

# Iteratieve Algoritmen

Bas Terwijn <[b.terwijn@uva.nl](mailto:b.terwijn@uva.nl)>

Heuristieken / Programmeertheorie

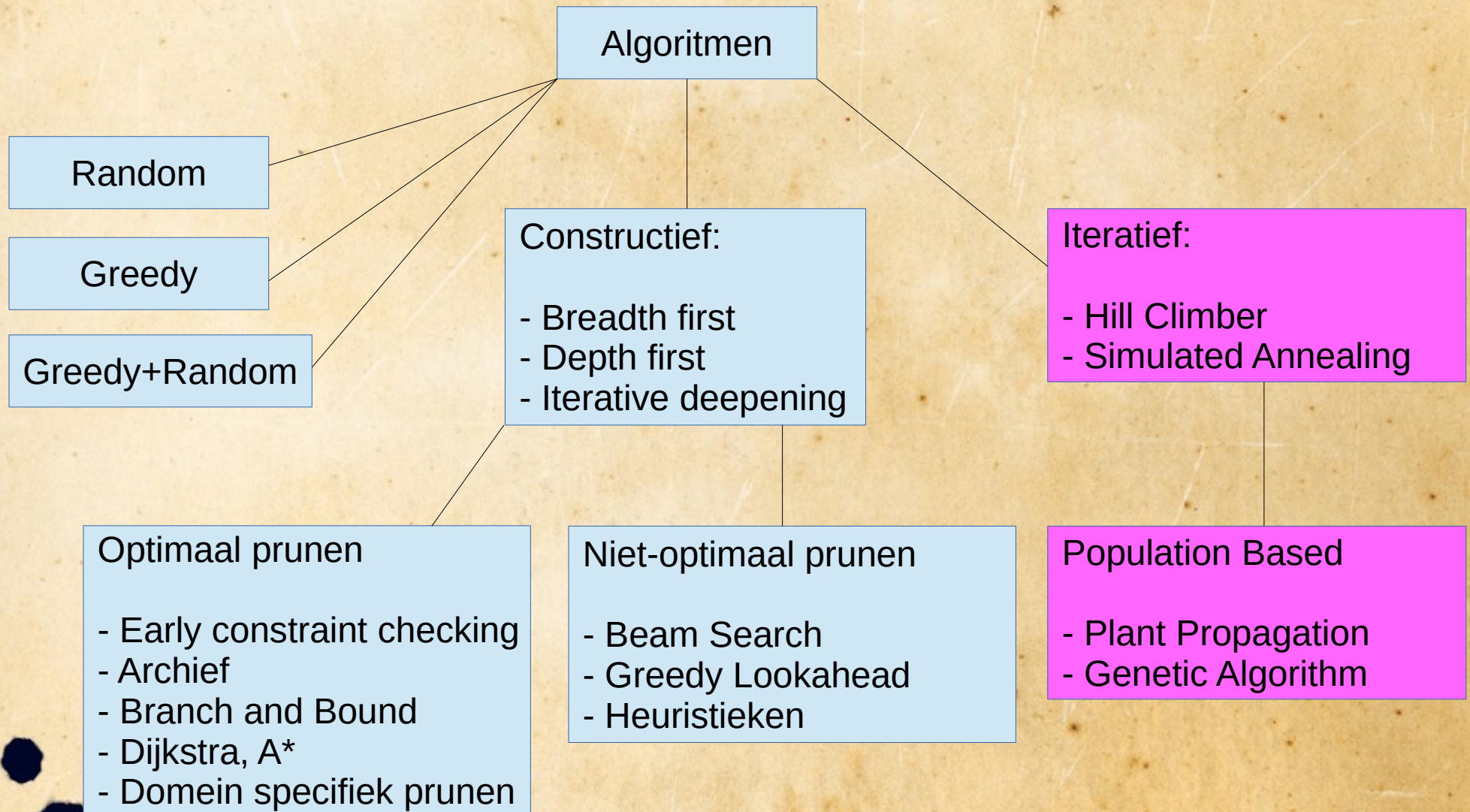
Minor Programmeren

Universiteit van Amsterdam





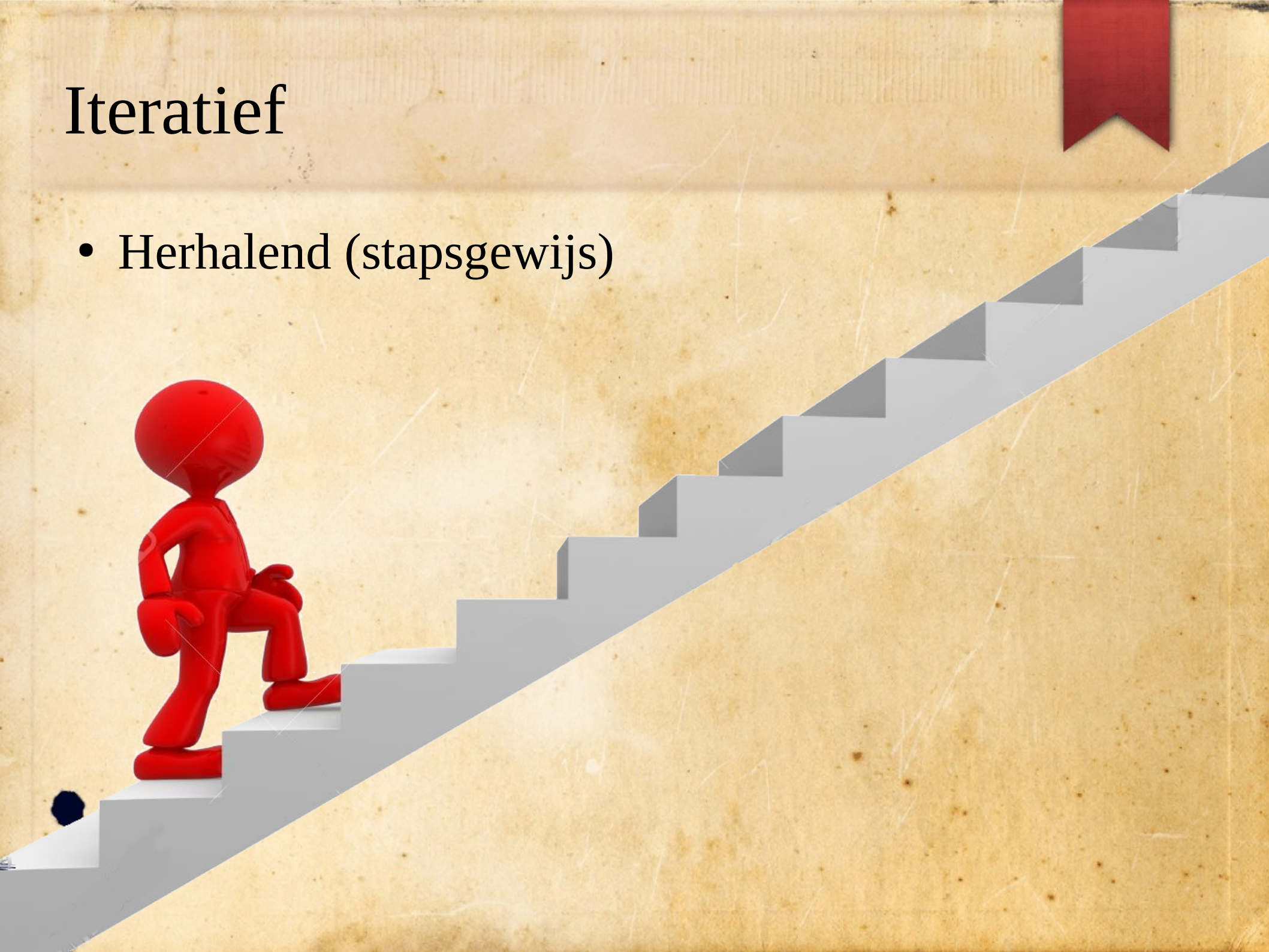
# Algoritmen





# Iteratief

- Herhalend (stapsgewijs)





# Hill Climber, pseudo code

Kies een random start state

Herhaal:

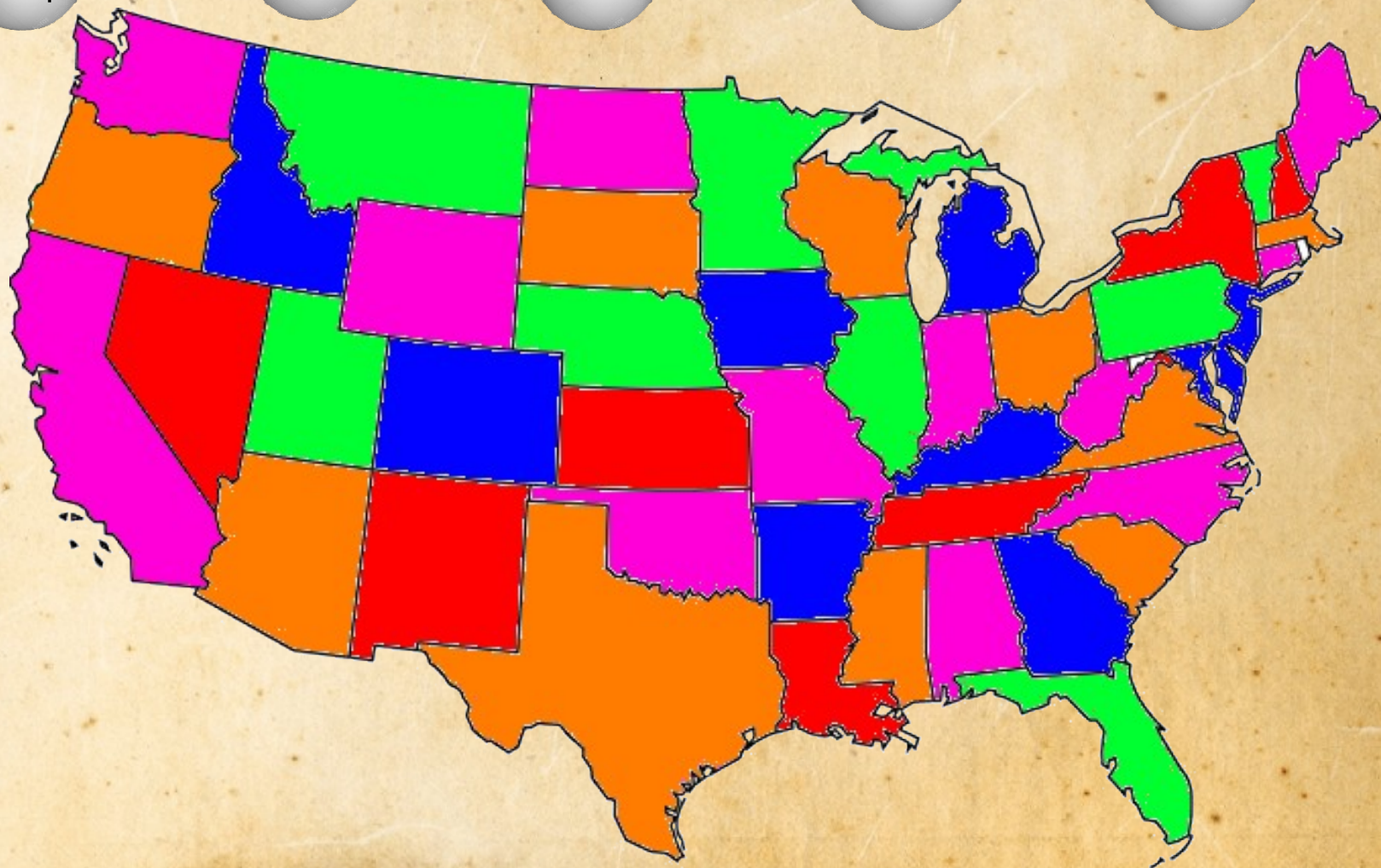
- Doe een kleine random aanpassing

- Als de state is verslechterd:

