

Exercise 1:

a.) supervised learning:

- b.) - learning a function with a set of input-output pairs
- the data from which the function is learned is already tagged with the correct answer
- "supervised" because the correct answers are already given

Problem:

- The website classification in Task 6
- At first collect the Training Data via manual annotation of the websites
- determine the features that should be used in the model
- train the model with the Training Data
(Algorithms could be SVM, Neural Networks, Regression, Classification trees)
- for new data the trained model can estimate a class as an output

unsupervised learning:

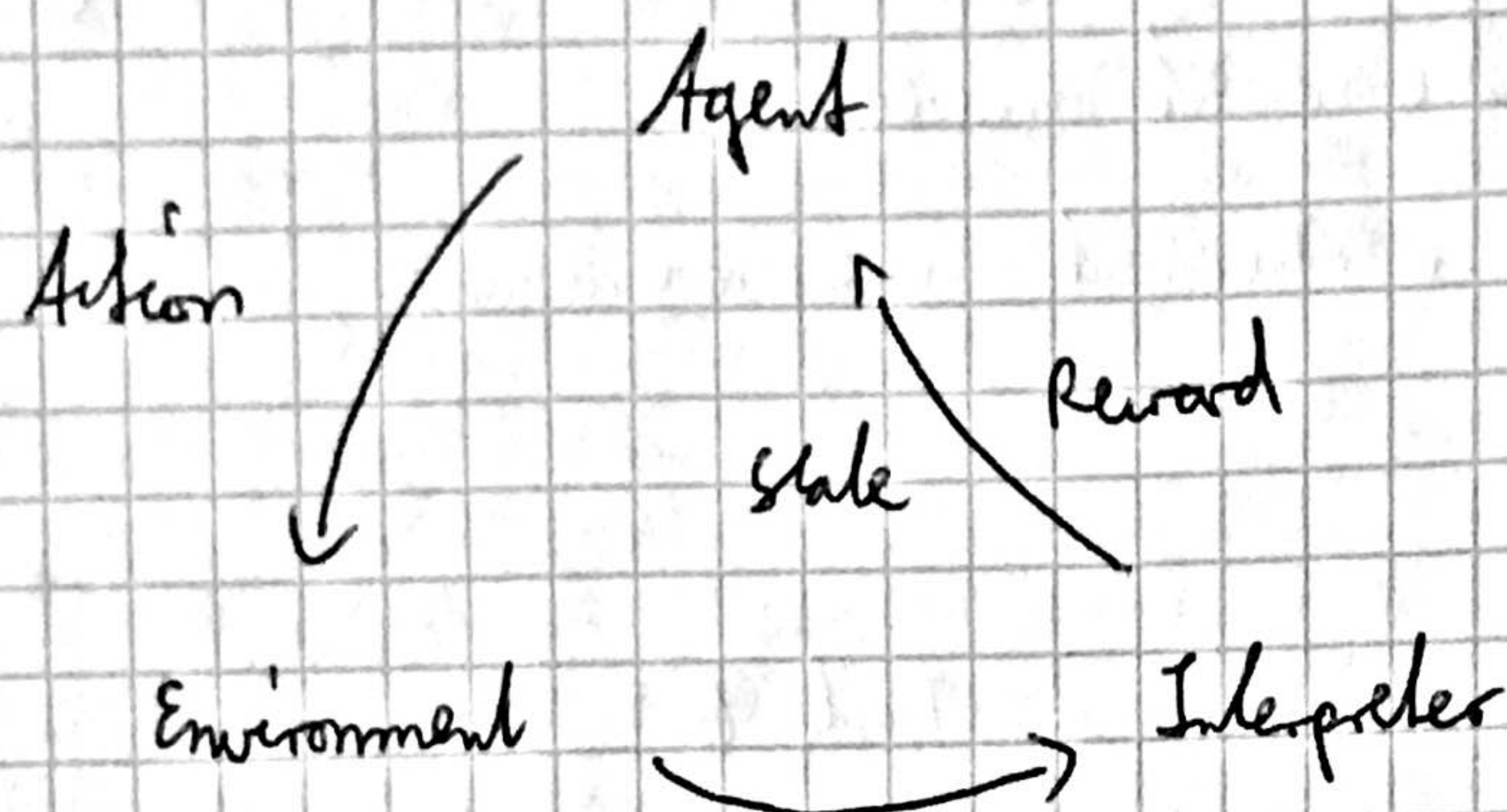
- identifying structures in unlabeled/uncategorized data (e.g. clustering)
- only use input data to learn, no output data
- important subareas: ^{eg:} - automated categorization (via cluster analysis)
- parameter optimization ^{eg:} (via expectation minimization)
- feature extraction (e.g. via factor analysis)

