Group 3: Regression

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Introduction

A Brief Recap

- Initial Plan:
 - Focus: Linear Regression
 - Task: Implement five regression models
 - Validation: Compare with established implementations
- Received Feedback:
 - Focus on 1 or 2 models
 - Add educational value to the project

Introduction

Our New Approach

- Chose OLS and LWR as focus models
- Retained comparative study
- Developed two web applications
 - One for education
 - One for model visualization

2. Coding

- Dev setup
- SWE principals

2.1 Coding - Dev setup

- Fixed python version via pyenv (python 3.11)
- Dependencies loaded via pip into a virtual env of the venv module of the standard lib
- Code formatting with black
- Linting with pylint
- Reviews of pushed code
- Usage of github issue tracker

2.2 SWE principals -- Separation of concerns

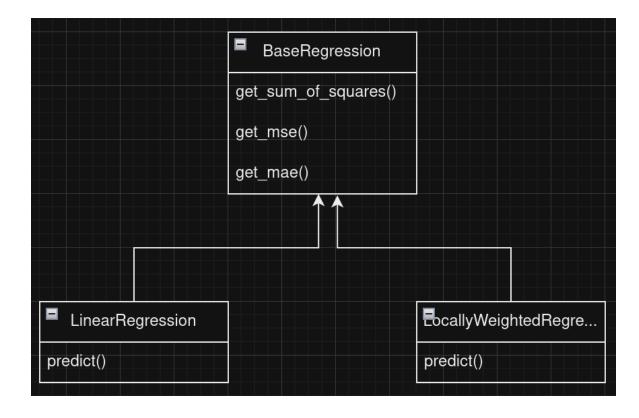
```
✓ kap src

  > IIII apps
    > 📴 data
    > evaluation

✓ models

         __init__.py
         base_regression.py
         linear_regression.py
         locally_weighted_regression.py
    > u visualization
      __init__.py
    __init__.py
```

2.3 SWE principals -- Inheritance



3. Locally Weighted Regression (LWR) in Detail

3.1 LWR in Detail

- 1. Divide intersections
- 2. For each section, calculate the weighted regression with weight

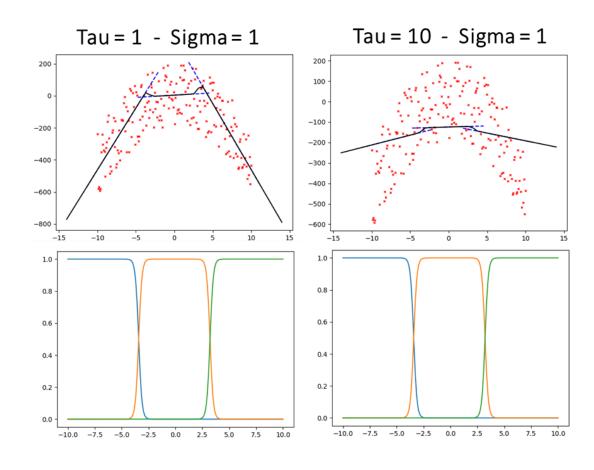
$$w_i(x) = e^{-rac{(centre_i-x)^2}{2 au^2}}$$

3. Smoothen the function with $gauss_{centre}(x) = e^{-\frac{(centre-x)^2}{2\sigma^2}}$ and normalising it by dividing through $\Sigma_{centre} \ gauss_{centre}(x)$ $f(x) = \frac{1}{\Sigma_i \ gauss_i(x)} \Sigma_i gauss_i(x) \cdot f_i(x)$

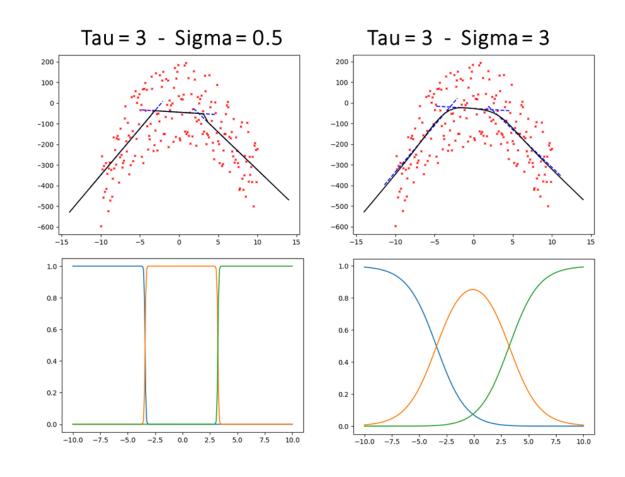
Hyperparameters:

- amount sections
- tau
- sigma

3. LWR in Detail - Influence of Tau



3. LWR in Detail - Influence of Sigma



4. Runtime Performance - Setup

Time Complexity:

- 50 iterations
- 50 data points
- 2 independent variables

Memory Consumption:

- 1 iteration
- 500 samples
- one independent variable

4. Runtime Performance - Results

	Time Complexity	Memory Consumption
Linear Regression	0.138s	OMiB
skearn.linear_model	0.060s	OMiB
Locally WeightedRegression	56.500s	45.3MiB
localreg	2.170s	OMiB

5. Live Demo