  - Maak de aanpassing ongedaan

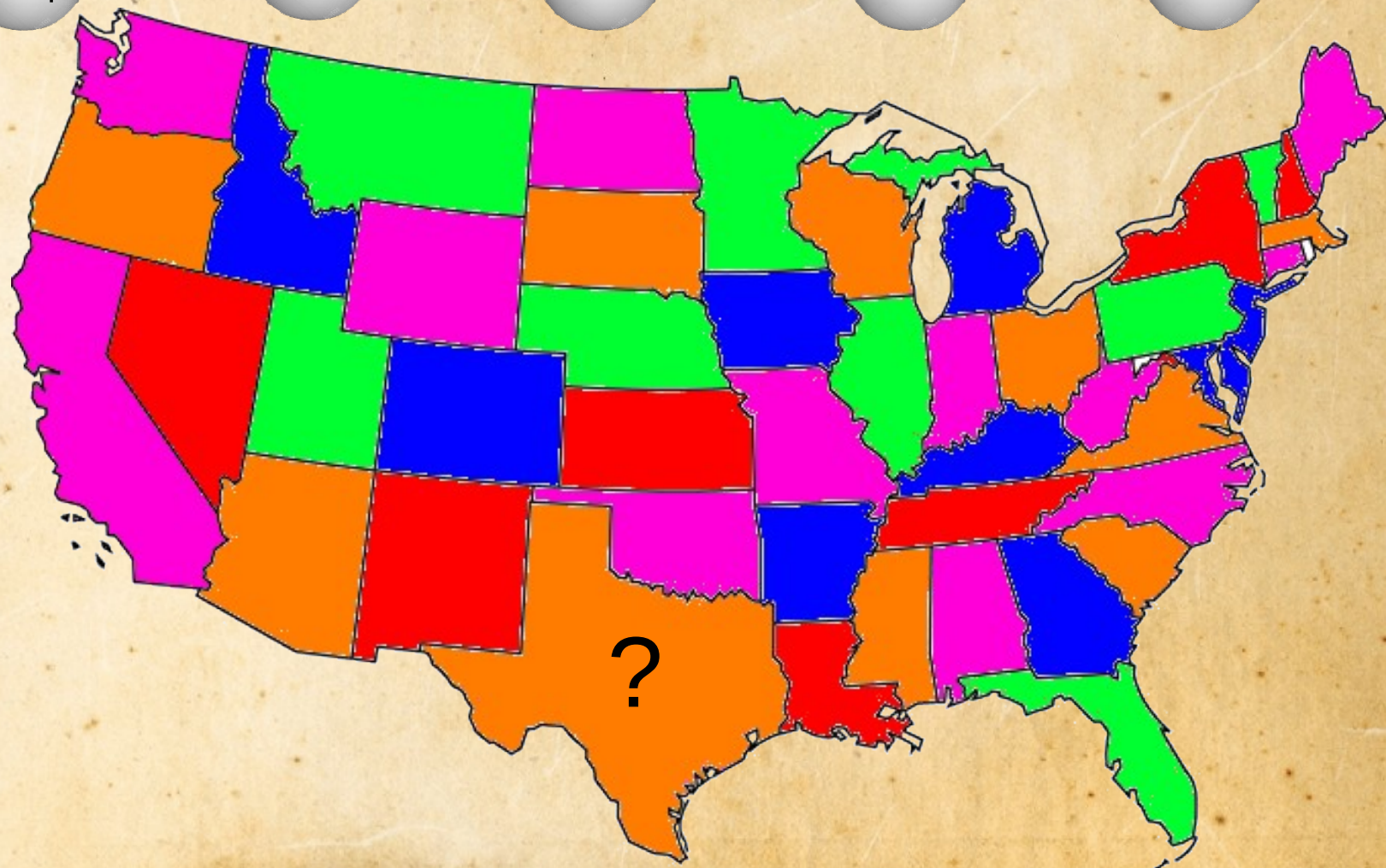


# Hill Climber



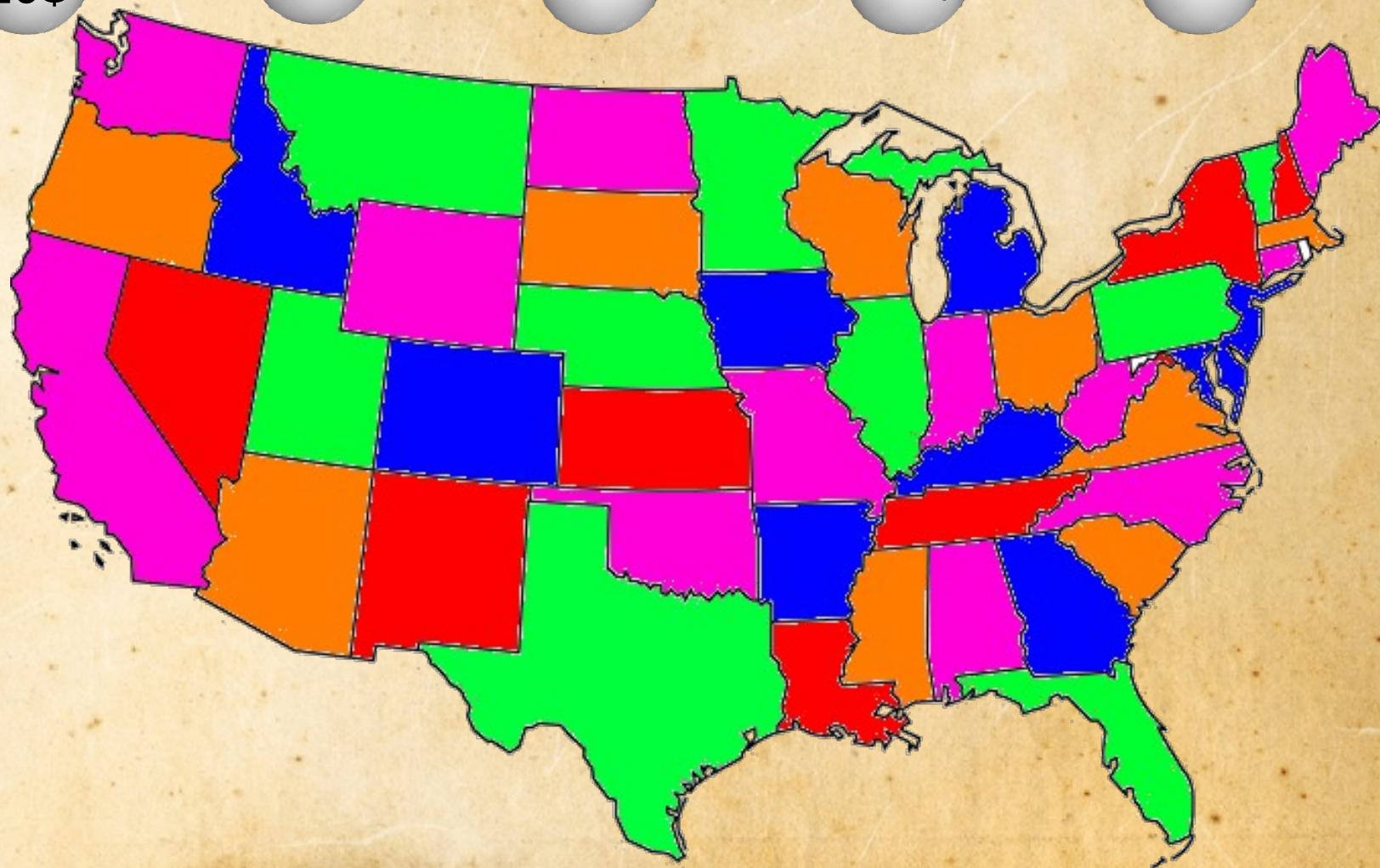


# Hill Climber



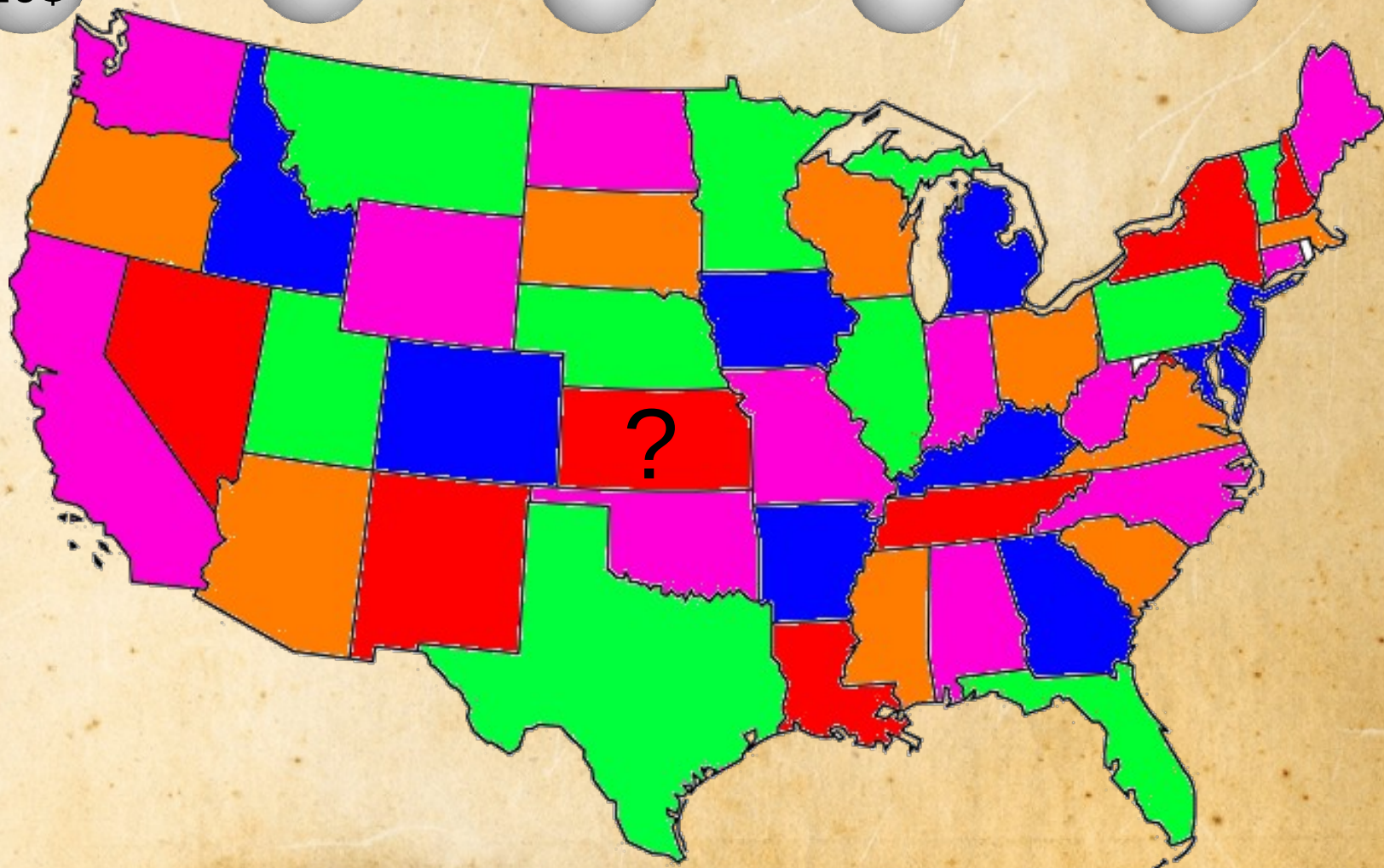


# Hill Climber



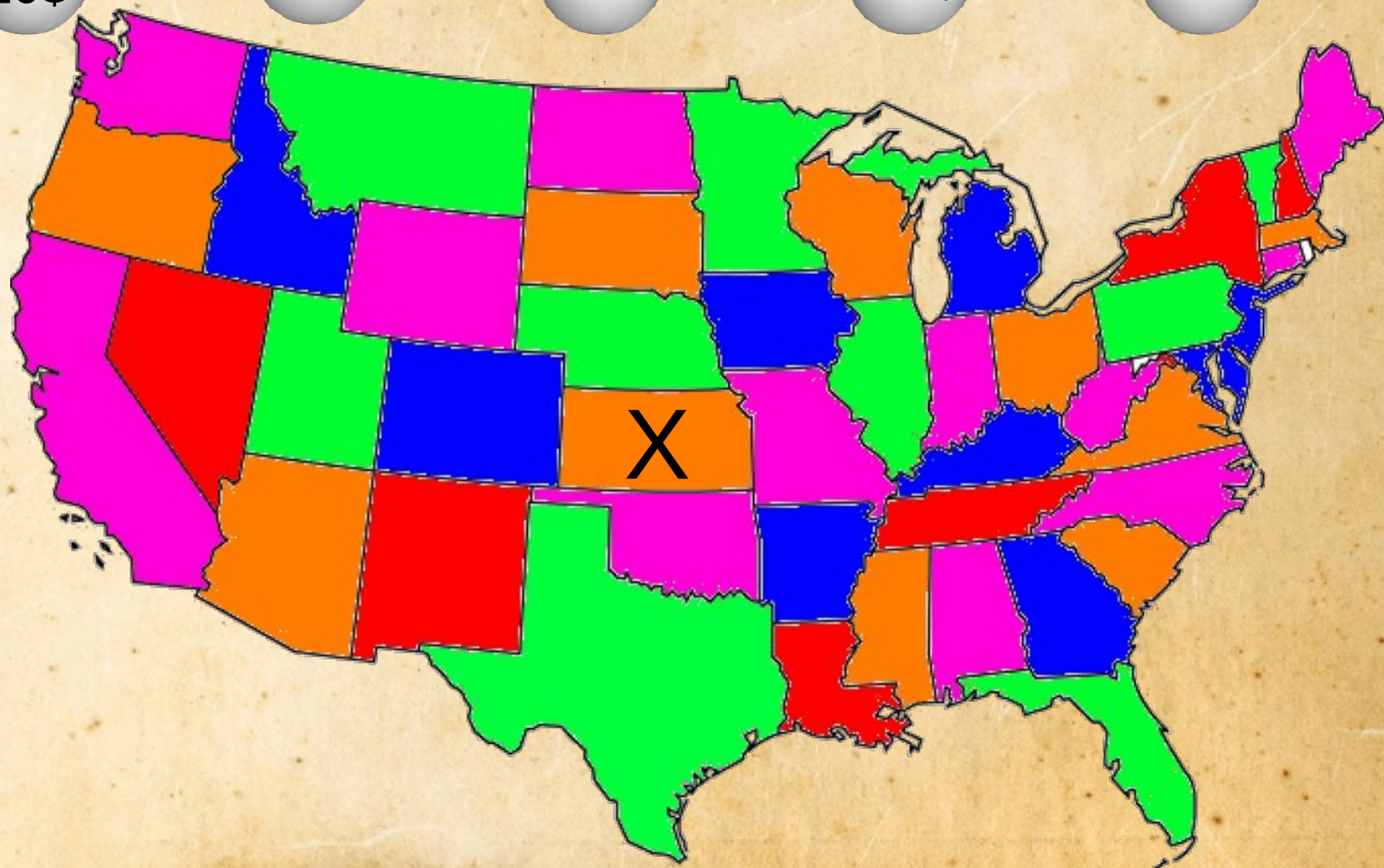


# Hill Climber



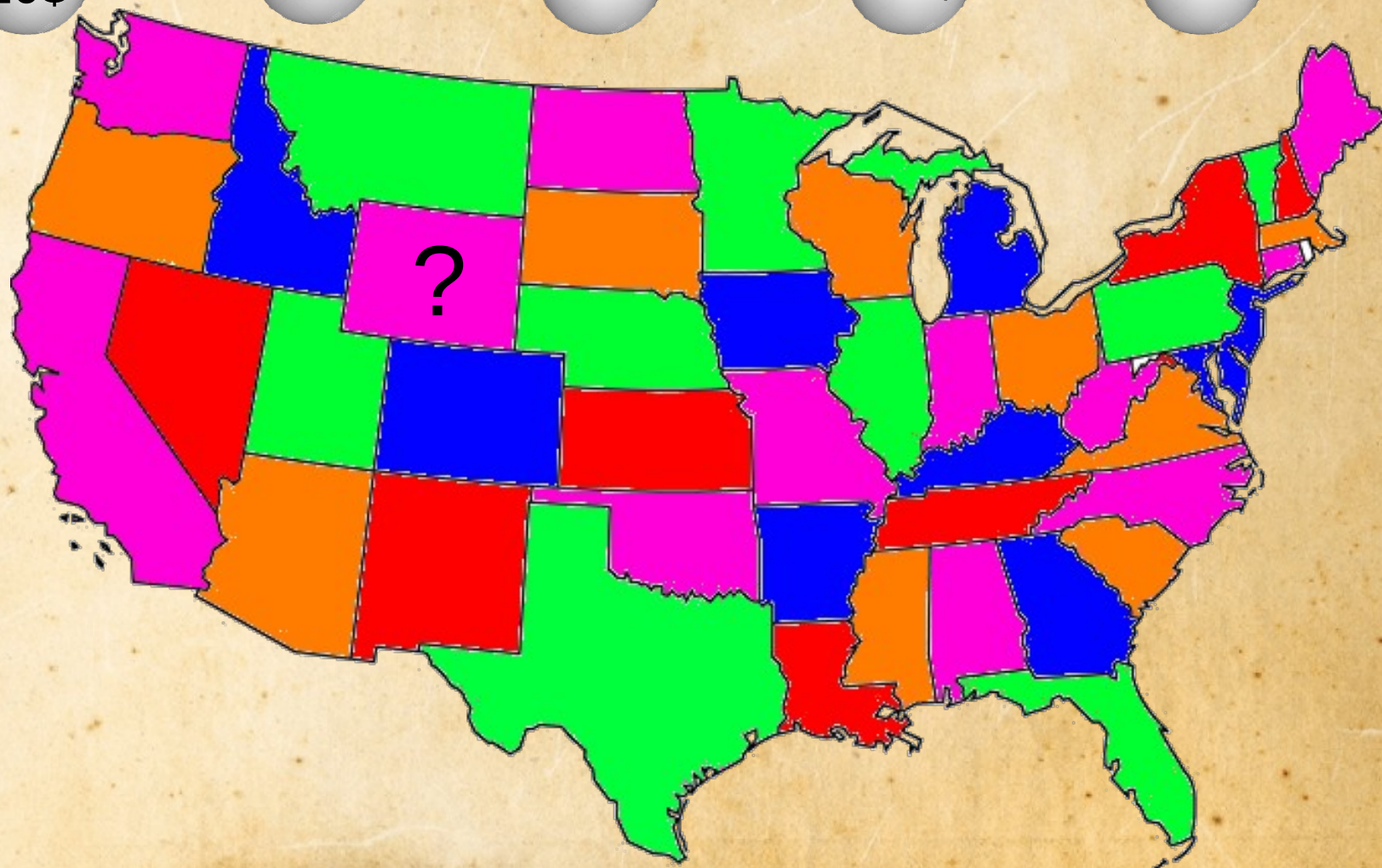


# Hill Climber



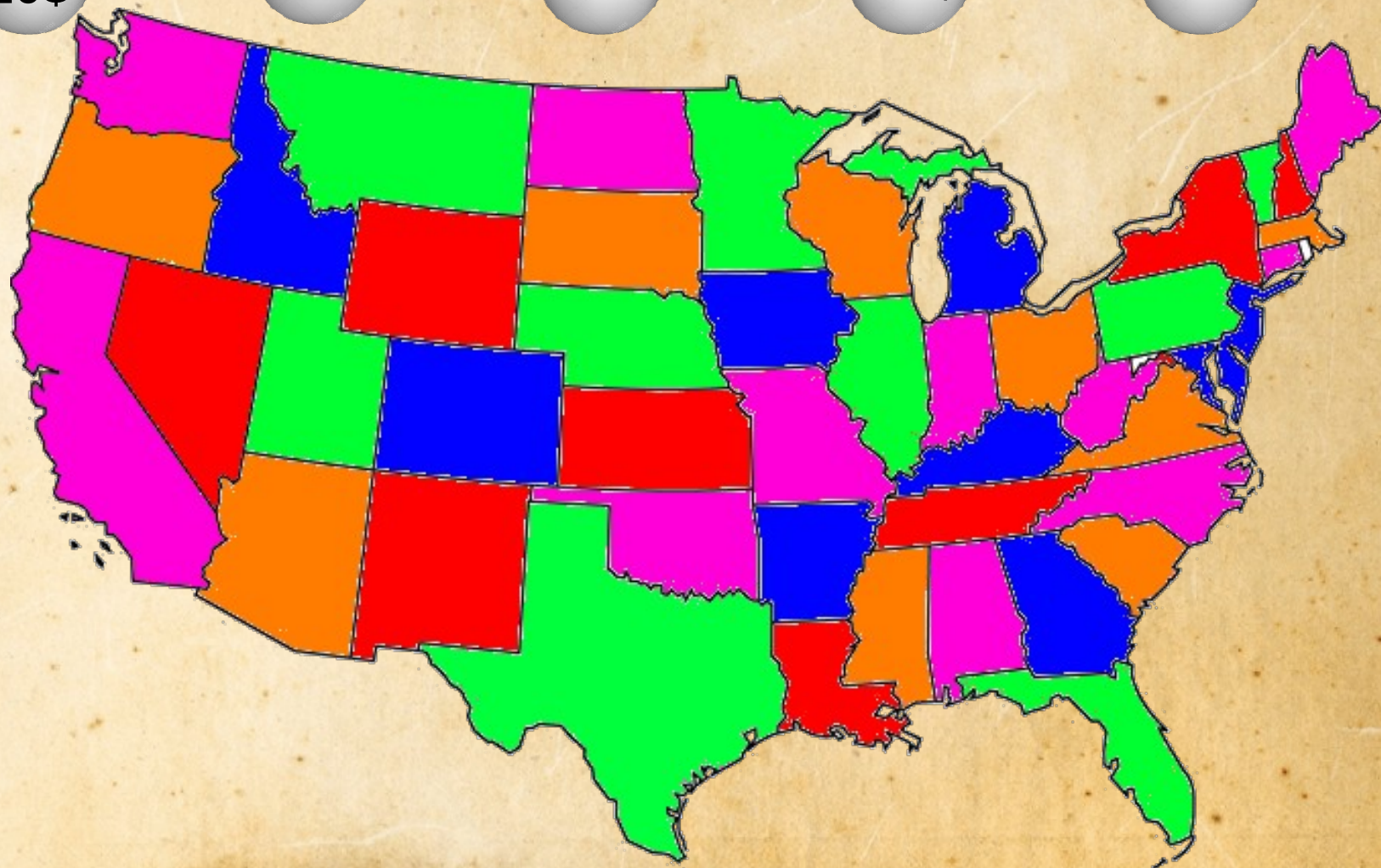


# Hill Climber





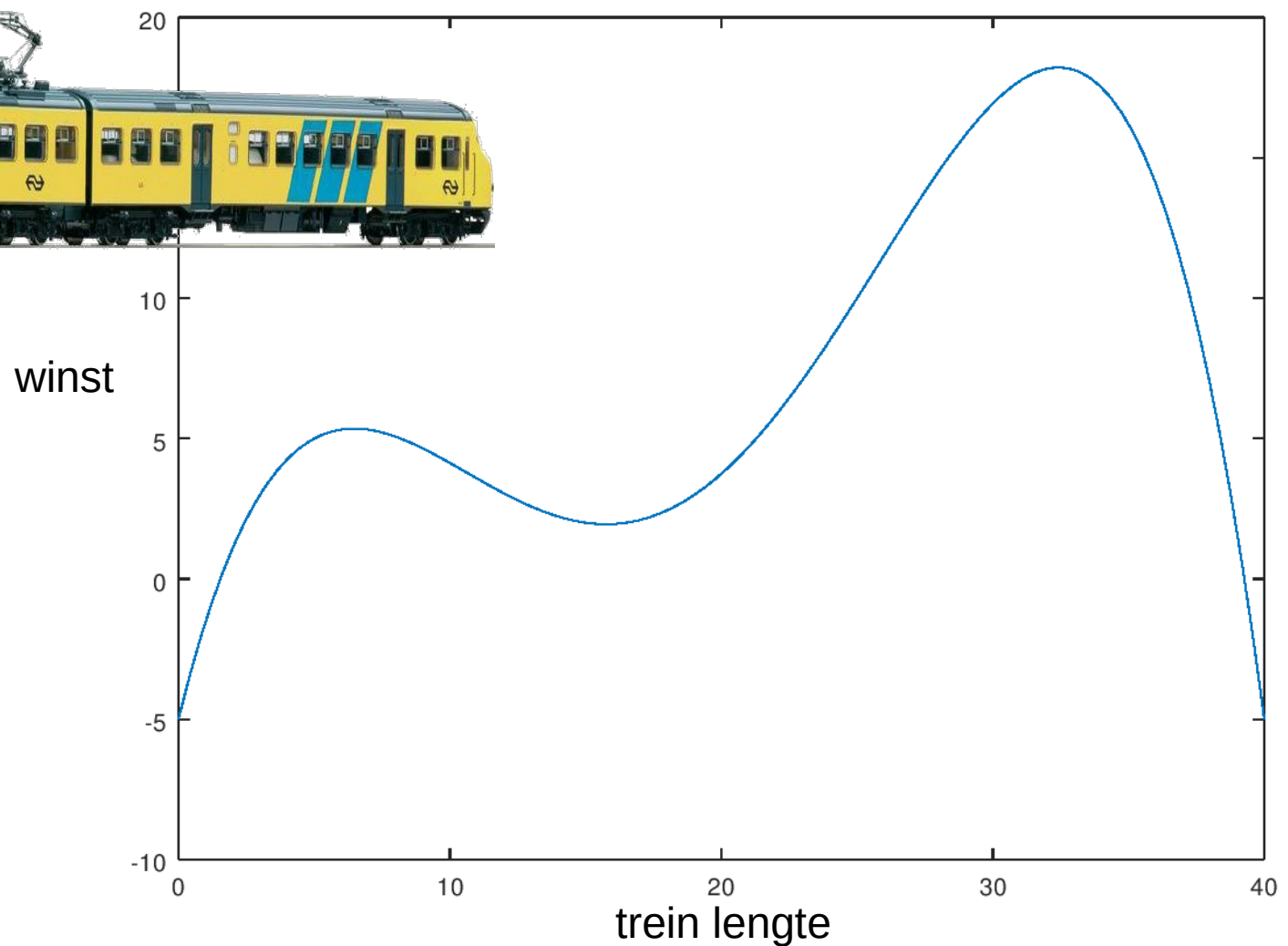
# Hill Climber





# Hill Climber

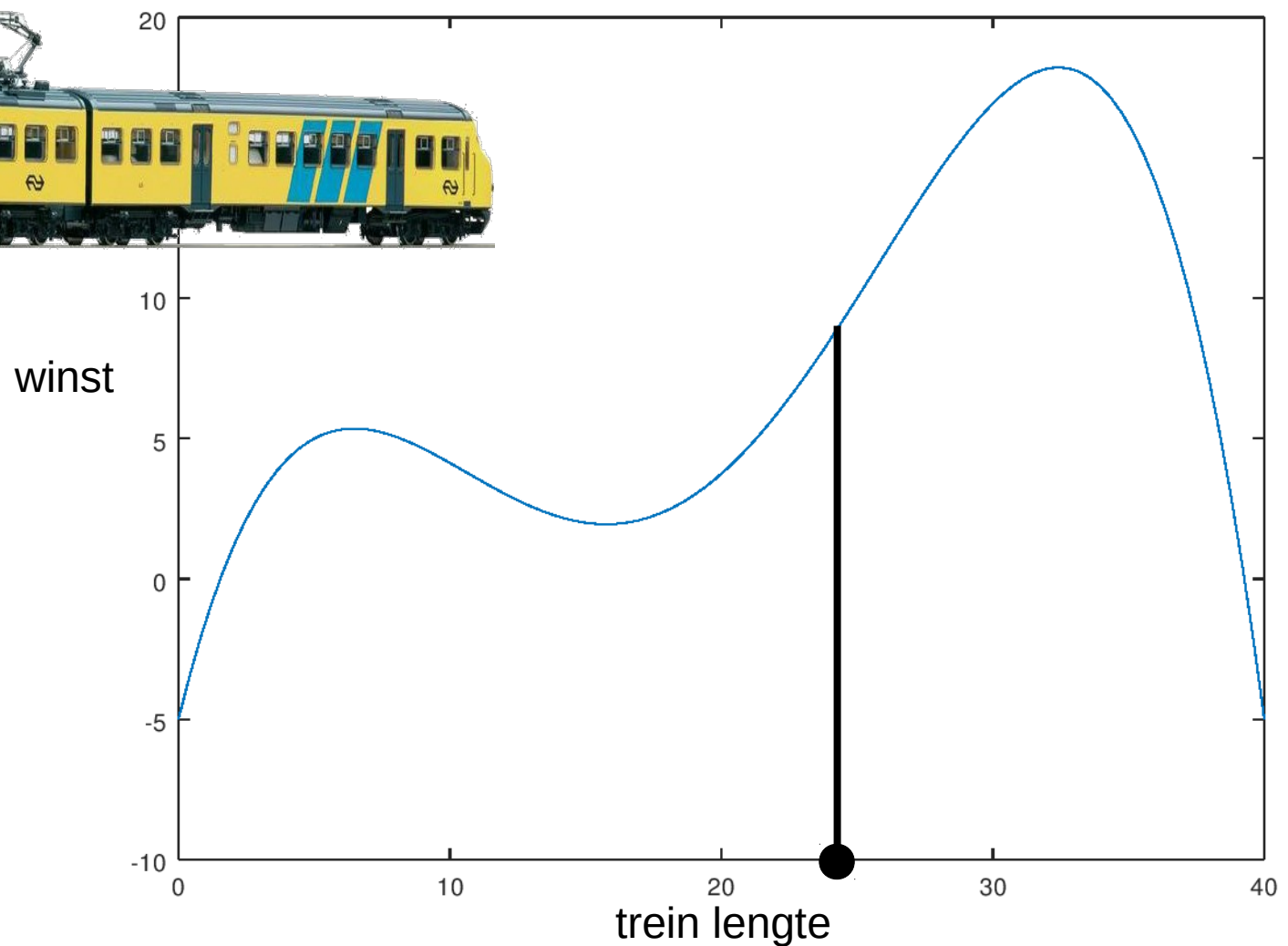
- 1 dimensionaal trein probleem