Problem:

- detecting different classes of wine based on example data of many different wines with a couple of features
- e.g. via ~~support vector machine~~ cluster analysis

Reinforced learning:

- learn / adapt / optimise a behaviour strategy in order to maximise the own benefit / reward by interpreting feedback from the environment
- basic reinforcement is modeled as a "Markov decision process" :-



Problem:

- (- computer agent learning to be "friendly" in a conversation ?)
- learning behaviour strategies in hostile environments
- chess - bots, Go - Bots, elevator scheduling, robot control, checkers, backgammon, (games in general ?)

Exercise 2:

- | | |
|---------------------|-----------------------|
| a1) a1) Objects O | a2) Feature Space X |
| a3) y | a4) x |
| a5) classes C | a6) $c \approx x$ |
| b.) 6a) y | 6b) x |

Exercice 3:

a.) $y(x) = w_0 + w_1 \cdot x$

$$RSS(w_0, w_1) = \sum_{i=1}^n (y_i - w_0 - w_1 x_i)^2$$

$$\hat{w}_0 = \bar{y} - \hat{w}_1 \cdot \bar{x}$$

$$\hat{w}_0 = \frac{553}{4} - \hat{w}_1 \cdot \frac{55}{4}$$

$$\hat{w}_1 = \frac{\sum_{i=1}^n (x_i - \bar{x}) \cdot (y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$\bar{x} = \frac{5 + 7 + 15 + 28}{4} = \frac{55}{4}$$

$$\bar{y} = \frac{50 + 79 + 124 + 300}{4} = \frac{553}{4}$$

$$\hat{w}_1 = \frac{13837}{1307} \approx 10,6$$

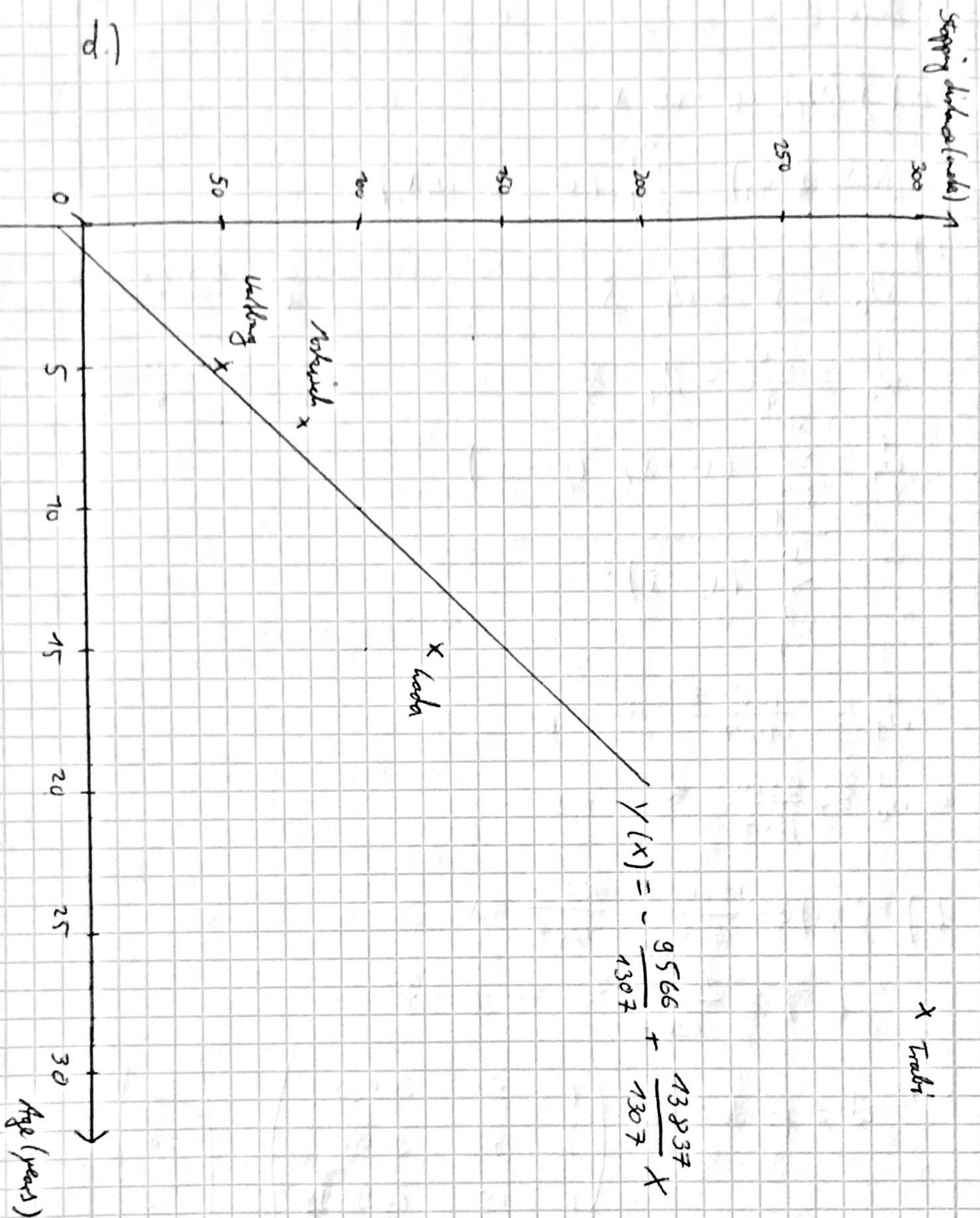
$$\hookrightarrow \hat{w}_0 = -\frac{9566}{1307} \approx -7,3$$

