

# Identifying dominant models when the noise context is known

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# Outline

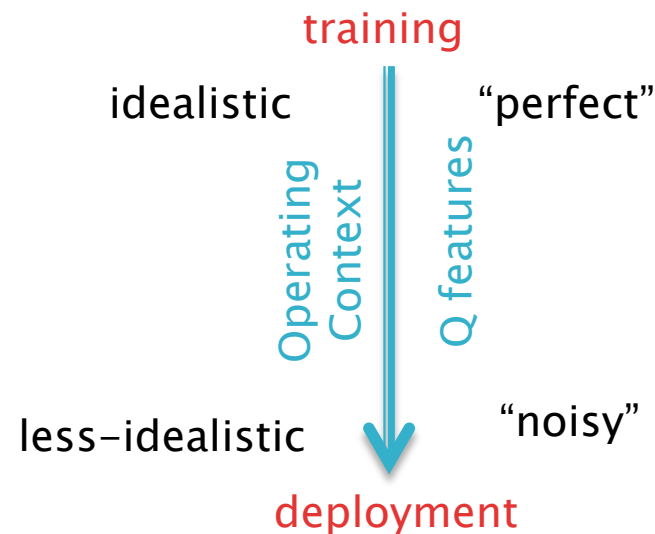
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- ❑ Motivation
- ❑ The Noise setting
- ❑ Context Plots and Dominance
- ❑ Experiments
- ❑ Conclusions and Future Work

# Motivation

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- Very often, operating contexts (OC) at the training and the deployment time are different.



- The error of the model depends on the level of noise introduced by the OC.

# Motivation: An example

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- Alarm system where two models (A and B) have been trained (ideal conds., OC=temperature in [0,30])
  - Validation:  $A > B$  (A is better!)
- Deployment
  - How these OC affect to sensors is also known
  - OC in deployment are given
- Which model is better for each OC? .... For several OC, model B could be better now!

# Motivation

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In order to answer this question we propose:

- ❑ To evaluate the models with different levels of simulate noise.
- ❑ To draw a context plot with all models, and to determine dominance regions.
- ❑ During deployment, the noise level is derived from the OC and the best model for that noise is applied.

# The Noise setting

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Noise is calculated by using probability distributions:

## □ Numerical attributes

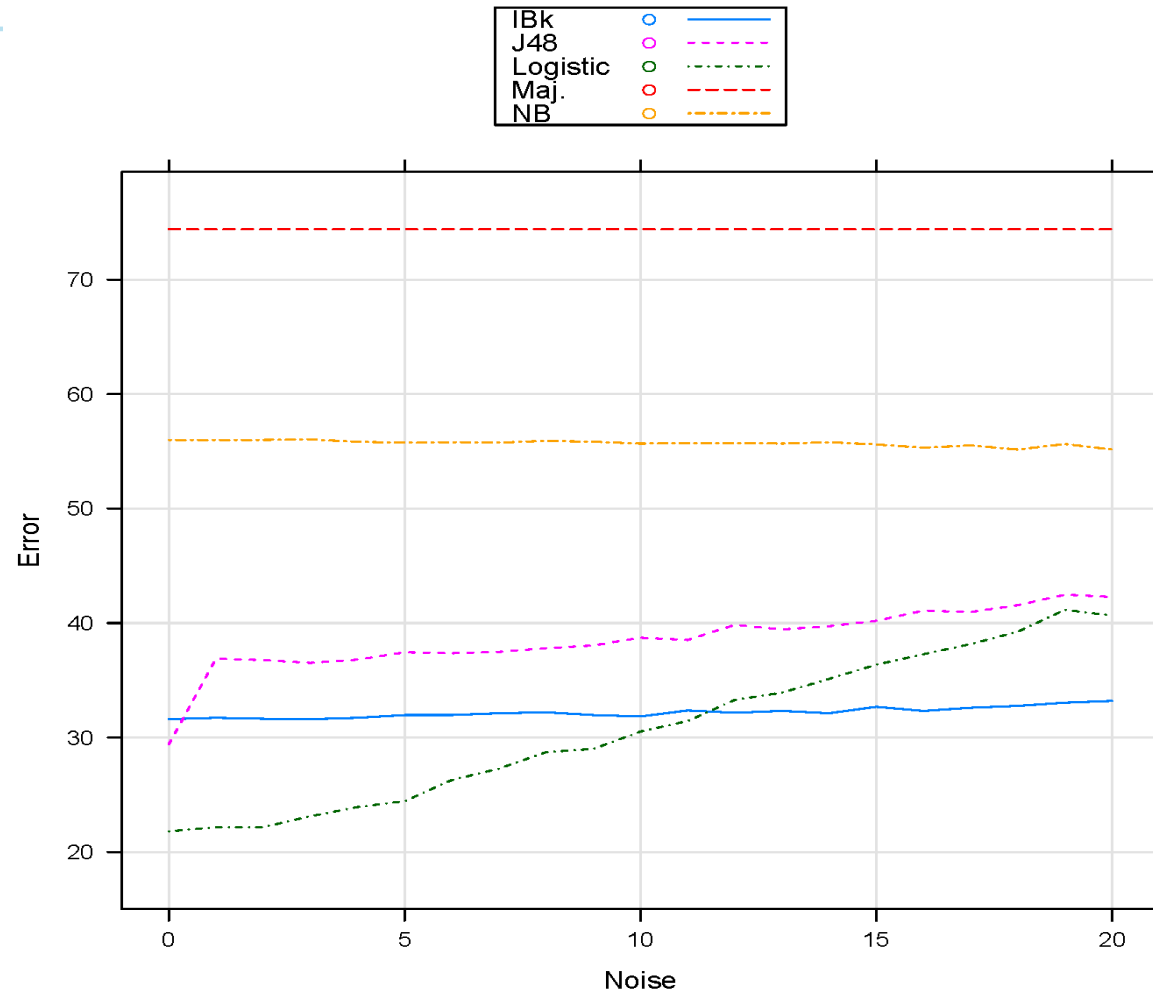
- we estimate the  $\sigma$  of all values of the attribute
- for a level of noise  $\nu$ , we modify a value  $x$  using a normal distribution  $x' \sim N(x, \sigma \cdot \nu)$

