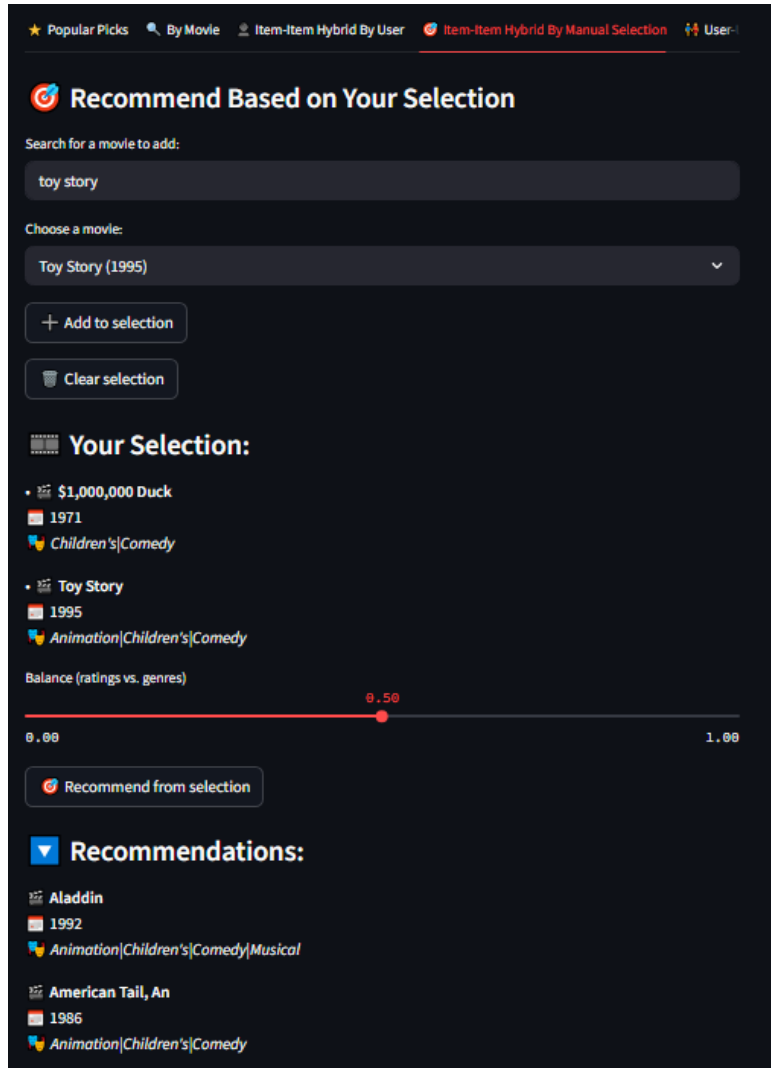
Balance (ratings vs. genres) 0.00 0.50 1.00

Recommend

#### Recommendations:

- Hunchback of Notre Dame, The  
1996  
Animation|Children's|Musical
- Jungle Book, The  
1967  
Animation|Children's|Comedy|Musical
- Sleepless in Seattle  
1993  
Comedy|Romance
- Cinderella  
1950  
Animation|Children's|Musical
- When Harry Met Sally...  
1989  
Comedy|Romance

# MY PROJECTS – THE MOVIE RECOMMENDER SYSTEM



# MY PROJECTS – MULTICLASS CLASSIFICATION FOR

## CONTEXT

- **Dataset:** [Multiclass Diabetes Dataset](#)
- **Goal:** Classify patients into several diabetes stages
- **Type:** Supervised, Multiclass classification
- **Size:** 264 patients, 12 features

## BUILT WITH

- Scikit-learn
- Pandas, NumPy
- Seaborn
- Matplotlib
- Streamlit

[Open the repository](#)

[Open the interface](#)

## APPROACH

- **EDA & Preprocessing:** Analyzed feature distributions, handled missing values, balanced classes, and scaled data.
- **Model tested:** Logistic Regression, Random Forest and K-Nearest Neighbour
- **Cross Validation:** Ensured robust performance and avoided overfitting
- **Evaluation:** Confusion Matrix, Classification report (F1-Score, Accuracy, Recall)
- **Final model (Random Forest)**
  - **Accuracy:** 97%
  - **Macro F1-score** (better suited to class imbalance): 0.98