b.) $y(x) = -\frac{9566}{1307} + \frac{13837}{1307} \cdot x$

$$y(15) = \frac{197989}{1307} \approx 151,5$$

c.) \hat{w} = pseudo inverse

$$\begin{pmatrix} 1 & 5 & 30530 \\ 1 & 7 & 90000 \\ 1 & 15 & 159899 \\ 1 & 28 & 270564 \end{pmatrix} \cdot \begin{pmatrix} 50 \\ 79 \\ 124 \\ 300 \end{pmatrix}$$



- e.) - can't model the exact searched function with linear regression, only a linear estimate
- one should be aware that the extrapolations can be different from the real values (searched function could be much more complex and depending on many more features)

4a) with EXAMPLE: $(x, c(x))$ // concept is attribute of EXAMPLE
 $u(p) = u_1 \cdot \dots \cdot u_p \cdot 2$ // u_i is number of values for an attribute
 $= \prod_{i=1}^p u_i \cdot 2$ // concept $c(x)$ has 2 values

4b) syntactically distinct hypotheses $= \prod_{i=1}^p (u_i + 2)$
 semantically distinct hypotheses $= 1 + \prod_{i=1}^p (u_i + 1)$

4c) $\begin{cases} u(1) = u_1 \cdot 2 \\ u(p+1) = u(p) \cdot u_{p+1} \end{cases}$
 $\begin{cases} |H_1| = u_1 + 2 \\ |H_{p+1}| = ((|H_p| - 1) \cdot (u_{p+1} + 1)) + 1 \end{cases}$

// a hypothesis is described as without concept, but as a tuple of the remaining other feature values
 $\leftarrow \text{with } x \in \{1, \dots, u_i\}$ the remaining other feature values
 (vgl. page 9 < sunny, ?, ?, strong, ?, same ? >)
 // 1 because of contradiction, \perp
 // $u_i + 1$ because of 2^n

5a) consistent hypothesis: $h(x) = c(x)$
 Inconsistency is caused, when $c(x) = 0$, then the inner if-cause isn't entered and reached
 and the chance of correcting an $h(x) = 1$ is zero.
 • IF $c(x) = 0$ then then if $h(x) = 0$ consistent, no chance of making it consistent
 $h(x) = 1$ inconsistent (with min-generalisation $Ch(x)$)
 (with min-generalisation $Ch(x)$)

5b.)

No, the order of the examples is not important.

There are 4 cases when looking at an attribute.

(written like: (attribute value hypothesis, attribute value example) -> attribute value hypothesis,

a, b and c are examples for literals)

1) most specific to less specific (\perp, a) -> a

2) same attribute-values: (a, a) -> a

3) different attribute values: (a, b) -> ?

4) most general and a any attribute value (?, c) -> ?

A) Given more than one example (- the first example changes the hypothesis to a copy of itself-),

the first value different to the value of the first example, will cause a change to ? (3))

everything afterwards is of no importance (4)).

-> not dependent on order because if there is a different value it will be taken into account at any time

B) If there are no different attribute values in the examples, the initial value will persist. (2))

-> not dependent on order, cause the attribute values are equal

So the outcome is either A) or B) and there is no Order