# Hill Climber

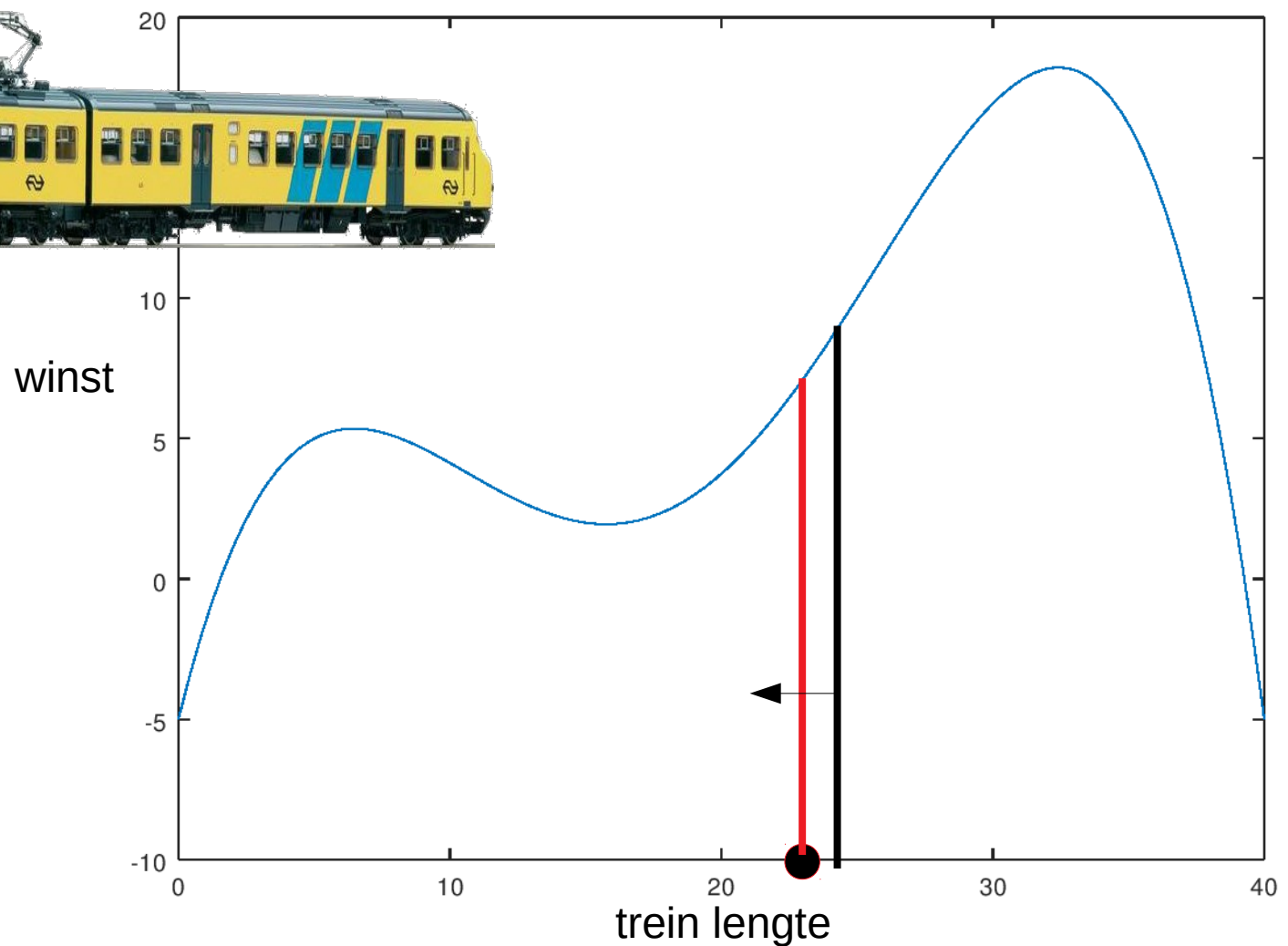
- 1 dimensionaal trein probleem





# Hill Climber

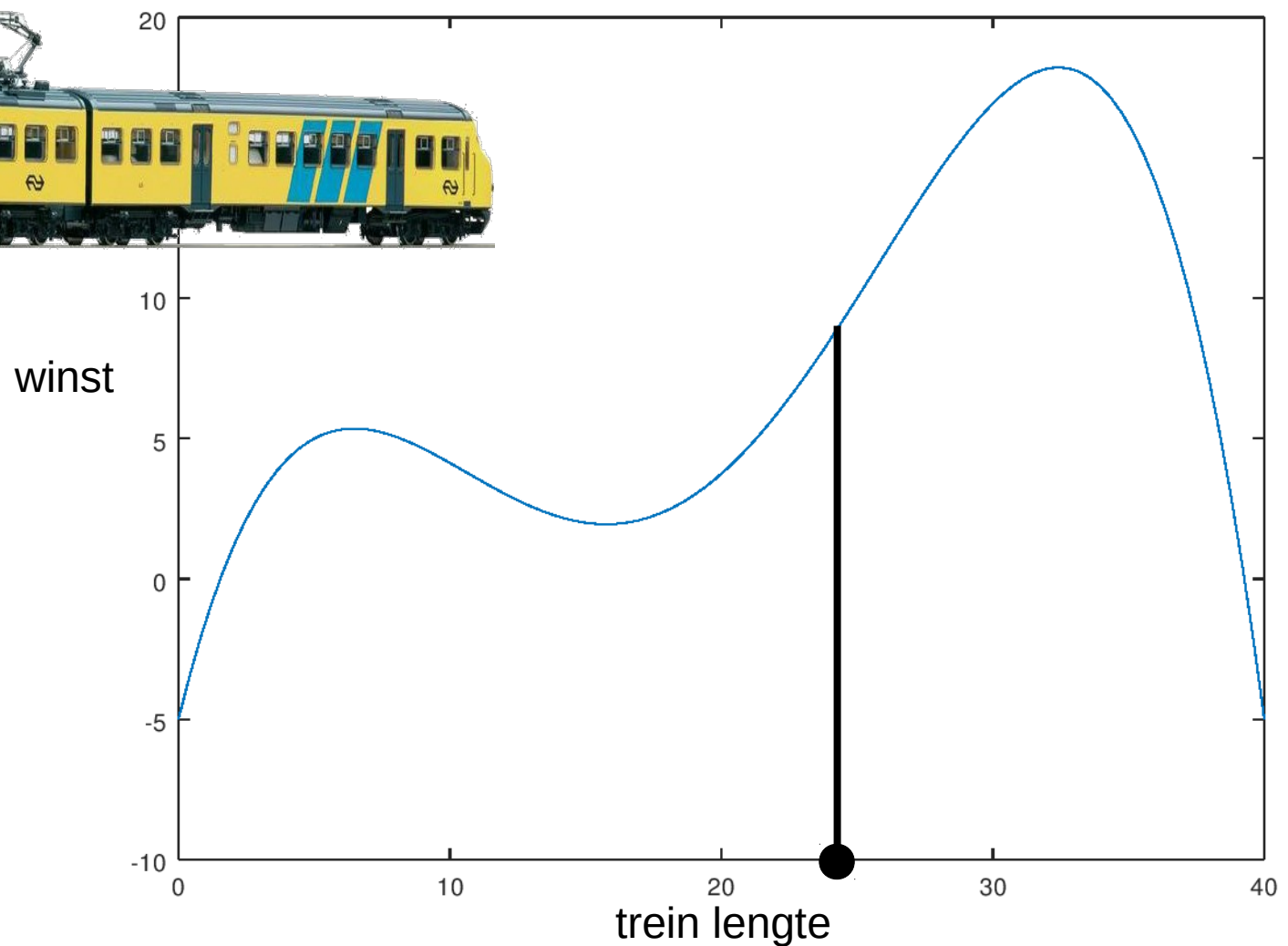
- 1 dimensionaal trein probleem





# Hill Climber

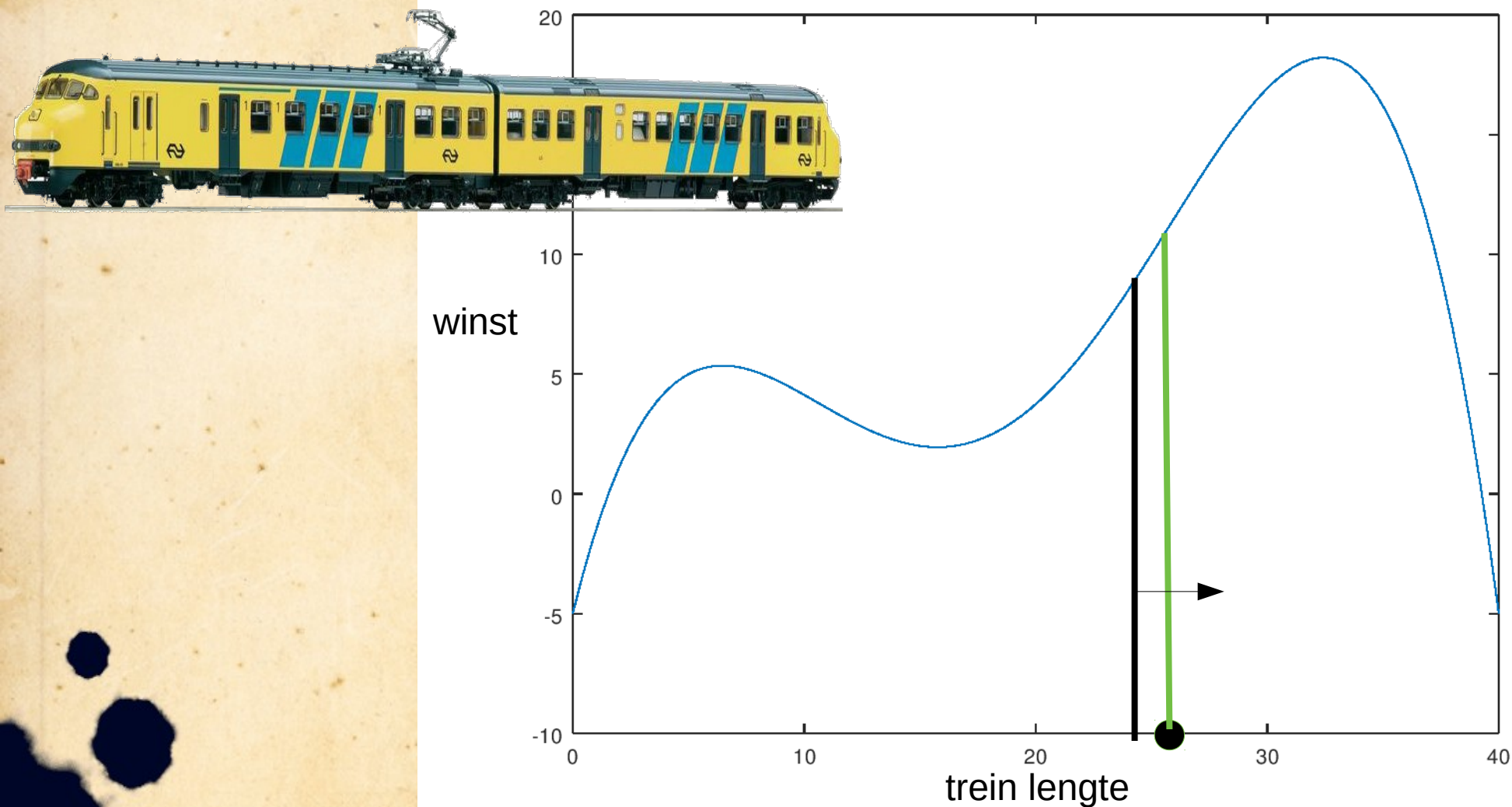
- 1 dimensionaal trein probleem





# Hill Climber

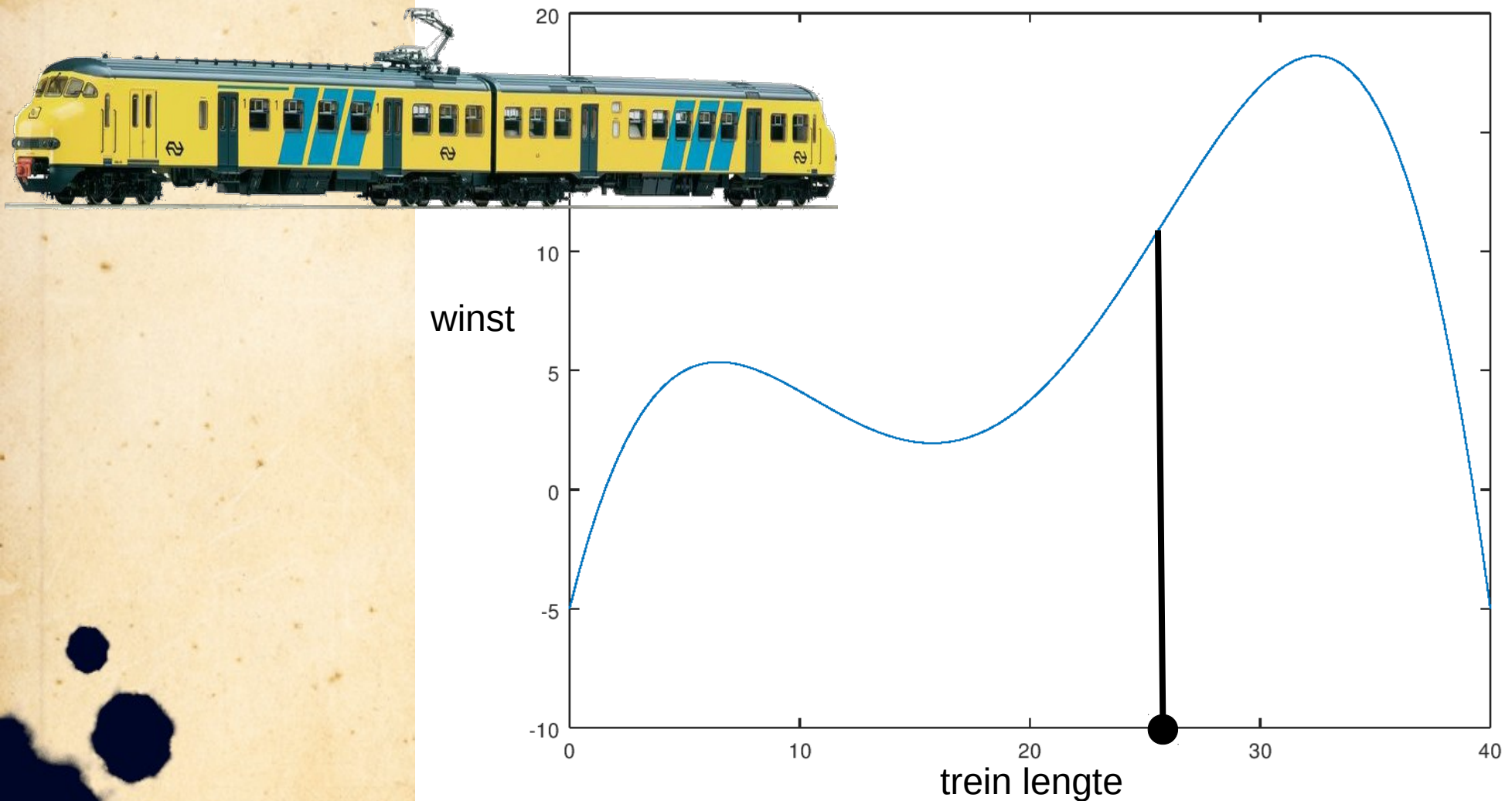
- 1 dimensionaal trein probleem





# Hill Climber

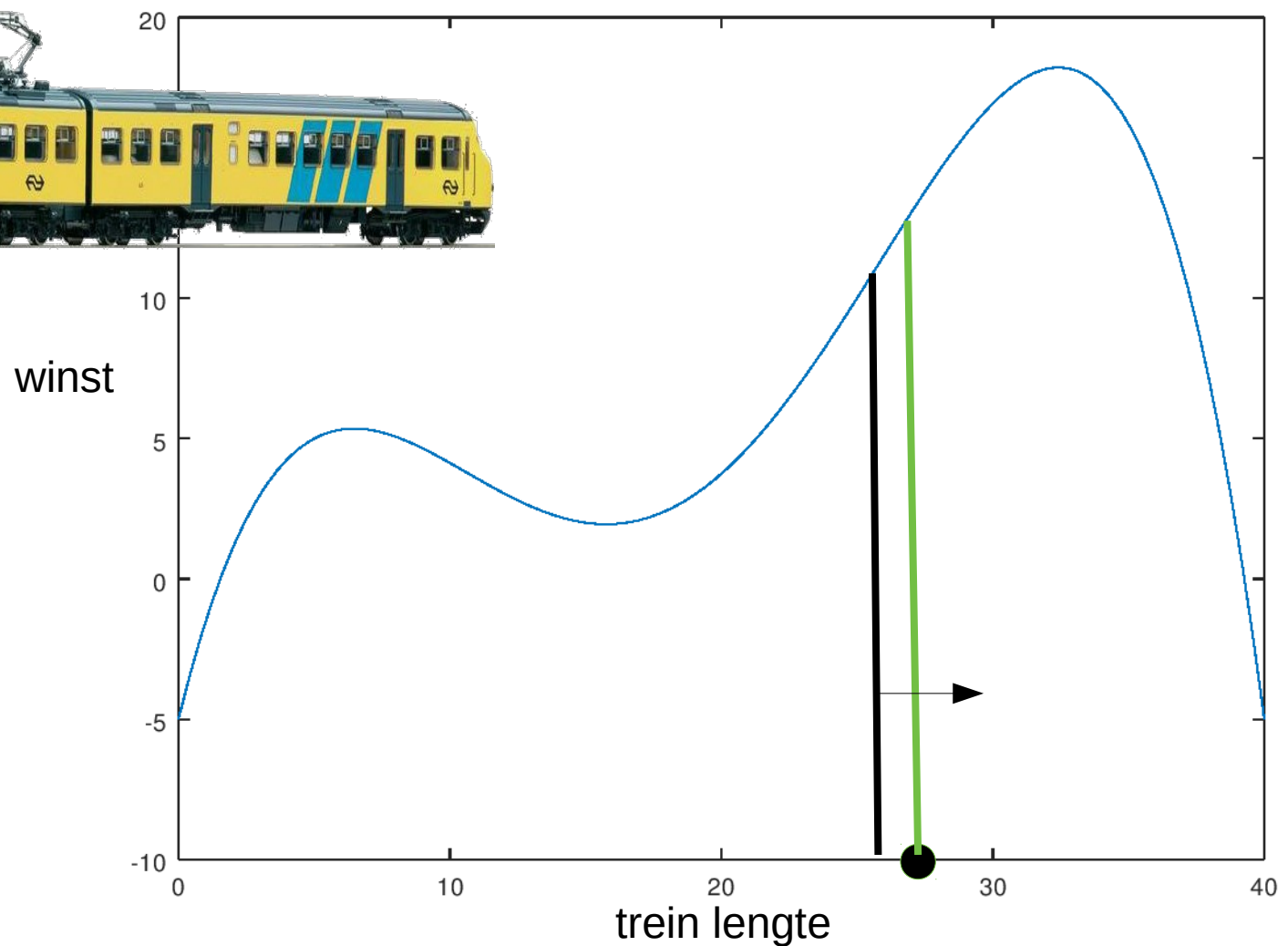
- 1 dimensionaal trein probleem





# Hill Climber

- 1 dimensionaal trein probleem

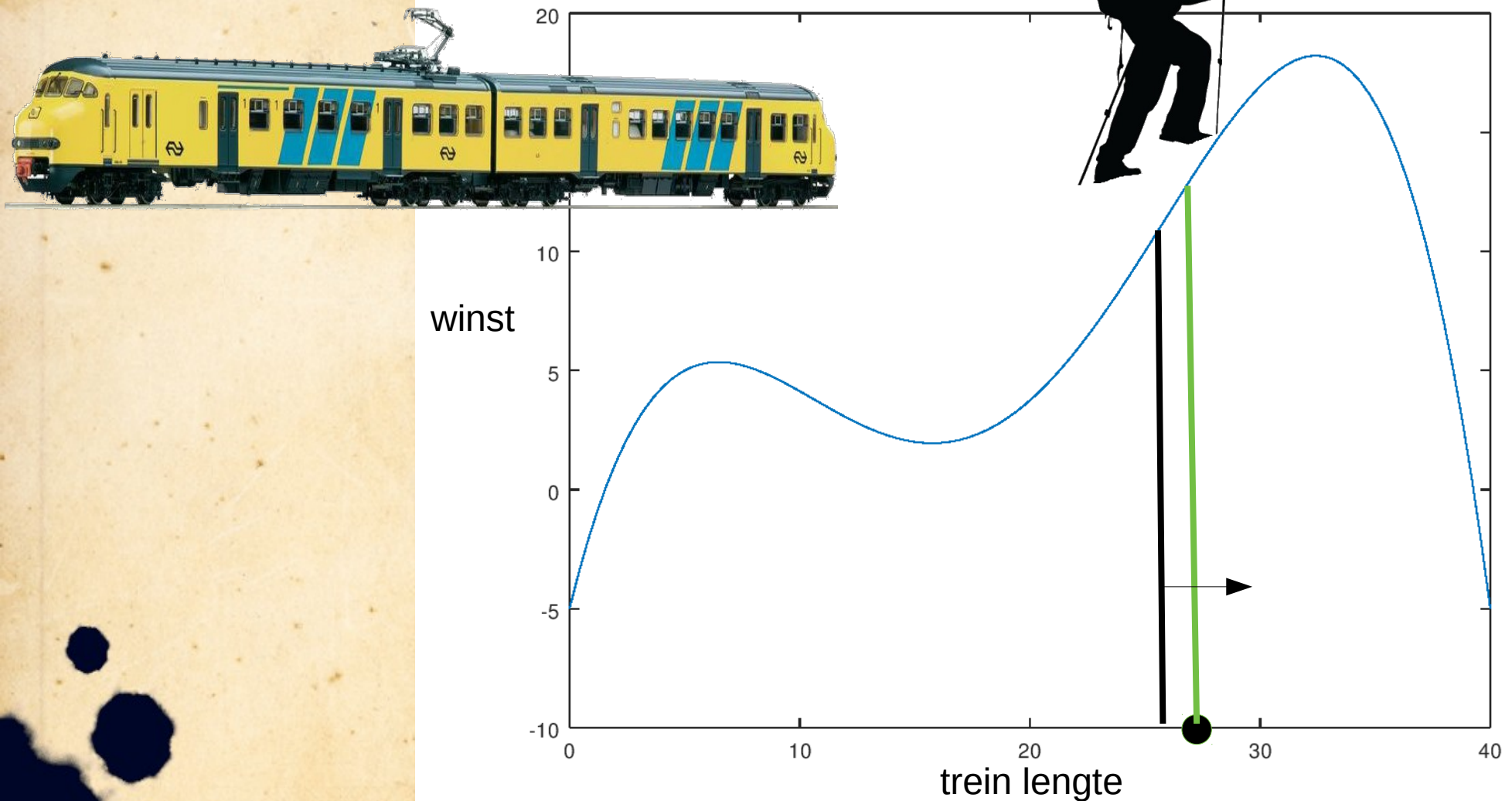