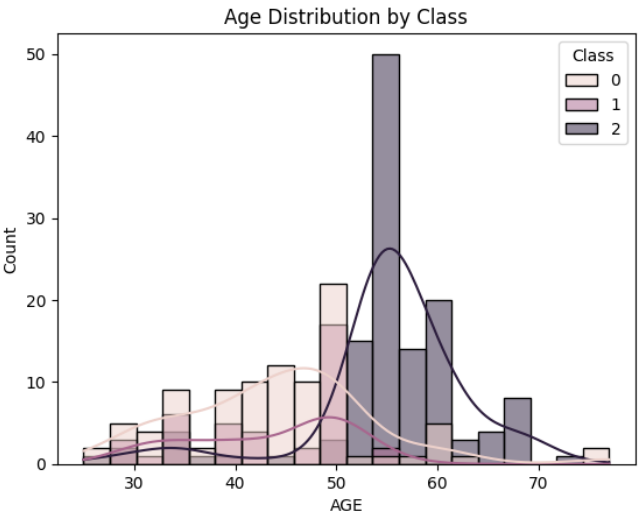
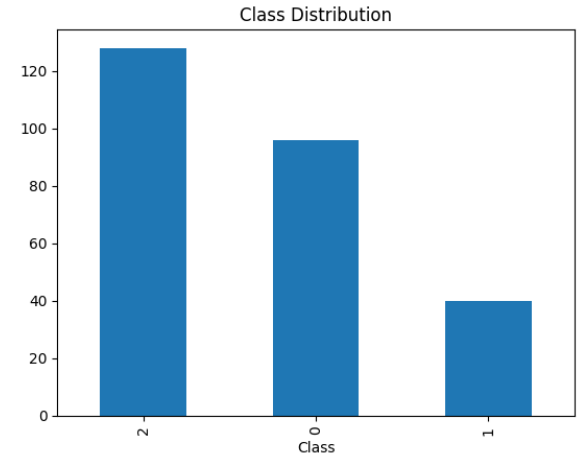
## WHAT I LEARNED

- **How to handle imbalanced multiclass data**
- The **importance** of **feature engineering** and **model tuning**
- Model explainability with **SHAP** or **feature importance**



# MY PROJECTS – MULTICLASS CLASSIFICATION FOR

Variable	F-statistic	p-value
AGE	60.368	0.00000
BMI	190.565	0.00000
TG	14.218	0.00000
HbA1c	200.415	0.00000
Chol	9.881	0.00007
Urea	9.115	0.00015
Cr	6.466	0.00182
VLDL	3.712	0.02573
LDL	0.957	0.38526
HDL	0.488	0.61457



### Diabetes Class Prediction

Enter patient data below to predict diabetes classification.

