

### Contact

Ypenburgbocht 9-A303 2316 WB Leiden, Zuid-Holland Netherlands

#### Socials

robkras.com Kaggle: robkraseu GitHub: rbkrs

robkraseu@gmail.com +31 6 27390907

#### Languages

Dutch – Native English – Bilingual German – Limited French – Limited

# Robin (R.P.M.) Kras

#### **EDUCATION**

Estimated finalisation Mid 2025

## MSc Computer Science, Rijksuniversiteit Leiden

- Data Science & Artificial Intelligence specialization track

Dec 2023

## BSc Computer Science, Vrije Universiteit Amsterdam

- Minor: Data Science

### **SKILLS**

Programming: Python, JavaScript, Assembly, C, C++, SCALA

**Technical skills:** Keras, NumPy, Pandas, TensorFlow, (My)SQL, data manipulation, data visualization, machine learning, GIT, PowerPoint, web scraping, data mining, natural language processing

**Personal skills:** love of learning, time management, communication, leadership, management, excellent swimmer, adaptability

# **CERTIFICATES**

## **English C1 Advanced**

May 2018

Issued by Cambridge Assessment International Education

## **PROJECTS**

## Kaggle

Competition entries are regularly updated and stored on my domain (robkras.com).

# Notable entries include:

Rainfall prediction (5/2529): rainfall (robkras.com)
Survival prediction (2359/15257): titanic (robkras.com)
House pricing prediction (37/3942): house-prices (robkras.com)

## Machine Learning & Data Science

- Developed supervised learning models (XGBoost, kNN, Random Forest, Linear Regression) to predict rainfall and classify Titanic survival outcomes.
- **Optimized models** using GridSearchCV, achieving improved accuracy and efficiency.
- Engineered new features, handled missing data, and applied one-hot encoding & label encoding for categorical variables.
- Trained and fine-tuned deep neural networks using TensorFlow/Keras.

#### Data Processing & Analysis

- Conducted Exploratory Data Analysis (EDA) using Seaborn & Matplotlib to identify trends and correlations.
- Cleaned and preprocessed datasets using Pandas & NumPy, ensuring high-quality input data.
- Scaled numerical features using **StandardScaler** to improve model convergence.

# Model Evaluation & Interpretability

- Assessed models with RMSE, R<sup>2</sup>, MAE, accuracy, and ROC-AUC scores for performance benchmarking.
- Applied SHAP values for explainability and feature importance analysis.
- Used SMOTE to balance imbalanced datasets, improving prediction robustness.