# Hill Climber

- 1 dimensionaal trein probleem

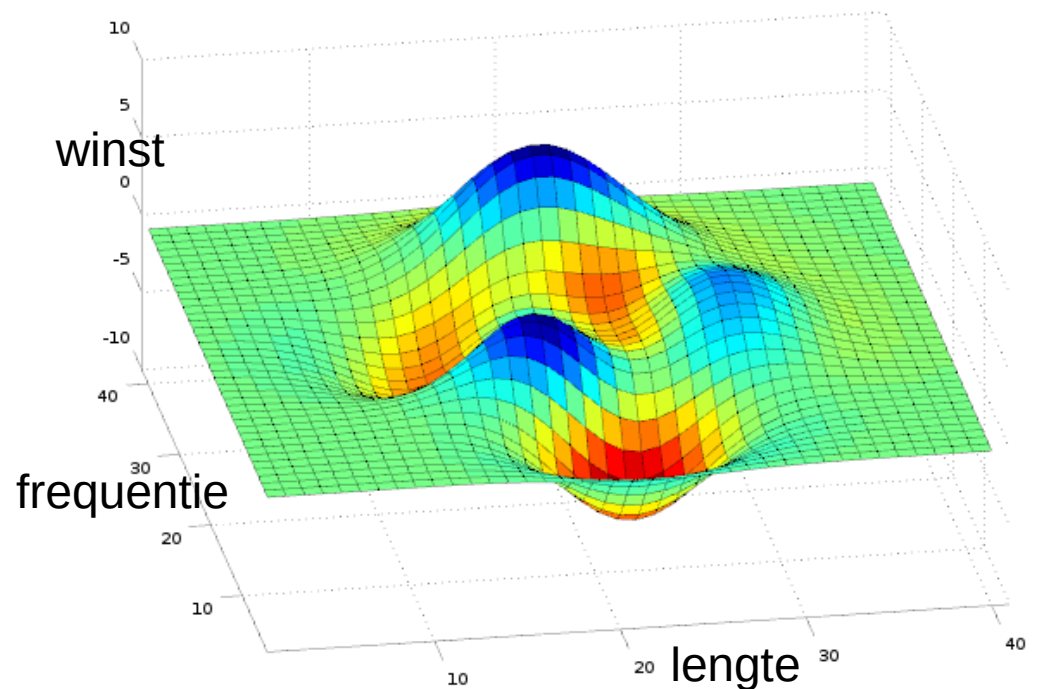
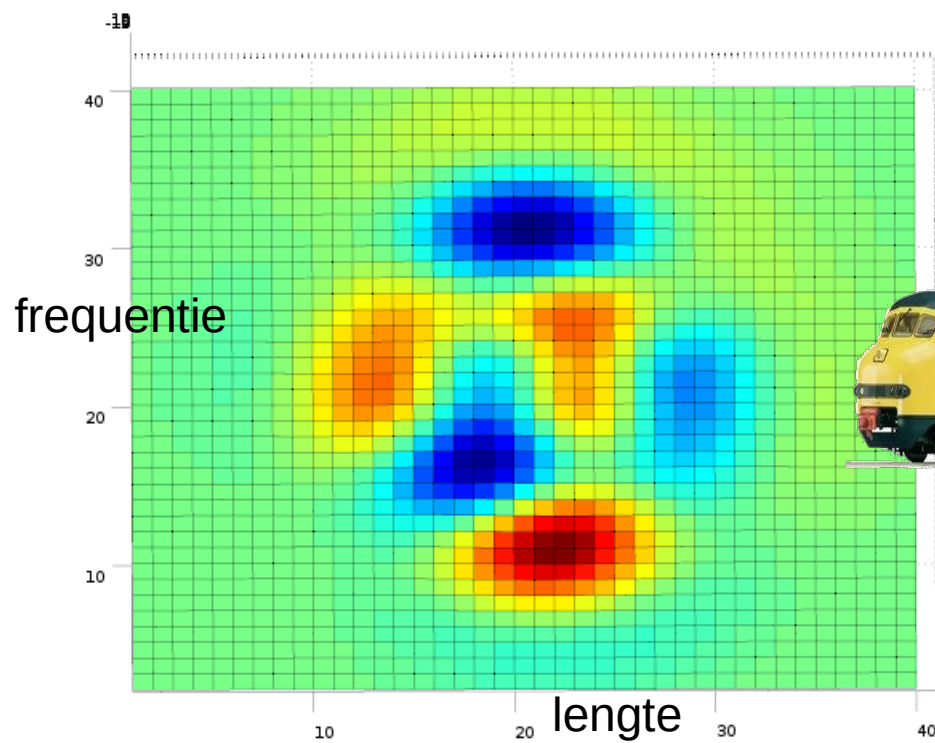
Gradient ascent





# Hill Climber

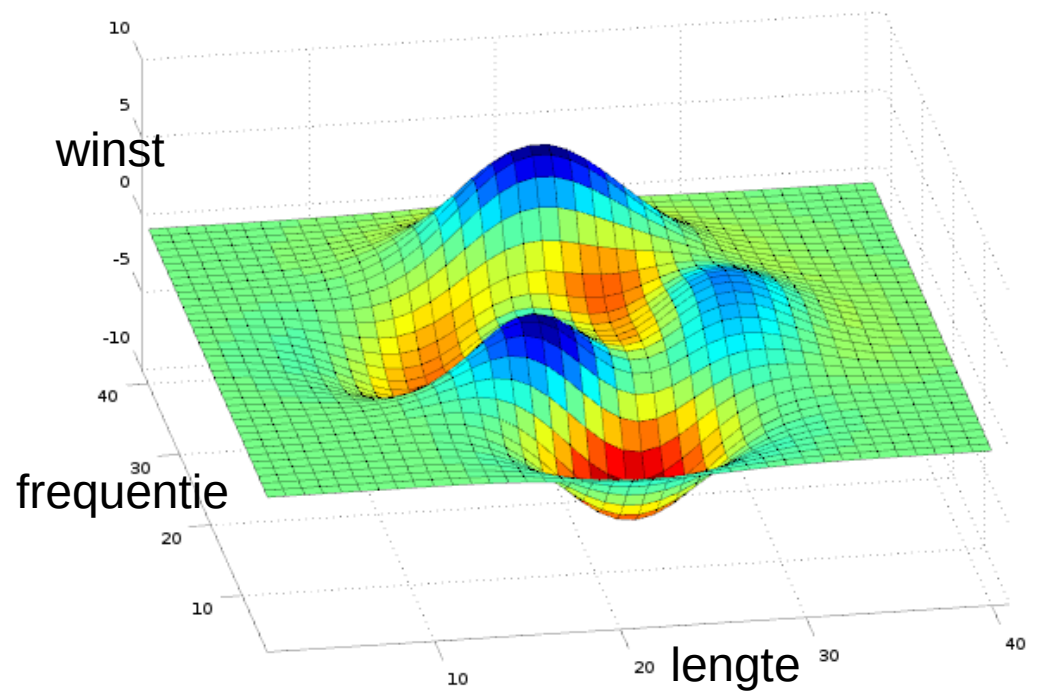
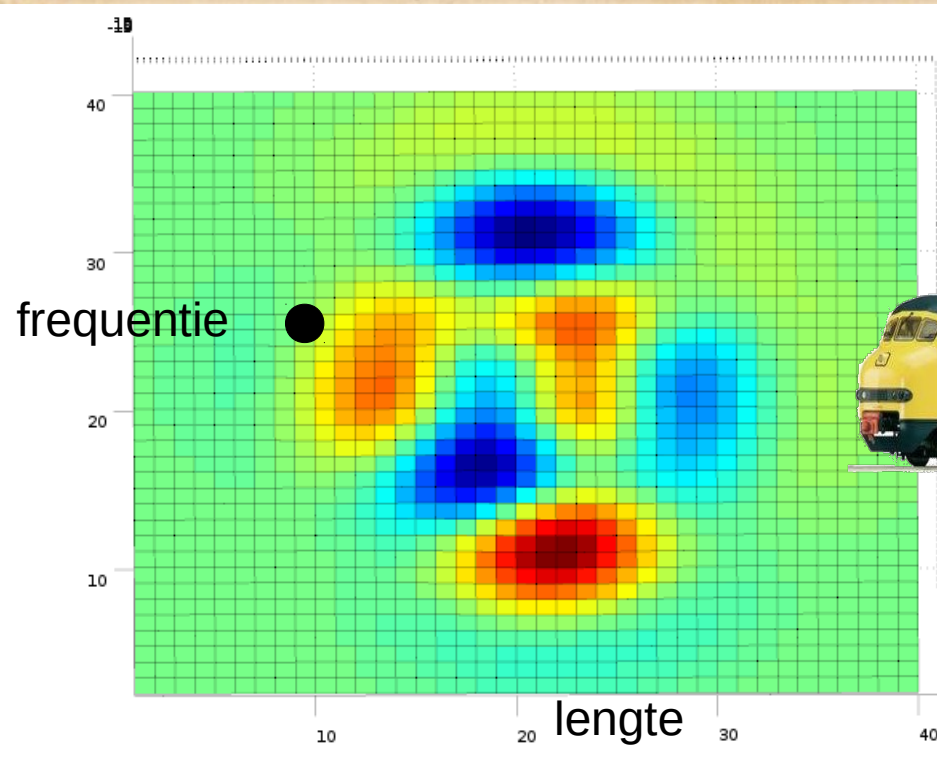
- 2 dimensionaal  
trein probleem





# Hill Climber

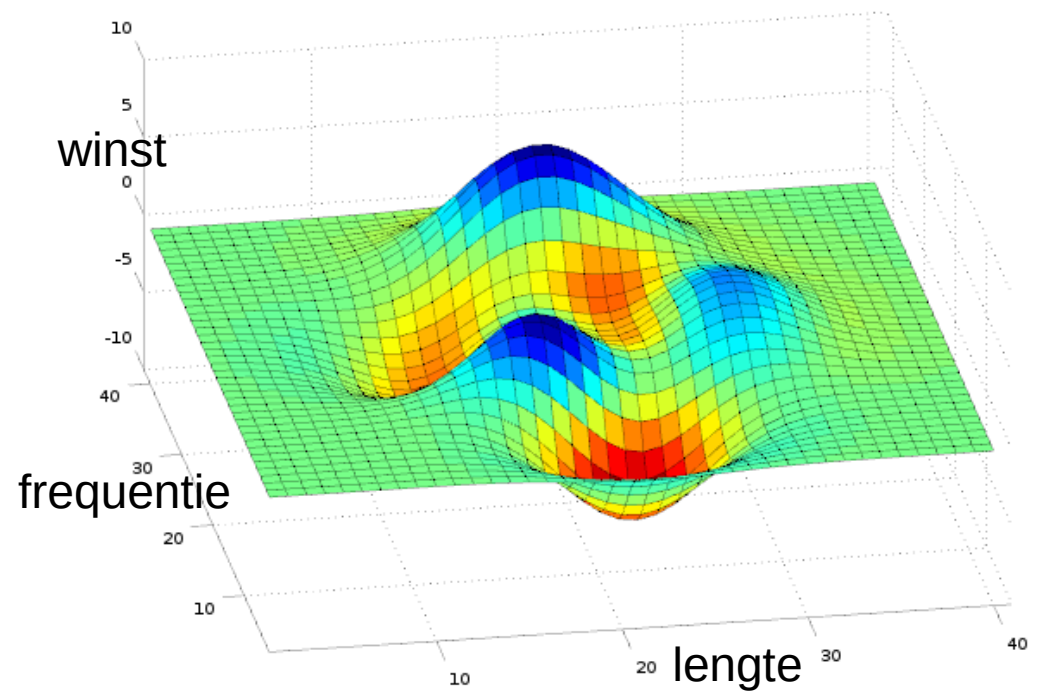
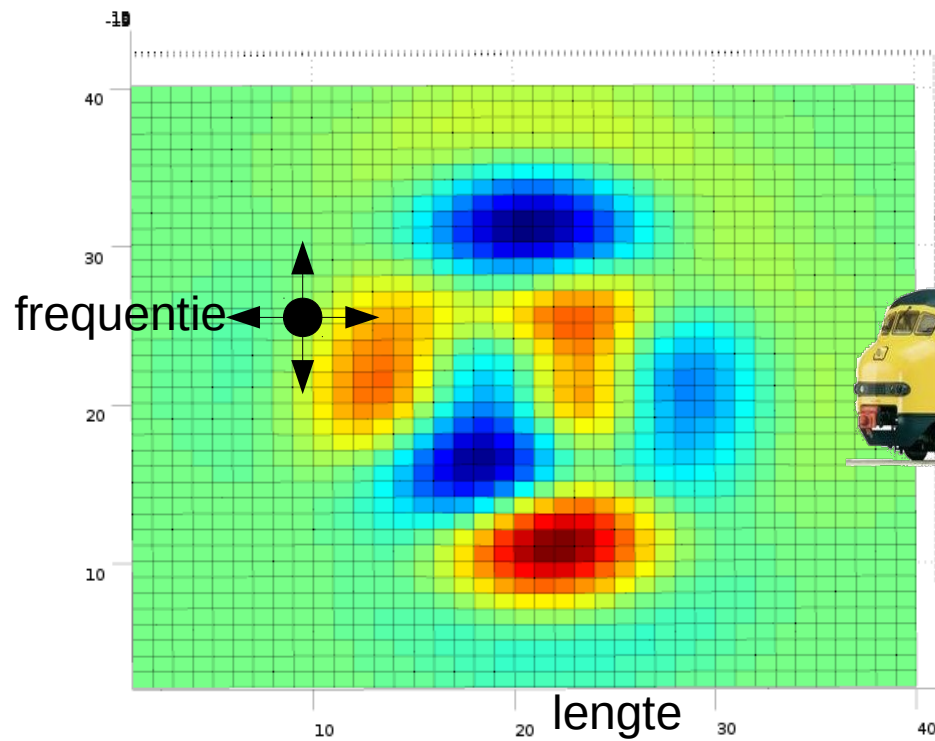
- 2 dimensionaal  
trein probleem





# Hill Climber

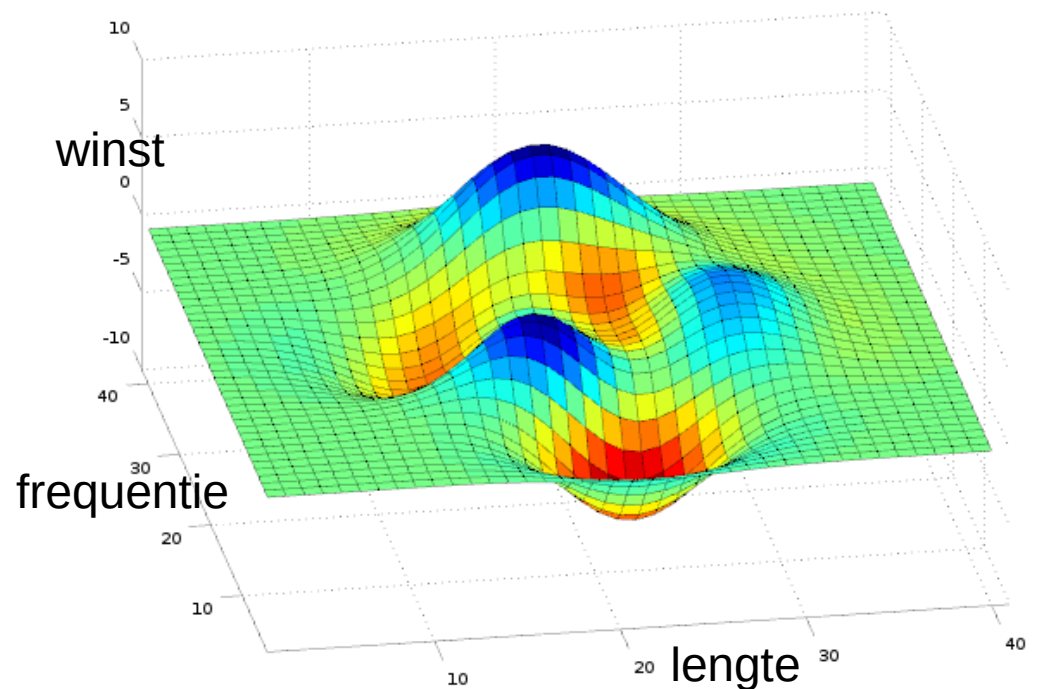
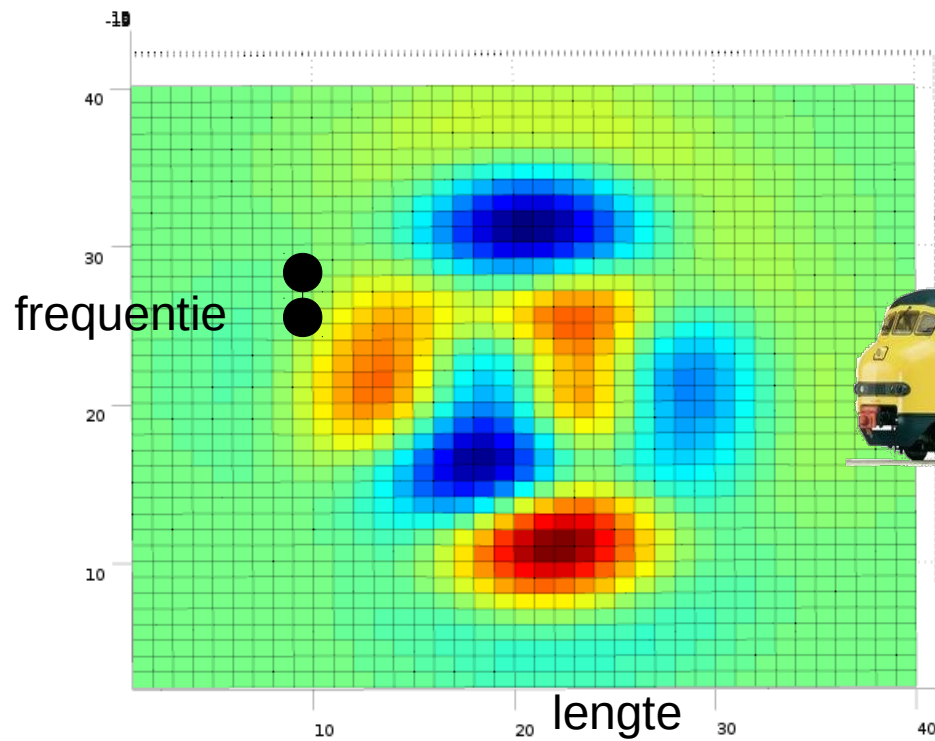
- 2 dimensionaal  
trein probleem