## □ Nominal attribute

- we estimate the frequency of each value  $v_i$ ,  $p = (p_{v_1}, \dots, p_{v_n})$
- for an instance  $x$  with value  $v_i$  we estimate the vector
$$t = (t_1, \dots, t_n) \quad t_i = 1 \text{ and } t_j = 0 \text{ if } i \neq j$$
- for a level of noise  $\nu$  we calculate  $p' = \alpha \cdot p + (1 - \alpha) \cdot t$  where  $\alpha = 1 - e^{(-\nu)}$
- we use  $p'$  to sample the new value  $x'$

# Context Plots and Dominance

## Vehicle



# Experiments

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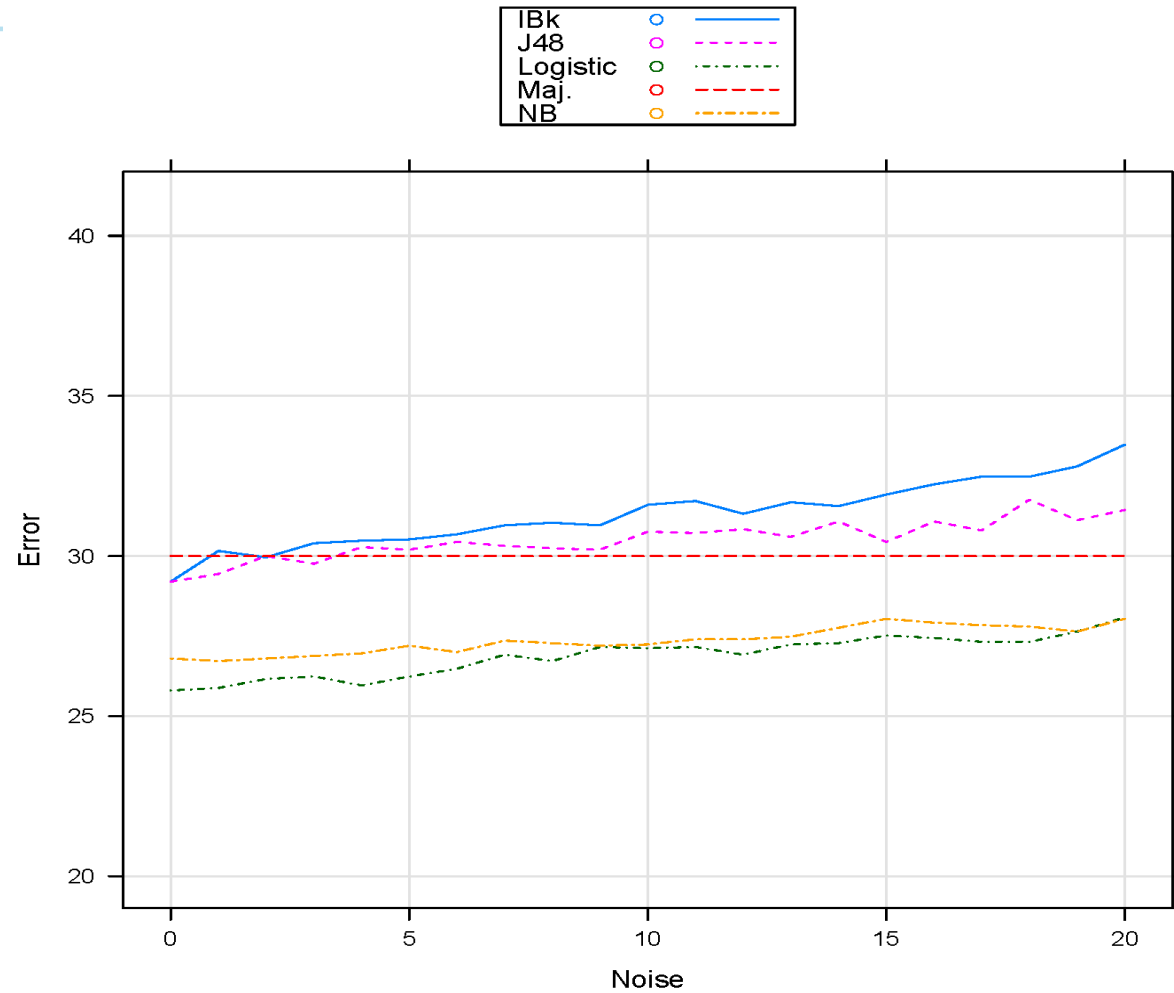
## Methodology:

- 12 datasets from the UCI repository
- 50% Train, 25% Validation, 25 % Test
- Classification:
  - J48, Naive Bayes, Logistic Regression and kNN.
  - Reference method: majority class.
  - Classification error
- Regression:
  - Linear Regression, M5P, kNN , SMOreg.
  - Reference method: ZeroR.
  - Relative absolute error