Gender  
☒ Female  
☐ Male

Age  
50

BMI  
25.00

HbA1c (%)  
6.00

Cholesterol  
5.00

Urea  
5.00

Creatinine  
1.00

Triglycerides  
2.00

HDL  
1.00

LDL  
2.00

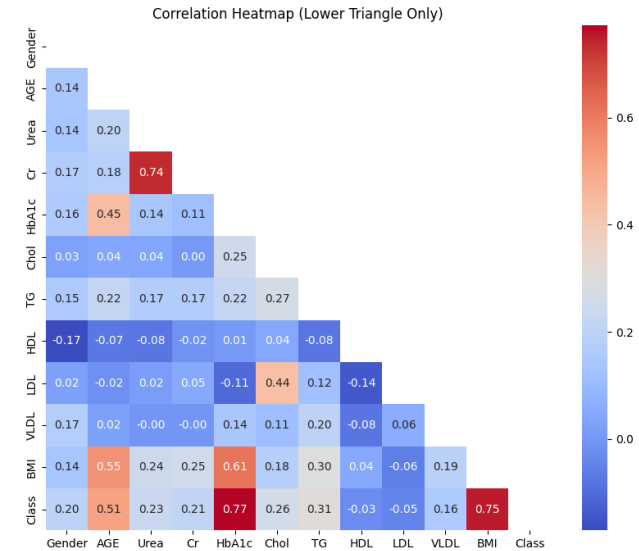
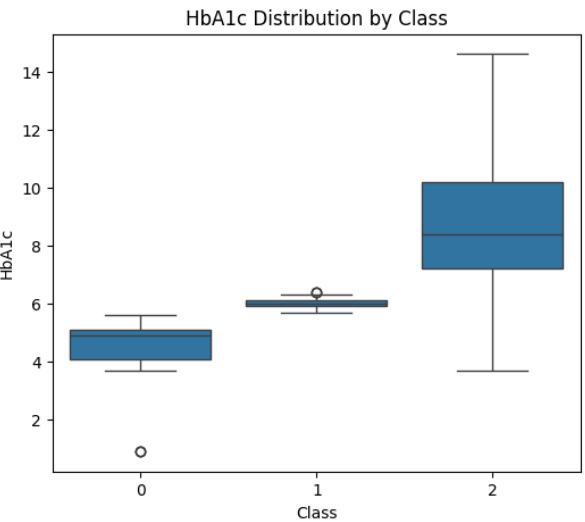
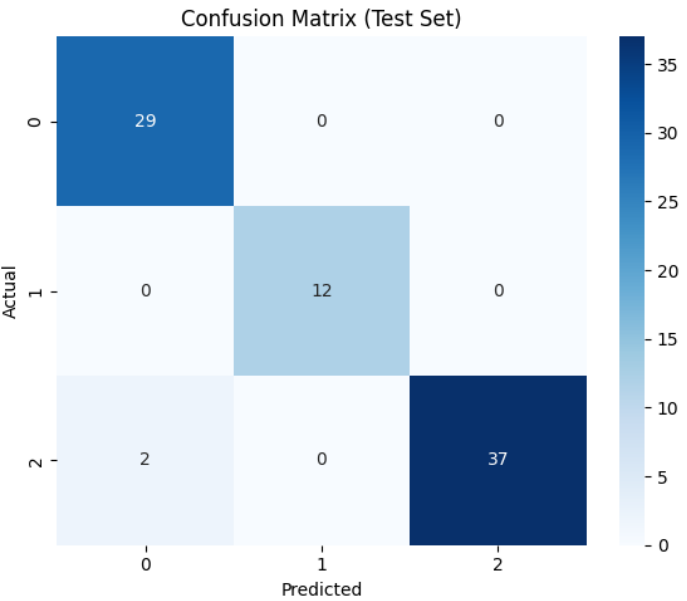
VLDL  
5.00

**Predicted Class: Diabetic**

⚠️ High likelihood of diabetes. Medical follow up recommended.

**Class Probabilities**

	Class	Probability
0	Non-Diabetic	11.00%
1	Diabetic	73.00%
2	Pre-Diabetic	16.00%



# MY PROJECTS – THE CKD AND DYALYSIS PREDICTION

## CONTEXT

- **Dataset:** [Kidney Disease Risk Dataset](#)
- **Goal:** Predict CKD status and dialysis need based on clinical and biological data
- **Type:** Supervised, Binary classification (2 targets: CKD\_Status, Dialysis\_Needed)
- **Size:** 2304 patients, 9 features

## BUILT WITH

- Scikit-learn
- Pandas, NumPy
- Seaborn
- Matplotlib
- XGBoost
- Streamlit

[Open the repository](#)

[Open the interface](#)

## APPROACH

- **EDA & Preprocessing:** Explored feature relationships, handled missing values, encoded categorical data, scaled numerical features.
- **Model tested:** Logistic Regression, Random Forest, Gradient Boosting, XGBoost and K-Nearest Neighbour
- **Cross Validation:** Ensured robustness and reduced overfitting risk.
- **Evaluation Classification Report, ROC-AUC, F1-Score, Accuracy**
- Best model was Gradient Boosting, but due to class imbalance, Random Forest gave more reliable results for generalization.
  - **Accuracy:** 100% for CKD\_Status — Accuracy 100% but a F1-Score 0.97 (class imbalance) for the Dialysis\_Needed
  - **Separate models trained for each target**