# Hill Climber

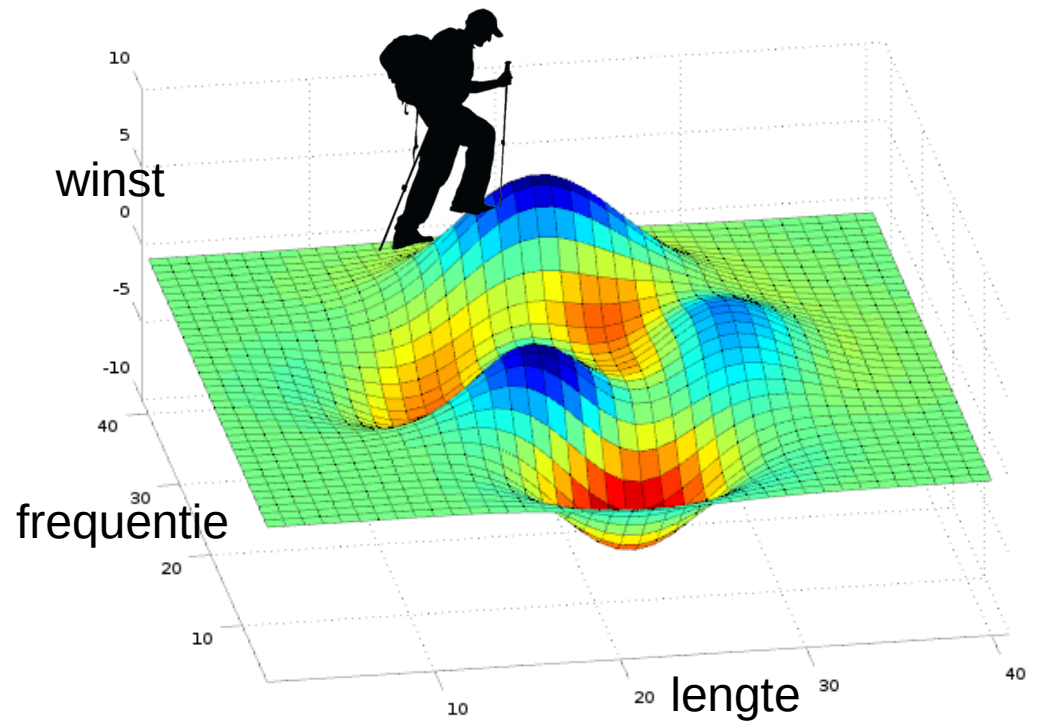
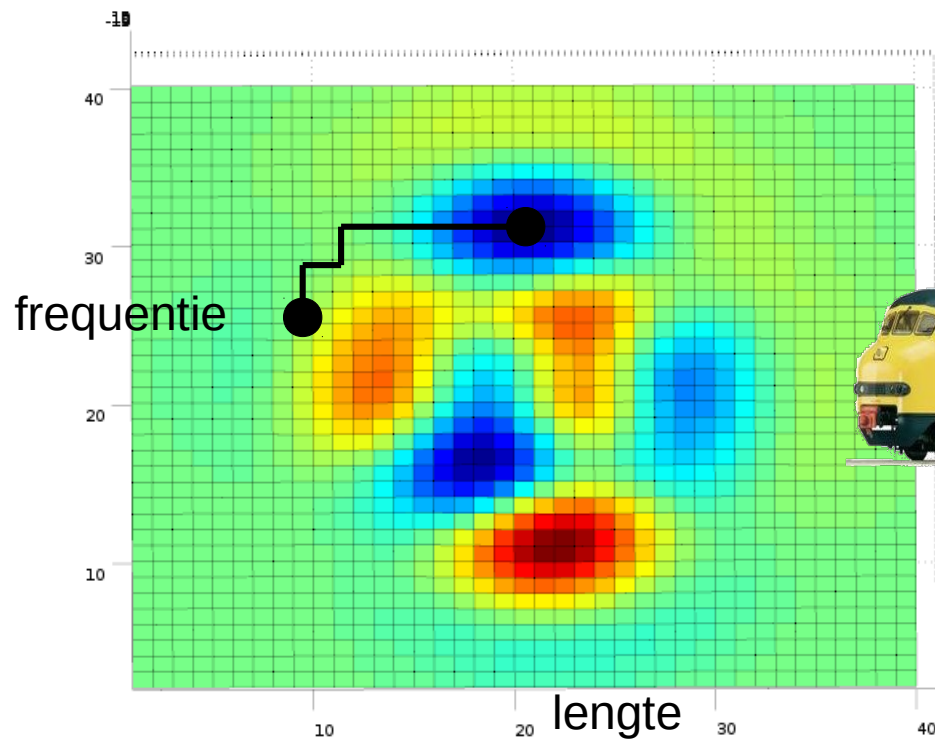
- 2 dimensionaal  
trein probleem





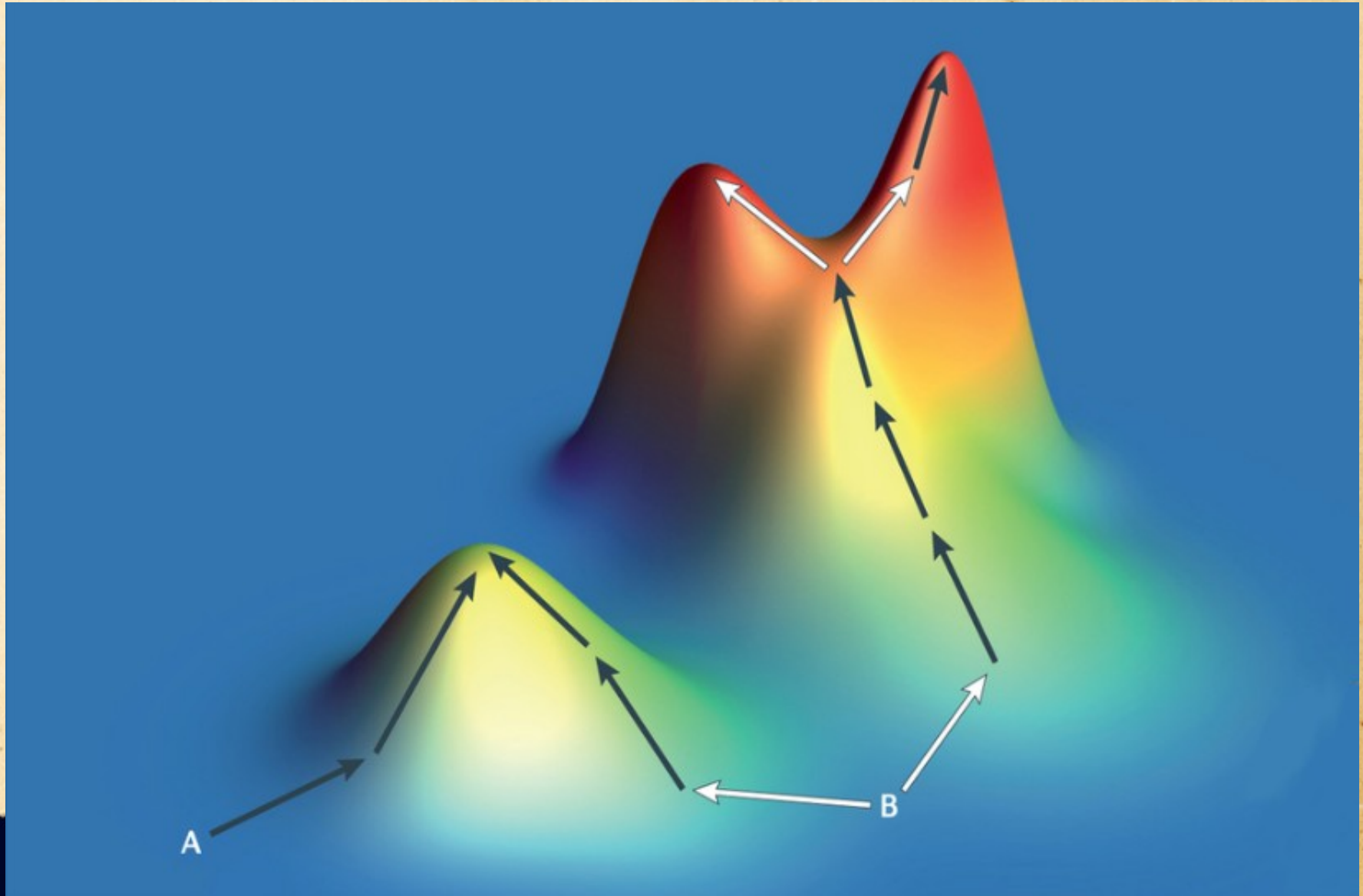
# Hill Climber

- 2 dimensionaal  
trein probleem



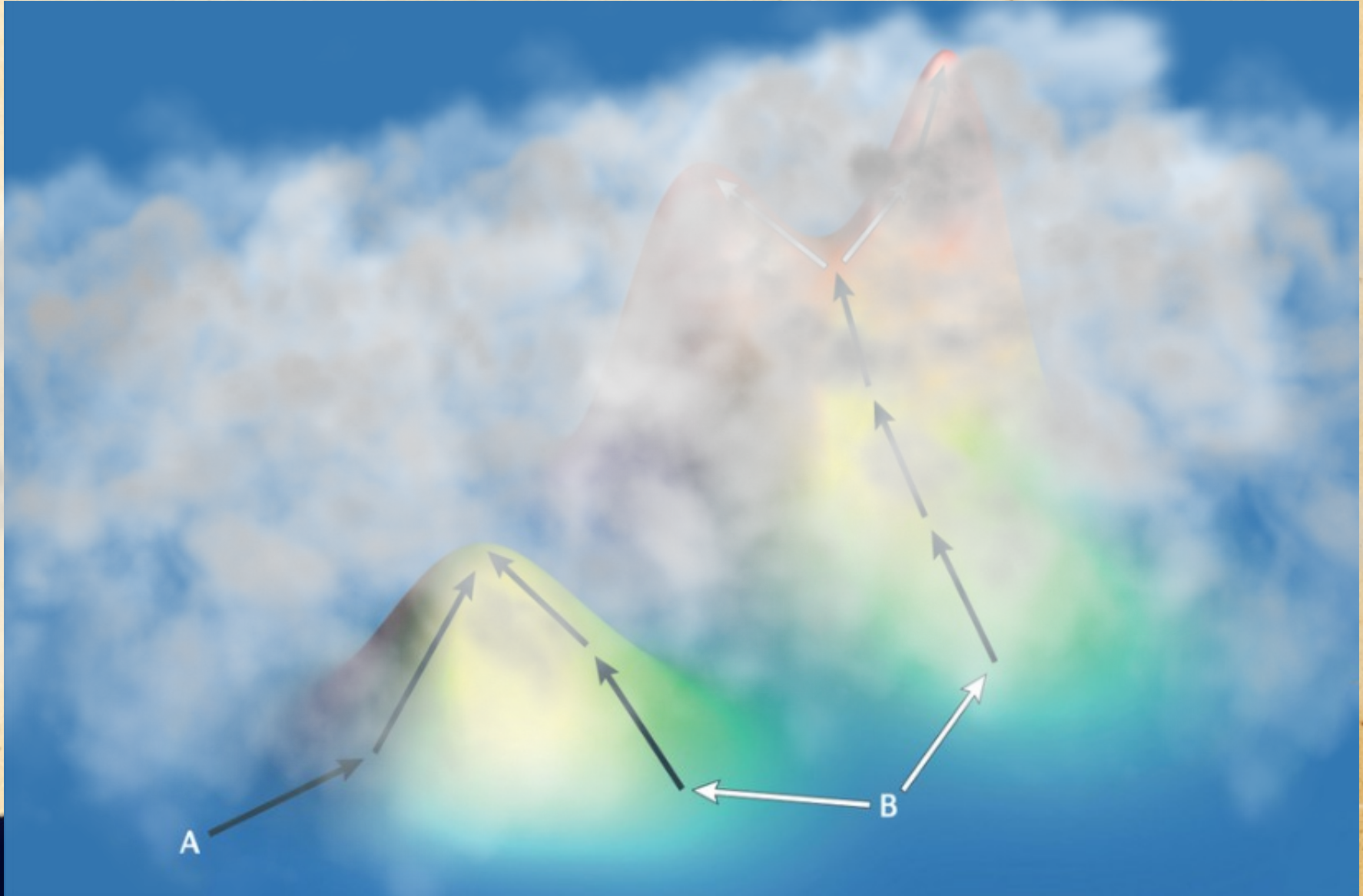


# Hill Climber



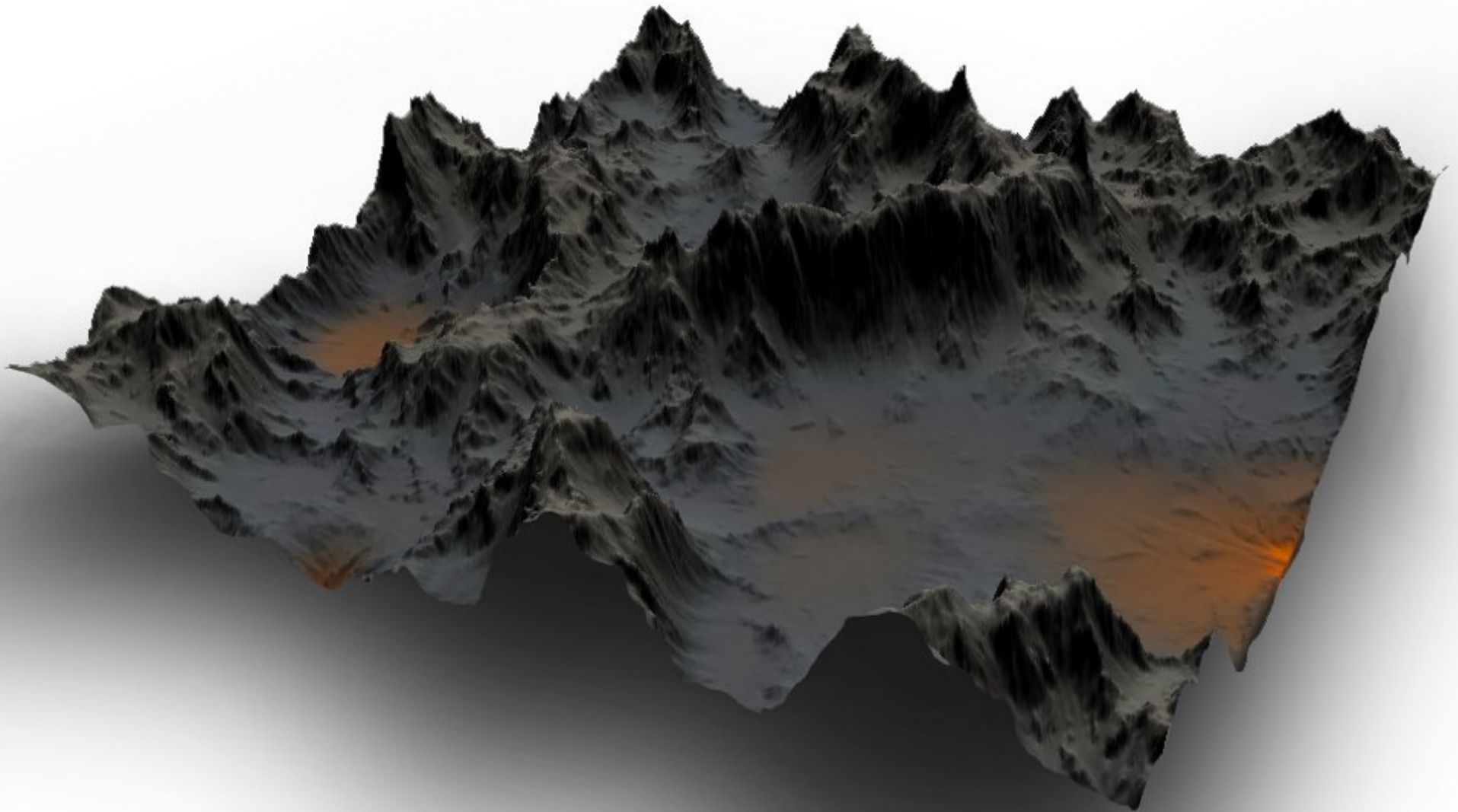


# Hill Climber





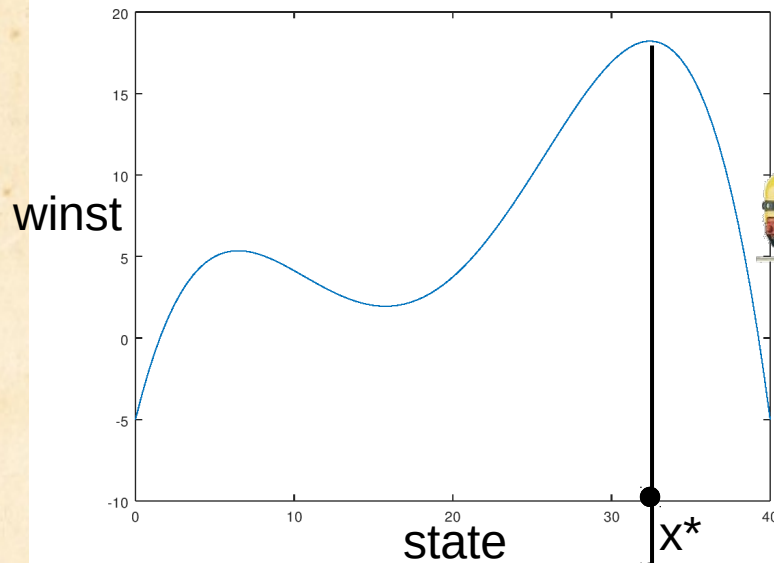
# Hill Climber



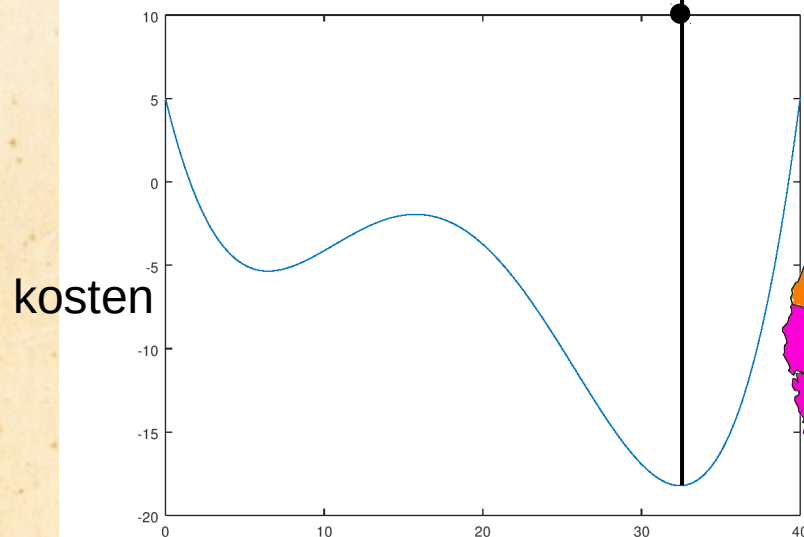
Groot, complex en hoog dimensionaal landschap