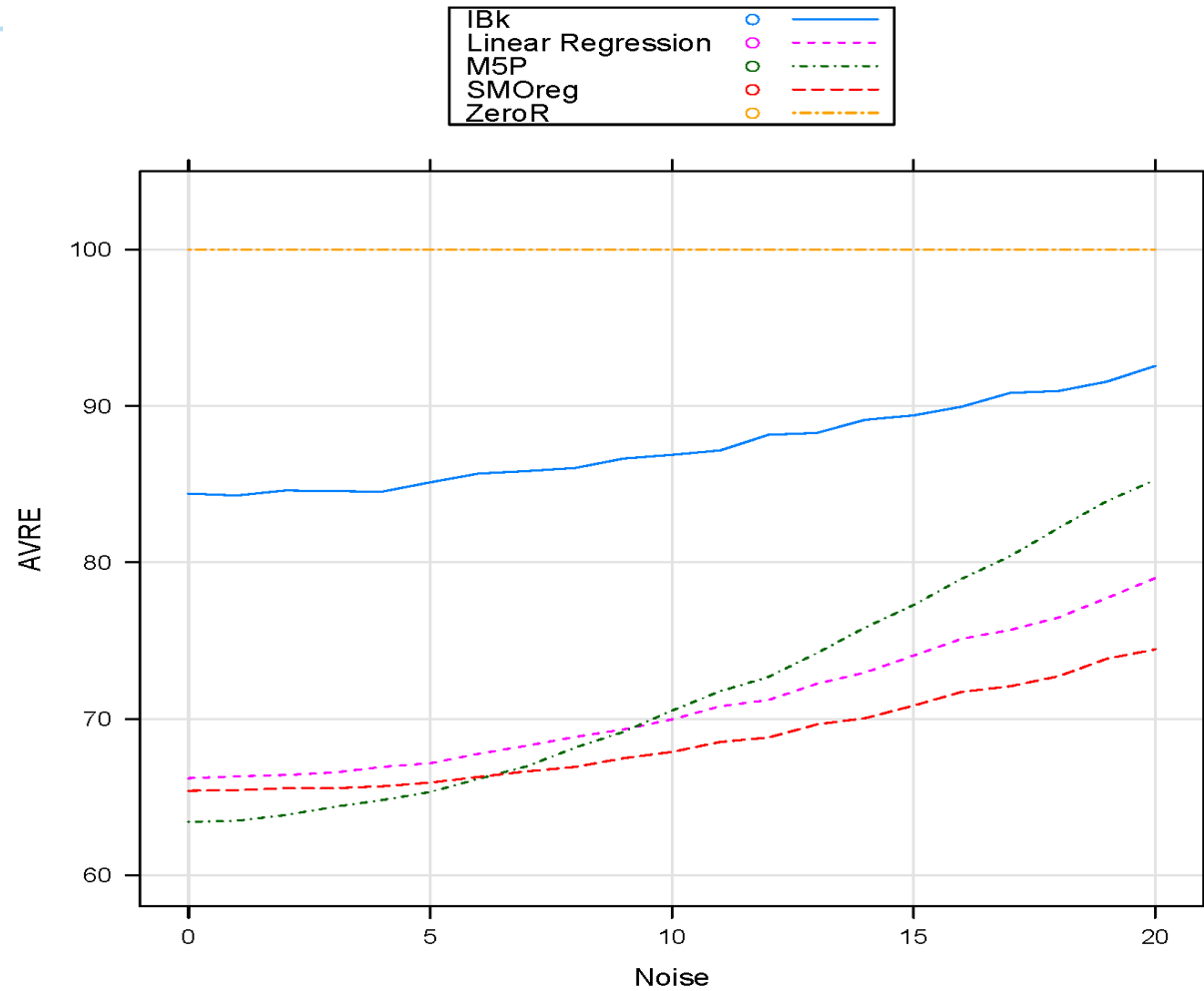
# Context Plots and Dominance

Creditg



# Context Plots and Dominance

## Abalone



# Experiments

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## Methods:

- **ValNoNoise:** For all the estimated values of noise, we select the method that obtains the **best performance without noise**.
- **ValBestArea:** For all the estimated values of noise, we select the method that obtains the best performance in the validation dataset by averaging all noise levels (i.e., **the curve with lowest area in the context plot**).
- **ValNoiseOpt:** For each of the estimated values of noise, we select the method with **best performance** in the validation dataset **with that value of noise**.
- **Idealistic:** For each of the estimated values of noise, we select the model **with best performance in the test dataset with that value of noise**. This strategy is not realistic (it cannot be done in practice). We just include this as a reference

# Experiments

| Dataset          | ValNoNoise | ValBestArea | ValNoiseOpt |
|------------------|------------|-------------|-------------|
| 1 cholesterol    | 97.5468    | 97.5468     | 97.5468     |
| 2 bodyfat        | 99.2007    | 99.2007     | 99.9875     |
| 3 wisconsin      | 91.9391    | 91.9391     | 92.1545     |
| 4 stock          | 100.0000   | 100.0000    | 100.0000    |
| 5 meta           | 100.0000   | 100.0000    | 100.0000    |
| 6 abalone        | 81.6546    | 99.9358     | 99.9919     |
| 7 iris           | 89.3179    | 89.3179     | 96.4559     |
| 8 white-clover   | 89.9206    | 89.9206     | 88.3807     |
| 9 hepatitis      | 100.0000   | 100.0000    | 100.0000    |
| 10 breast-cancer | 87.6362    | 87.6362     | 88.2716     |
| 11 vehicle       | 75.4220    | 96.1937     | 99.6372     |
| 12 credit-g      | 100.0000   | 100.0000    | 100.0000    |

- The performance results are normalised by the Idealistic performance

$$(\text{idealistic}/\text{method}) * 100$$

# Conclusions and Future Work

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1. In this paper we have analysed the case when:

- ❑ the noise level depends on a context
- ❑ we know the context in advance

2. The model that best behaves for each noise level situation is used.

3. CONS: It takes some time. PROS: Selection/application of the best model is straightforward with results close to an idealistic process.

4. As a future work we plan working on different noise models, derived from scenarios with real operating conditions and with real sensors.

- Each attribute will have a different operating range, but the context would still be given by a single parameter (e.g., temperature, or the number of measurements performed..).

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Thanks for your attention.....