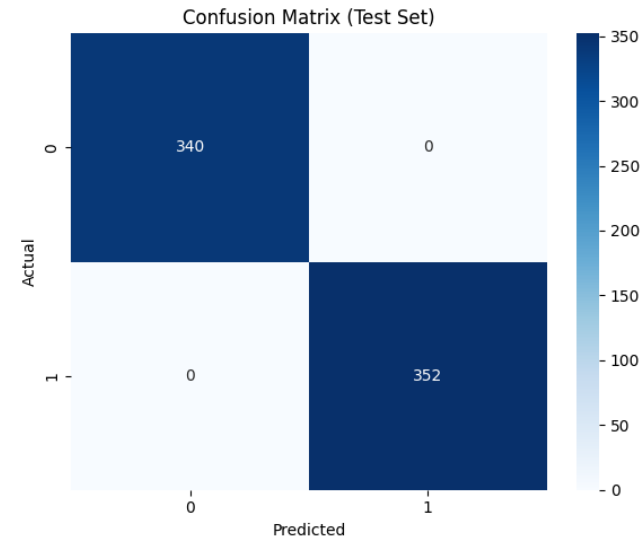
## WHAT I LEARNED

- **Managing dual target classification**
- **Handling noisy and medical data**
- Improving interpretability with **SHAP values**

# MY PROJECTS – THE CKD AND DYALYSIS PREDICTION

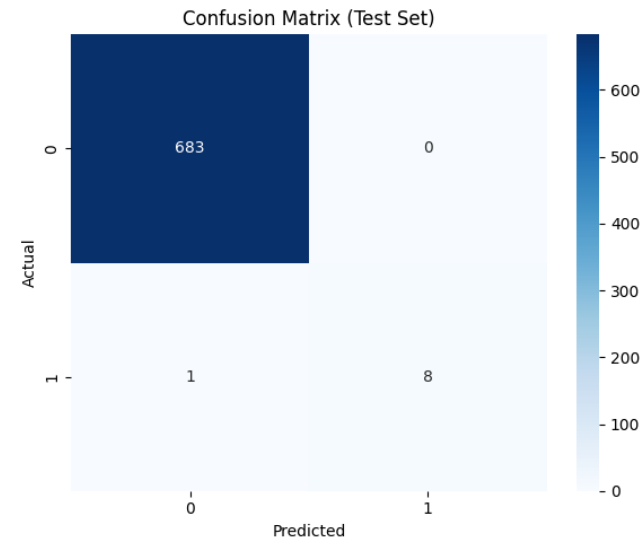
## Evaluation on Test Set – CKD\_Status


Class	Precision	Recall	F1-Score	Support
0	1.00	1.00	1.00	340
1	1.00	1.00	1.00	352
<b>Accuracy</b>			<b>1.00</b>	692
<b>Macro avg</b>	1.00	1.00	1.00	692
<b>Weighted avg</b>	1.00	1.00	1.00	692




## Evaluation on Test Set – Dialysis\_Needed

Class	Precision	Recall	F1-Score	Support
0	1.00	1.00	1.00	683
1	1.00	0.89	0.94	9
<b>Accuracy</b>			<b>1.00</b>	692
<b>Macro avg</b>	1.00	0.94	0.97	692
<b>Weighted avg</b>	1.00	1.00	1.00	692



 **Kidney Health Risk Prediction**

Predict the risk of Chronic Kidney Disease (CKD) and the potential need for dialysis based on patient clinical data.

 **Enter Patient Information**

Age (years)

50

Creatinine (mg/dL)

1.20

BUN (mg/dL)

20.00

Diabetes

☒ No ☐ Yes

Hypertension

☒ No ☐ Yes


GFR (ml/min/1.73m²)


90.00

Urine Output (ml/day)


1500.00


Predict

 **CKD Prediction**

 No signs of CKD detected at this time.

CKD Probability: 0.04%

 **Dialysis Risk Prediction**

 No immediate indication of dialysis need.

Dialysis Probability: 0.00%

Portfolio

Nasser CHAOUCHI

**DON'T HESITATE TO  
REACH ME OUT**