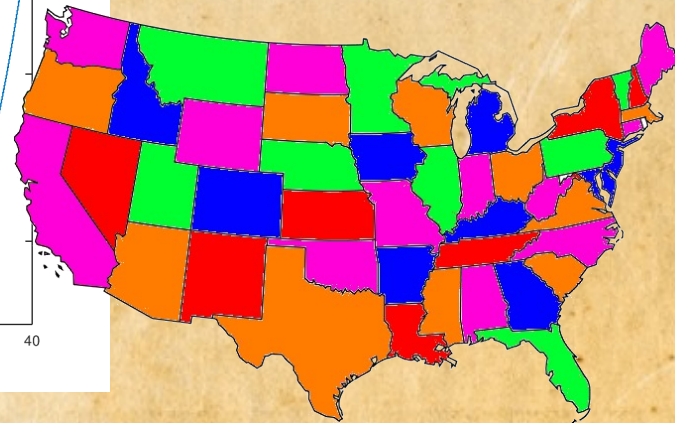
# Maximaliseren of Minimaliseren



$$x^* = \arg \max_x f(x)$$

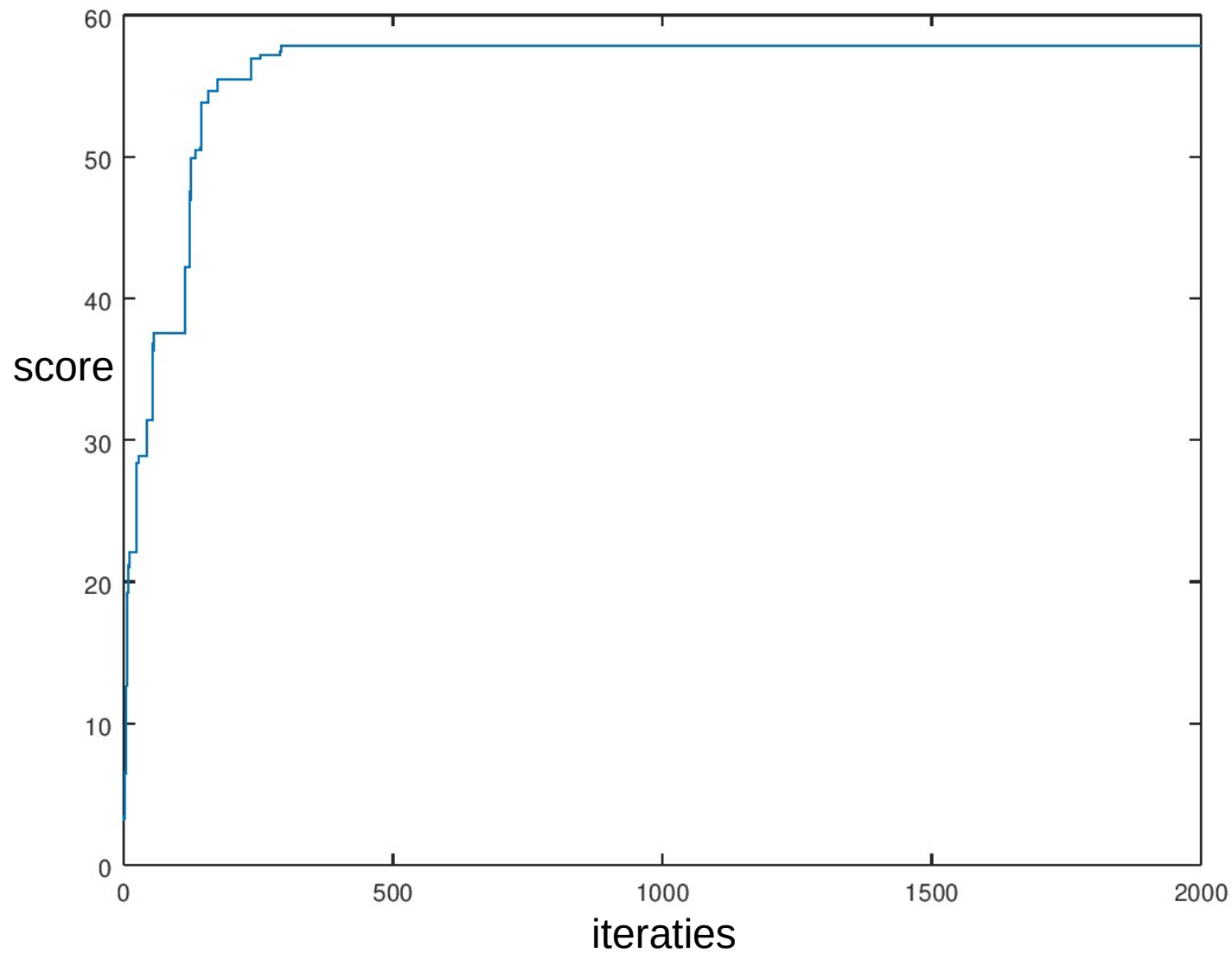


$$x^* = \arg \min_x -f(x)$$



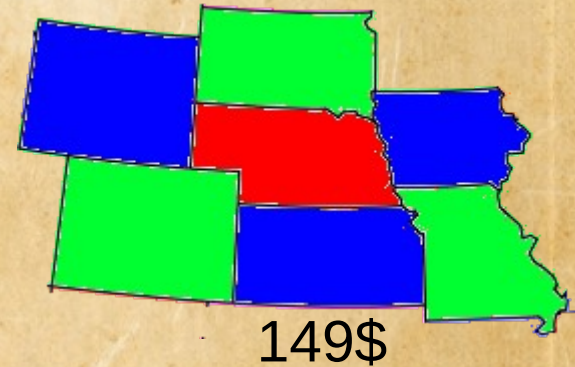
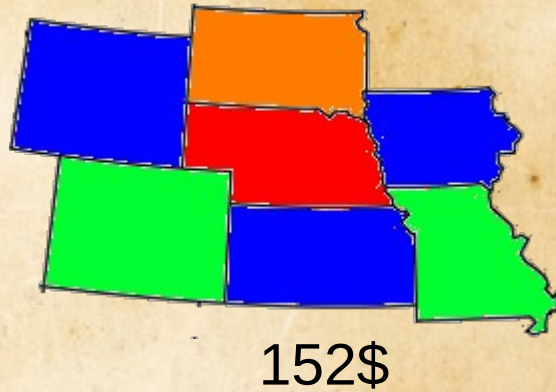
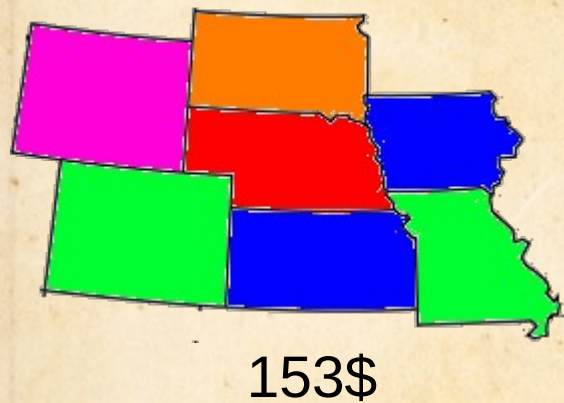


# Convergentie





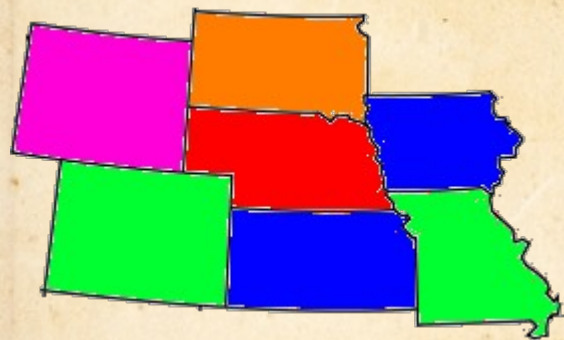
# Convergentie



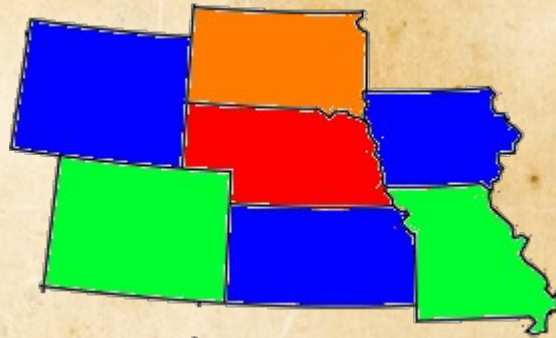
Lokaal optimum?  
Globaal optimum?



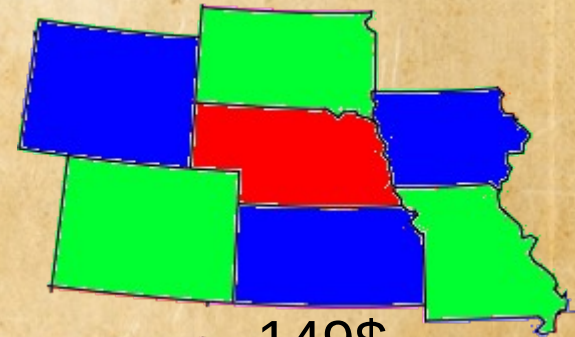
# Convergentie



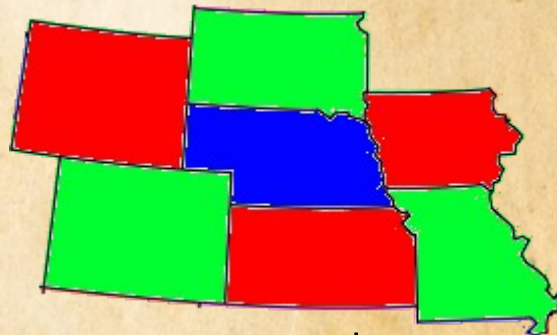
153\$



152\$



149\$



145\$

Lokaal optimum? ✓  
Globaal optimum? ✗



# Restart Hill Climber, pseudo code

Kies een random start state

Herhaal **tot na N-keer niet meer verbetert**:

    Doe een kleine random aanpassing

    Als de state is verslechterd:

        Maak de aanpassing ongedaan

# Restart Hill Climber, pseudo code

## **Herhaal:**

Kies een random start state

**Herhaal tot na N-keer niet meer verbetert:**

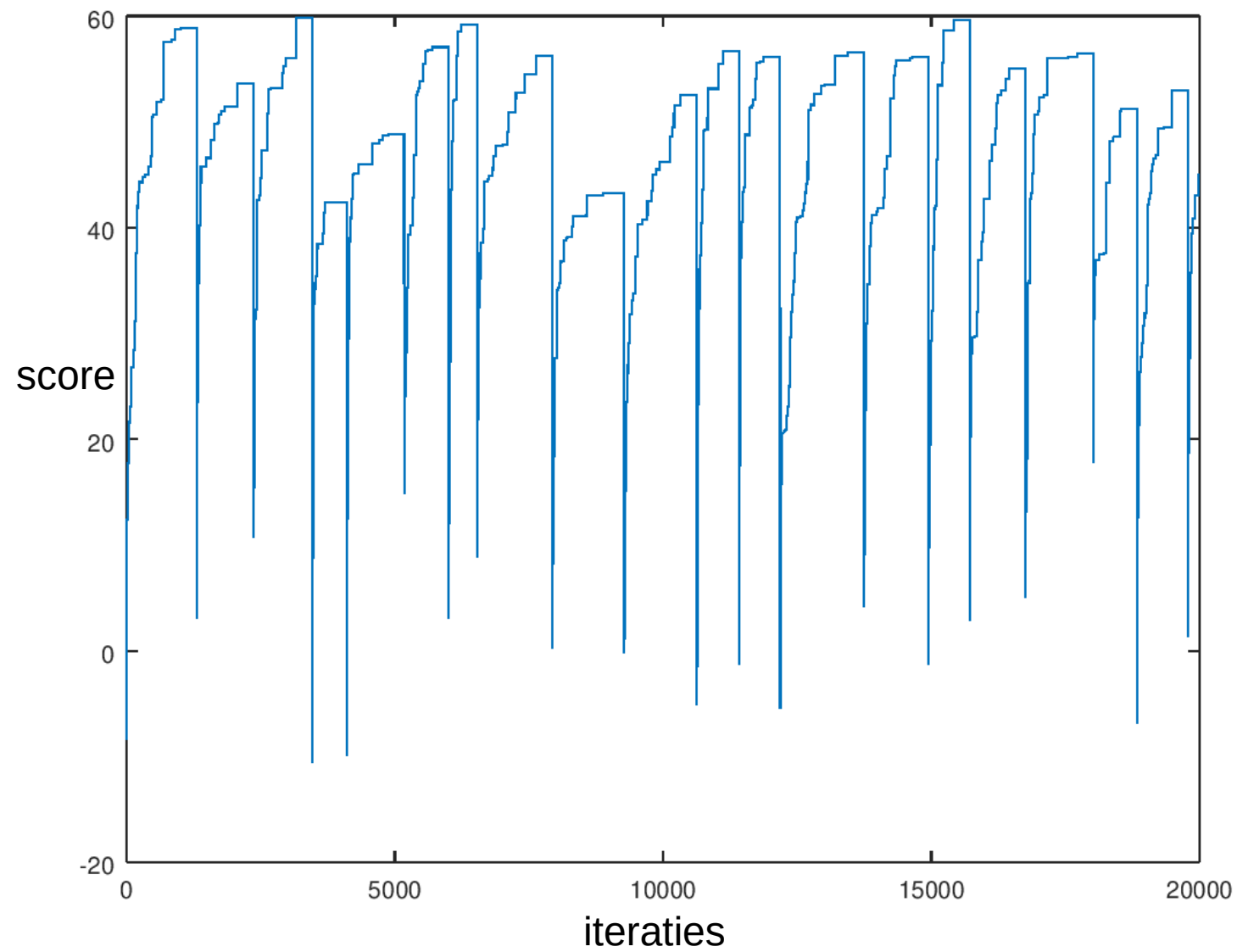
Doe een kleine random aanpassing

Als de state is verslechterd:

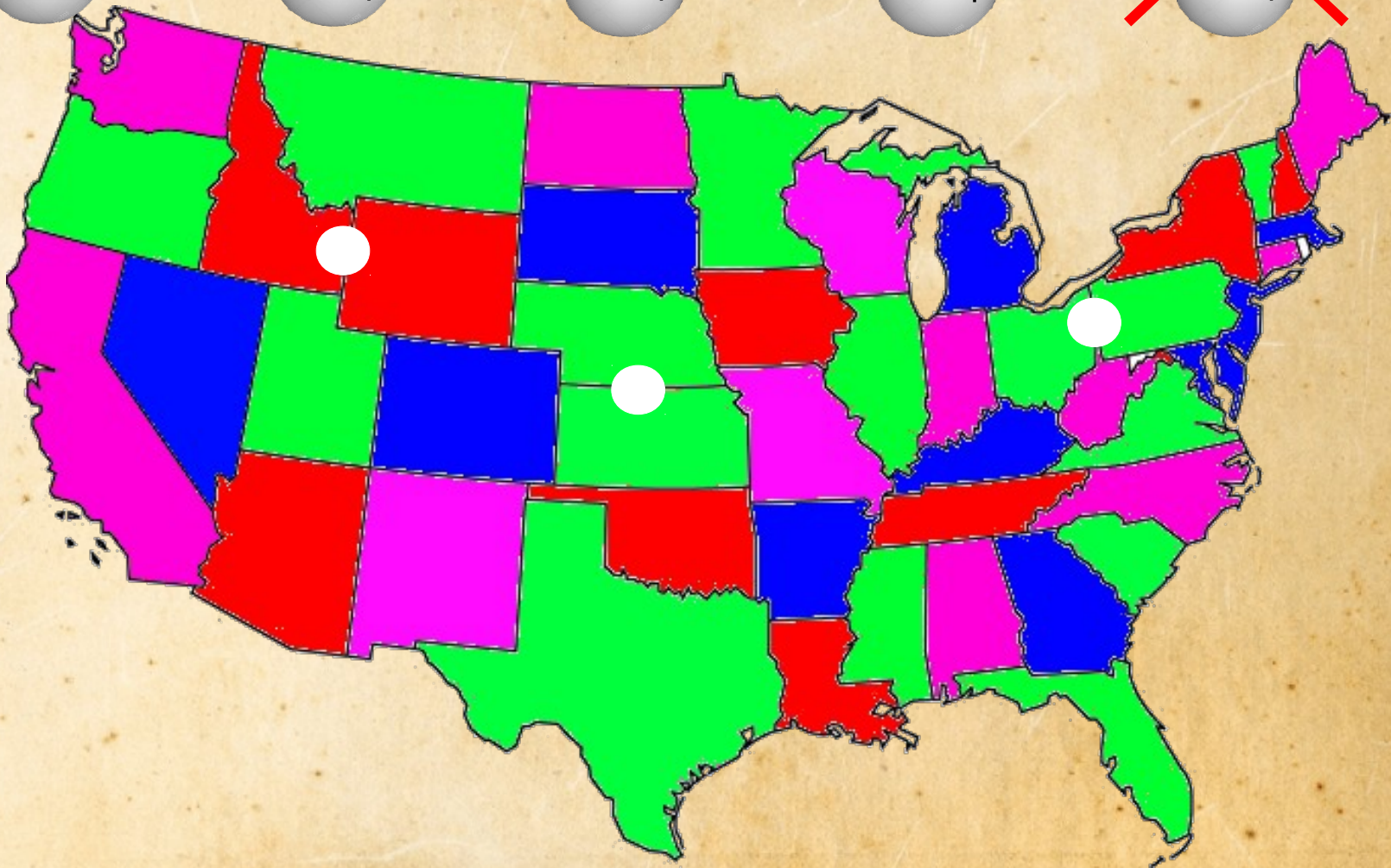
Maak de aanpassing ongedaan



# Restart

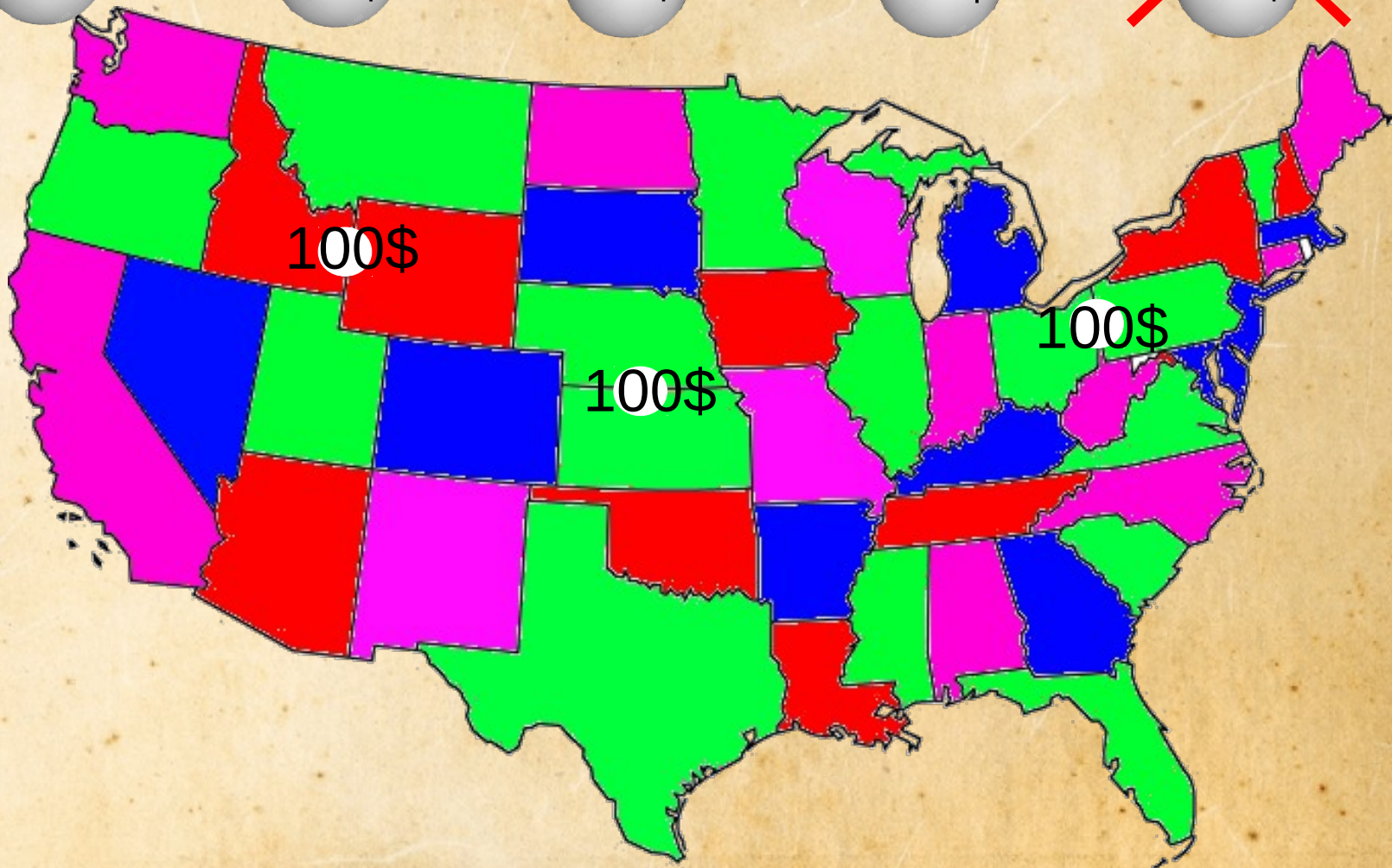


# Constraint relaxation





# Constraint relaxation





# Constraint relaxation, pseudo code

Herhaal:

Kies een random start state **zonder constraint checks**

Herhaal totdat N-keer niet meer verbetert:

Doe een kleine random aanpassing

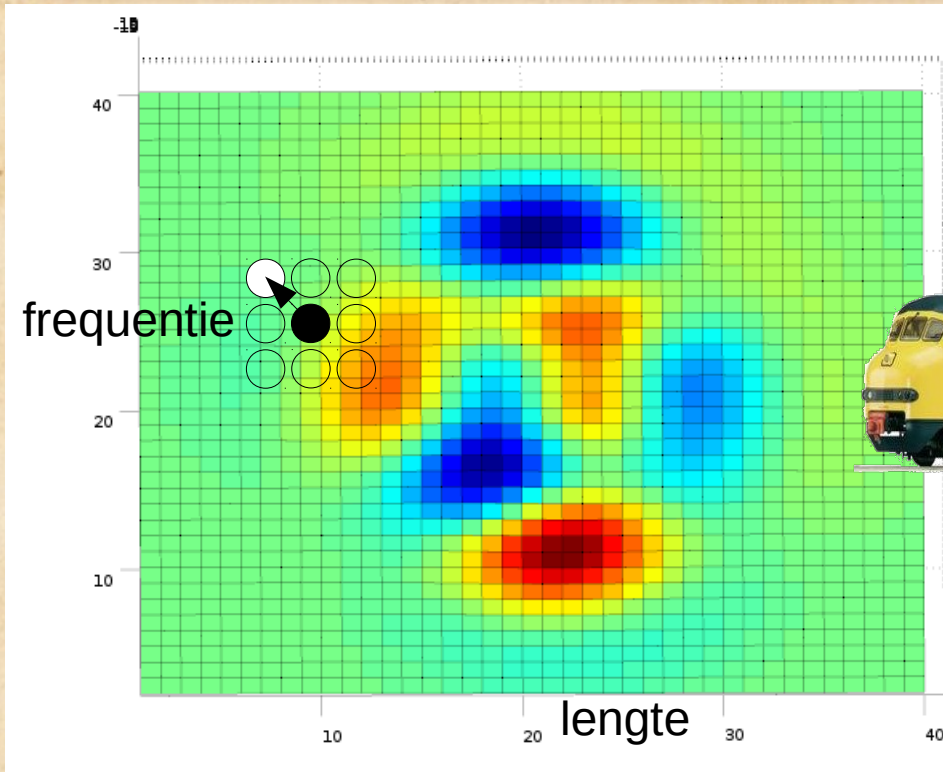
**Geef strafpunten voor constraint-schendingen**

Als de state is verslechterd:

