



# Robin (R.P.M.) Kras

## Contact

Ypenburgbocht 9-A303  
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## Socials

robkras.com  
Kaggle: robkraseu  
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## Languages

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

## 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, C, C++, Assembly, SCALA

**Technical skills:** Keras, NumPy, Pandas, TensorFlow, PyTorch, (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, 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 (2331/15346): [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,  $R^2$ , 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.