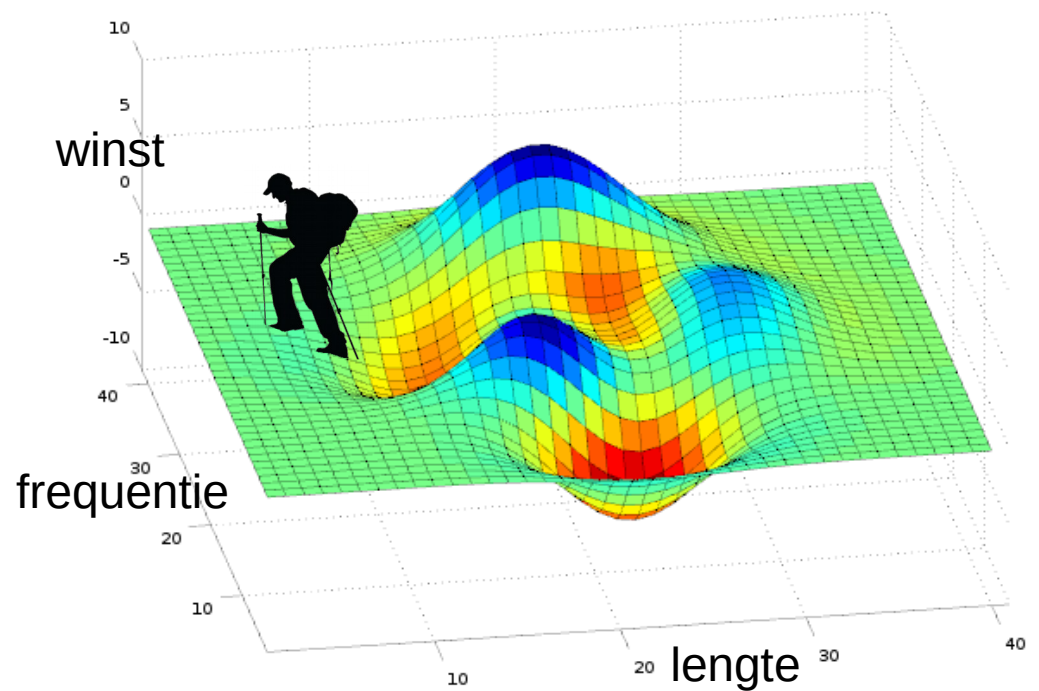
Maak de aanpassing ongedaan



# Steepest Ascent Hill Climber



Random step: **Stochastic Hill Climber**  
Best of all steps: **Steepest Ascent Hill Climber**





# Steepest Ascent Hill Climber

Herhaal:

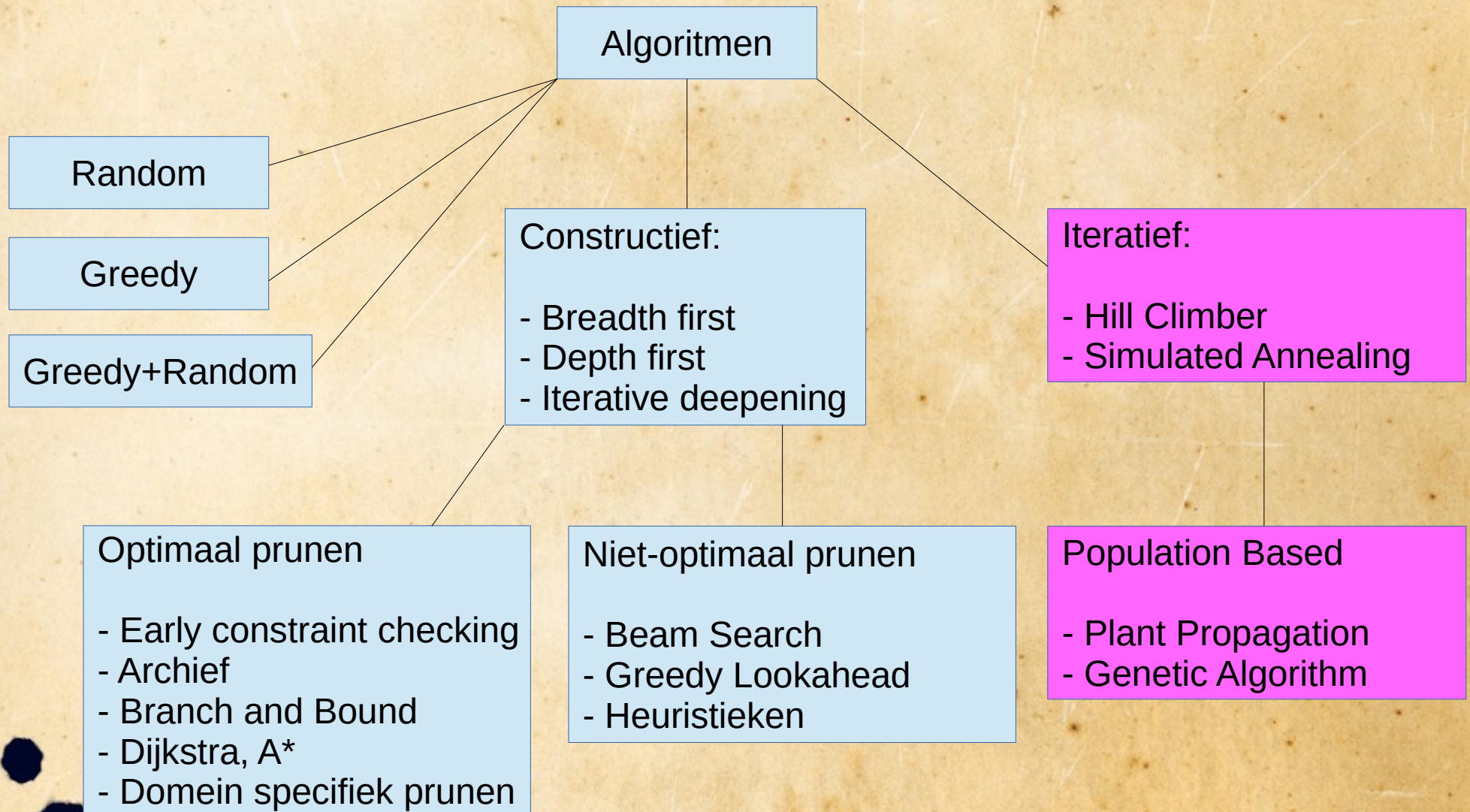
Kies een random start state

Herhaal **totdat niet meer** verbetert:

Doe **beste van alle mogelijke kleine aanpassingen**

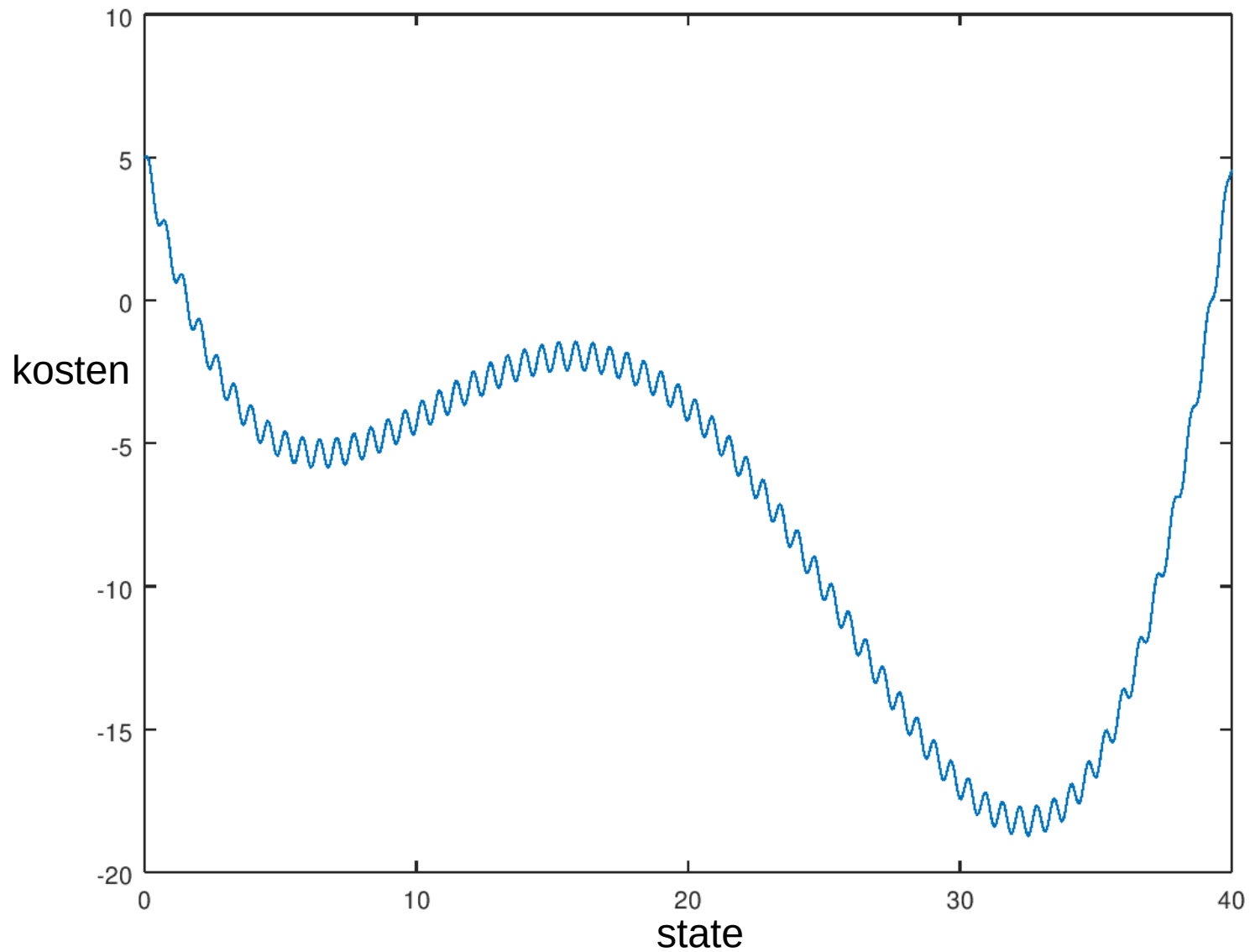


# Algoritmen





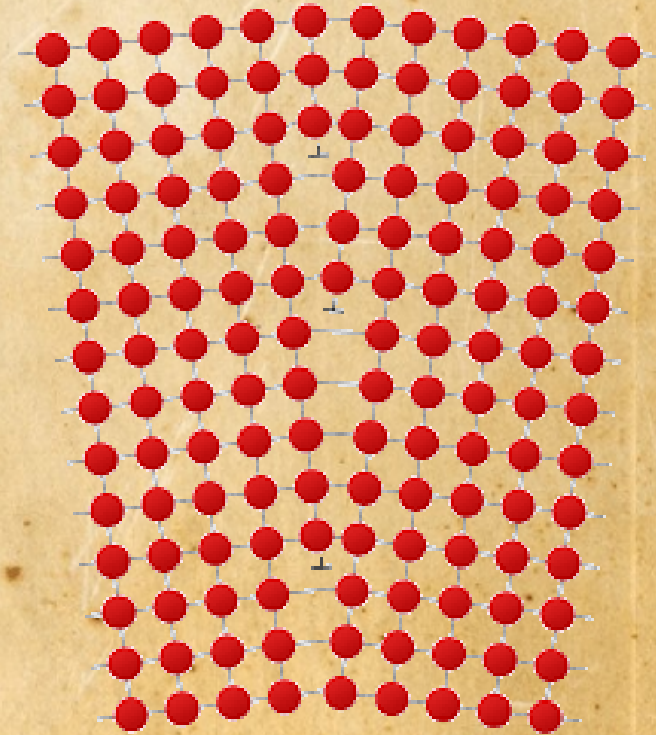
# Simulated Annealing





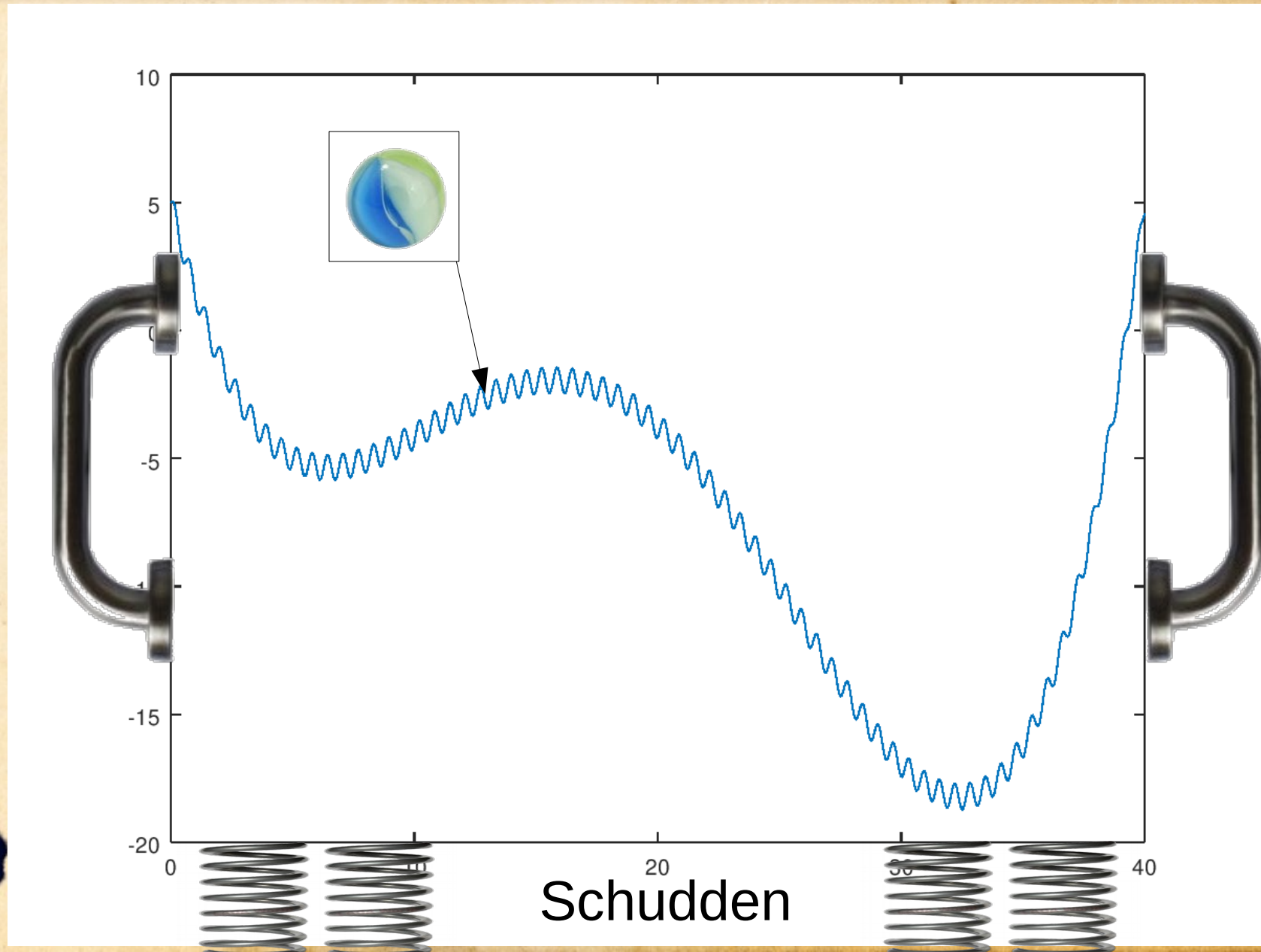
# Metaal Annealing

- Kristalstructuur van metaal versterken





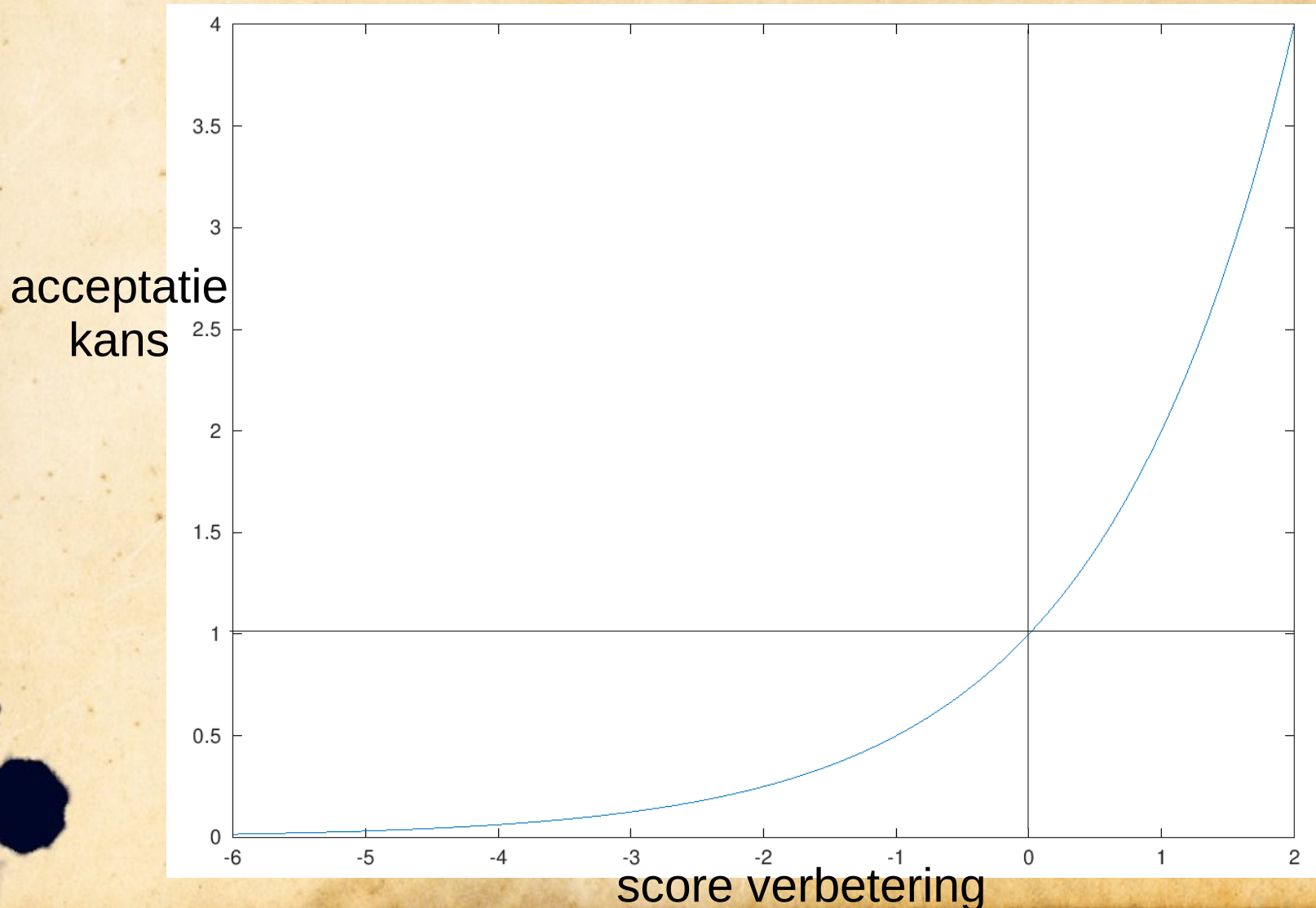
# Simulated Annealing





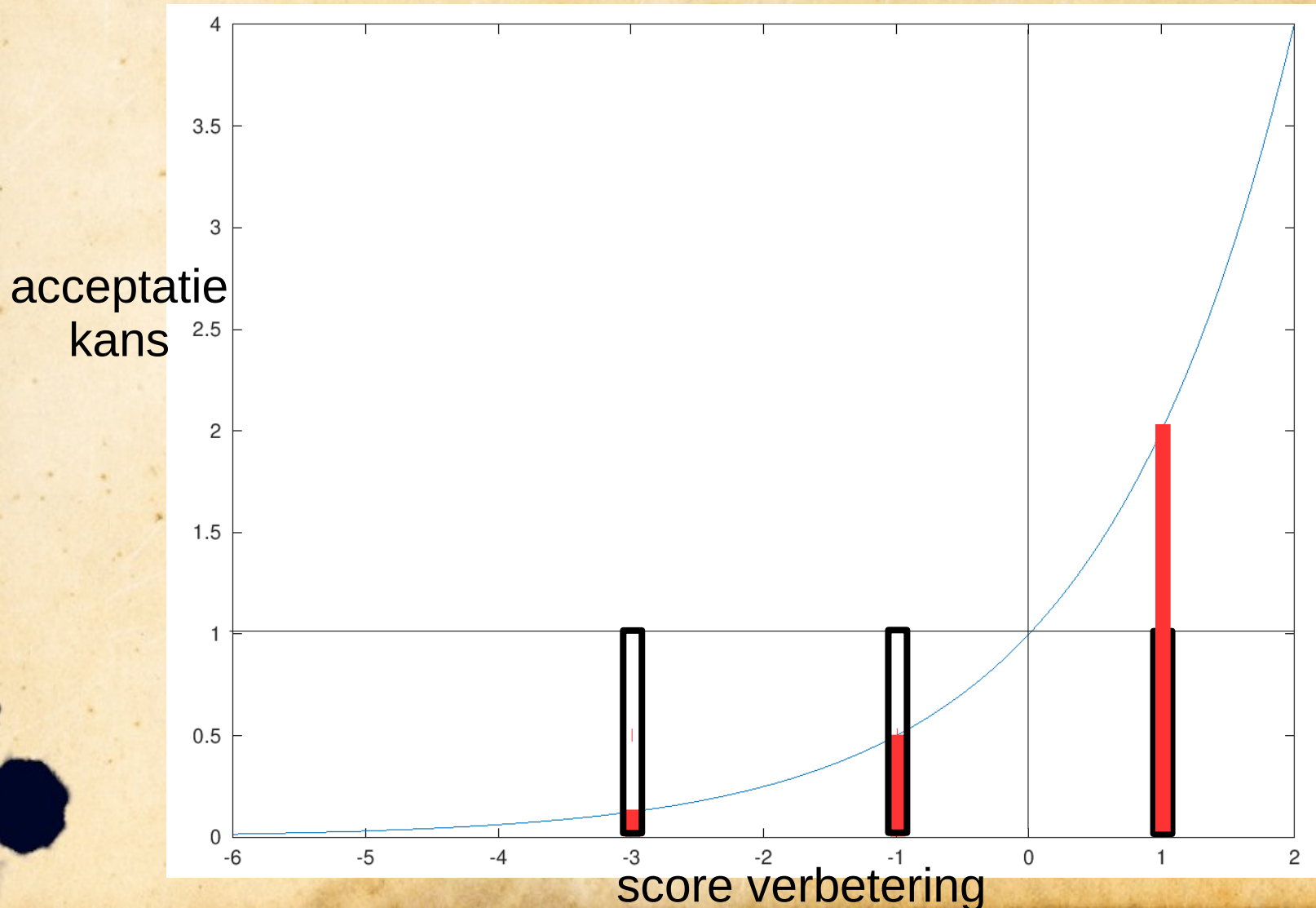
# Simulated Annealing

- Minimaliseren:  $\text{acceptatiekans} = 2(\text{score\_old} - \text{score\_new})$



# Simulated Annealing

- Minimaliseren:  $\text{acceptatiekans} = 2(\text{score\_old} - \text{score\_new})$





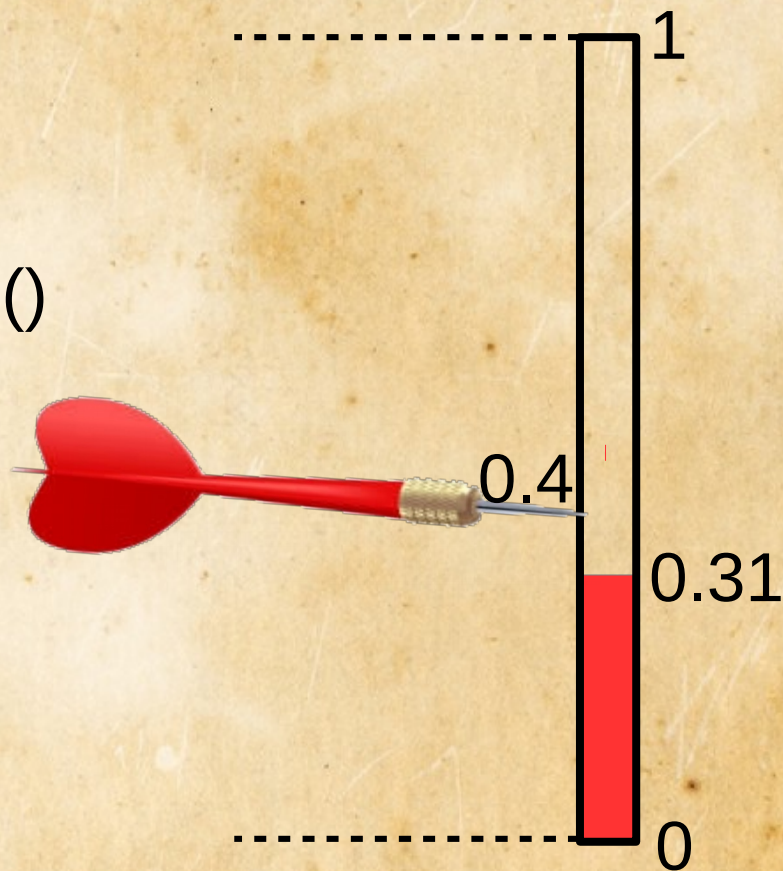
# Simulated Annealing

- Wel of niet accepteren?

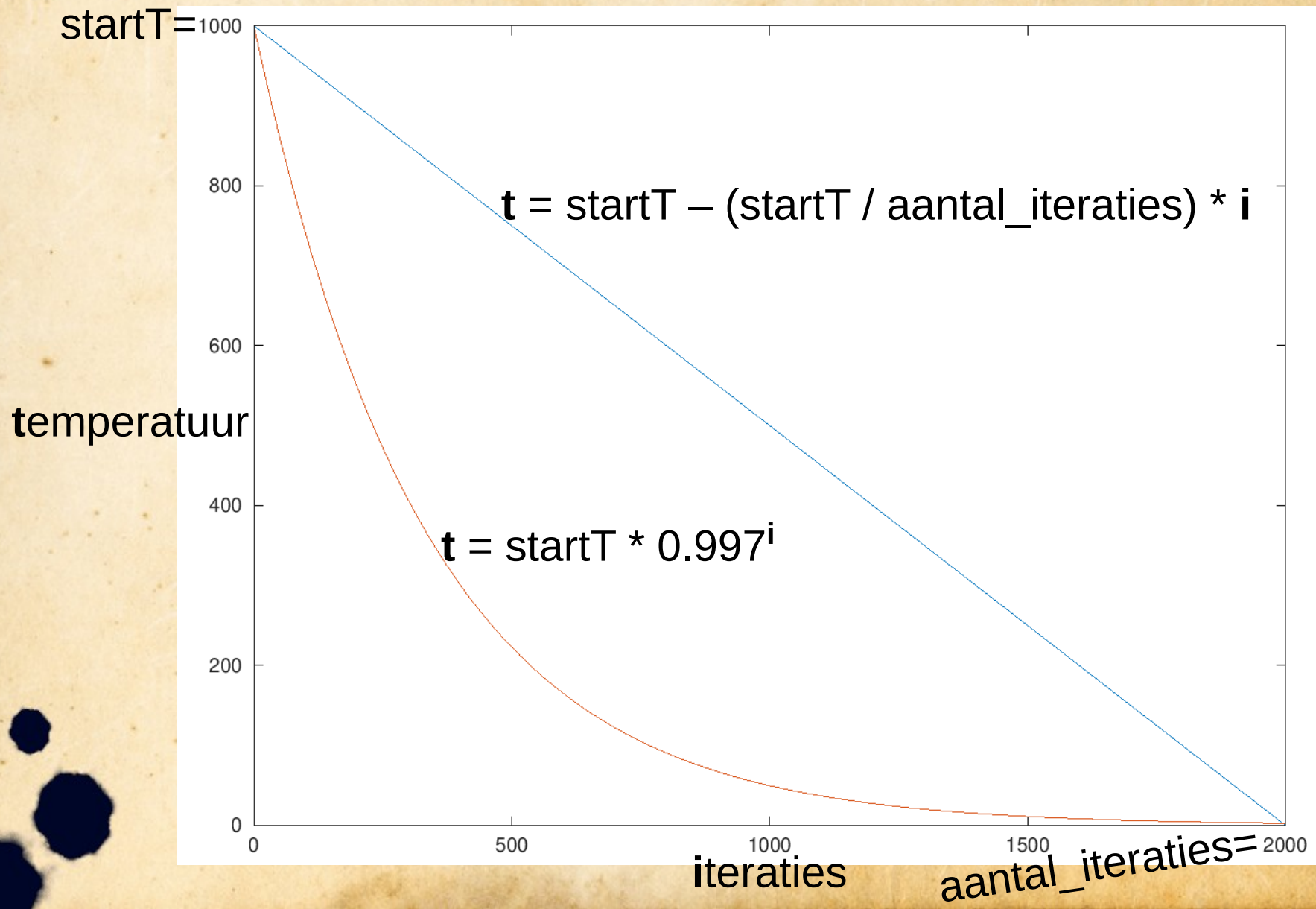
`r = random.random()`

if `r < kans`:

...



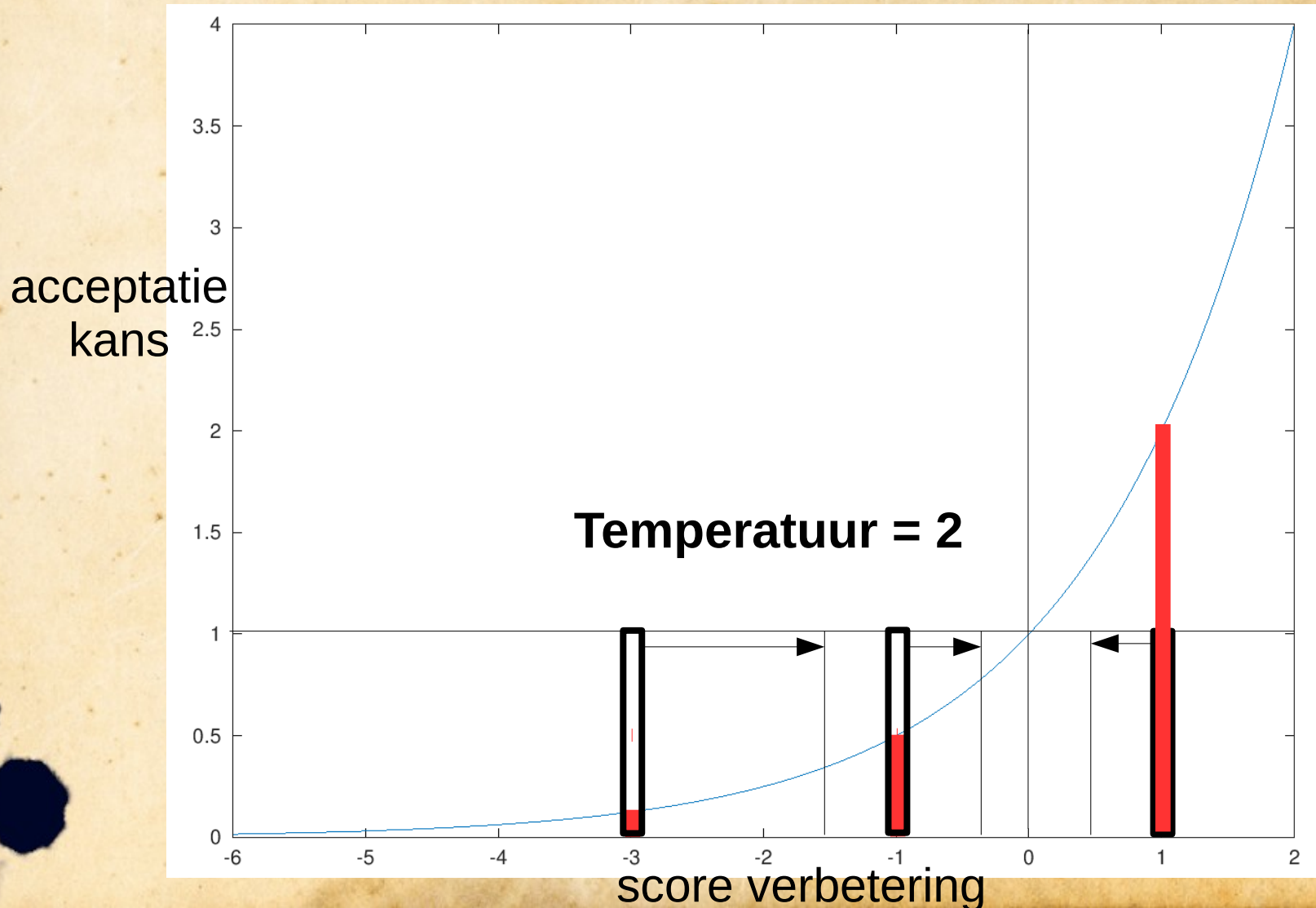
# Simulated Annealing





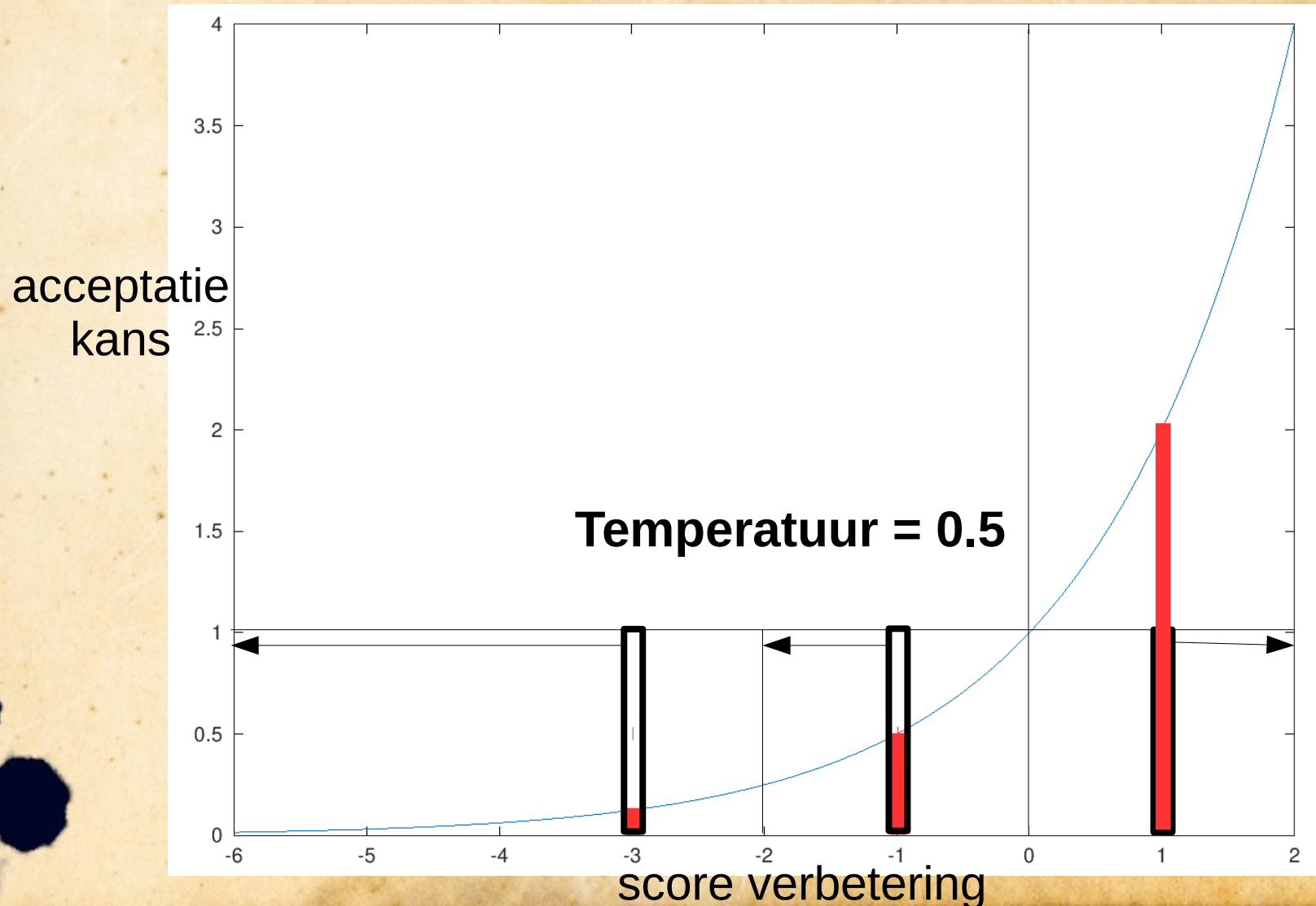
# Simulated Annealing

- acceptatiekans =  $2(\text{score\_old} - \text{score\_new}) / \text{temperatuur}$



# Simulated Annealing

- $\text{acceptatiekans} = 2(\text{score\_old} - \text{score\_new}) / \text{temperatuur}$





# Simulated Annealing, pseudo code

Herhaal:

Kies een random start state

**Kies start temperatuur**

Herhaal **N iteraties**:

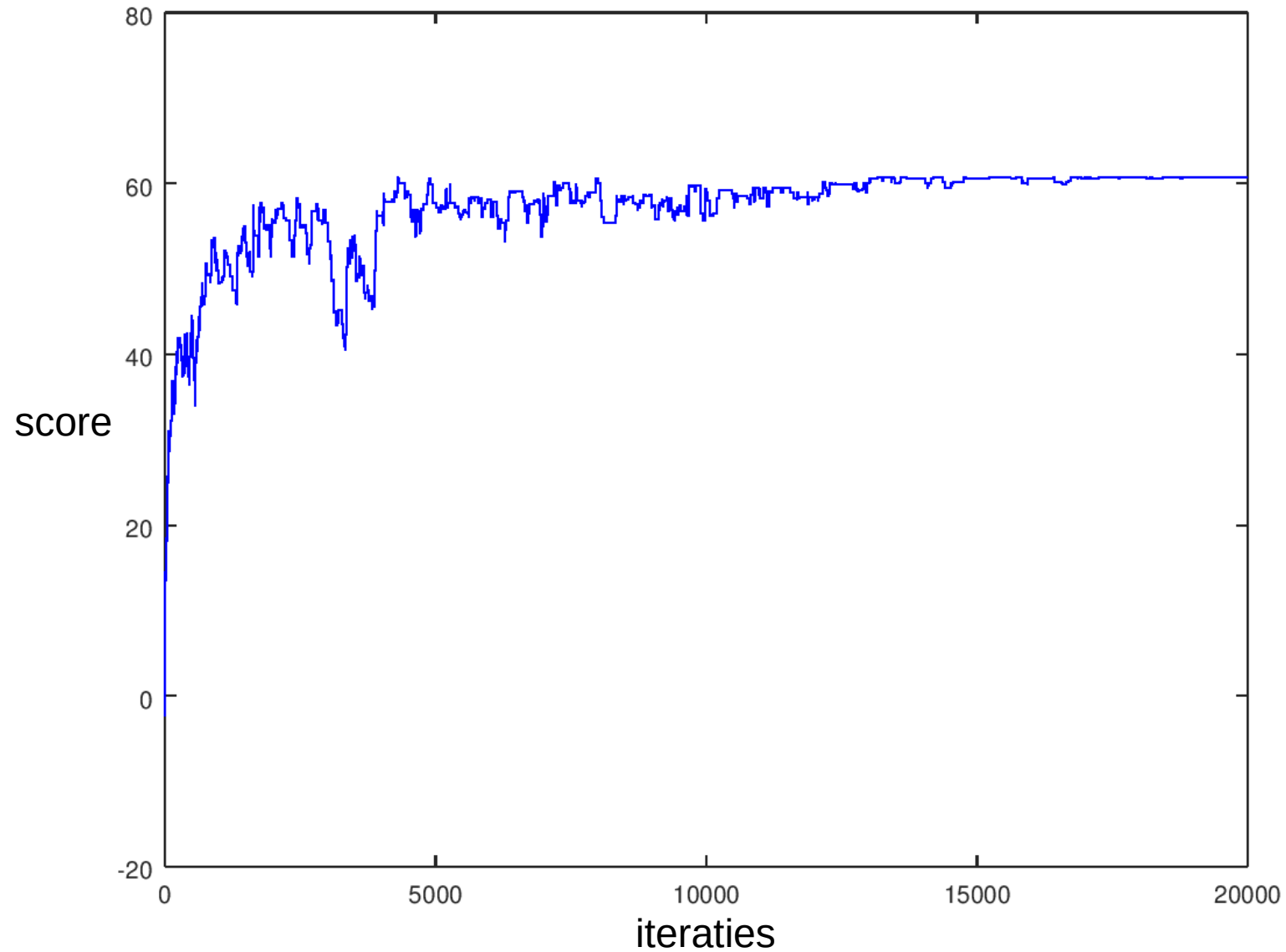
Doe een kleine random aanpassing

Als **random( ) > kans(oud, nieuw, temperatuur)**:

Maak de aanpassing ongedaan

**Verlaag temperatuur**

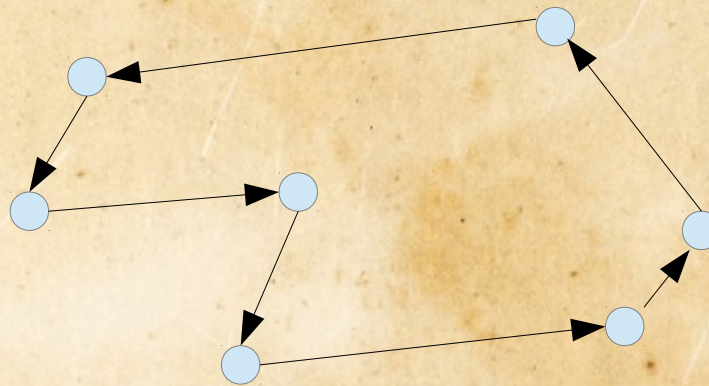
# Simulated Annealing





# Traveling Salesman, Demo

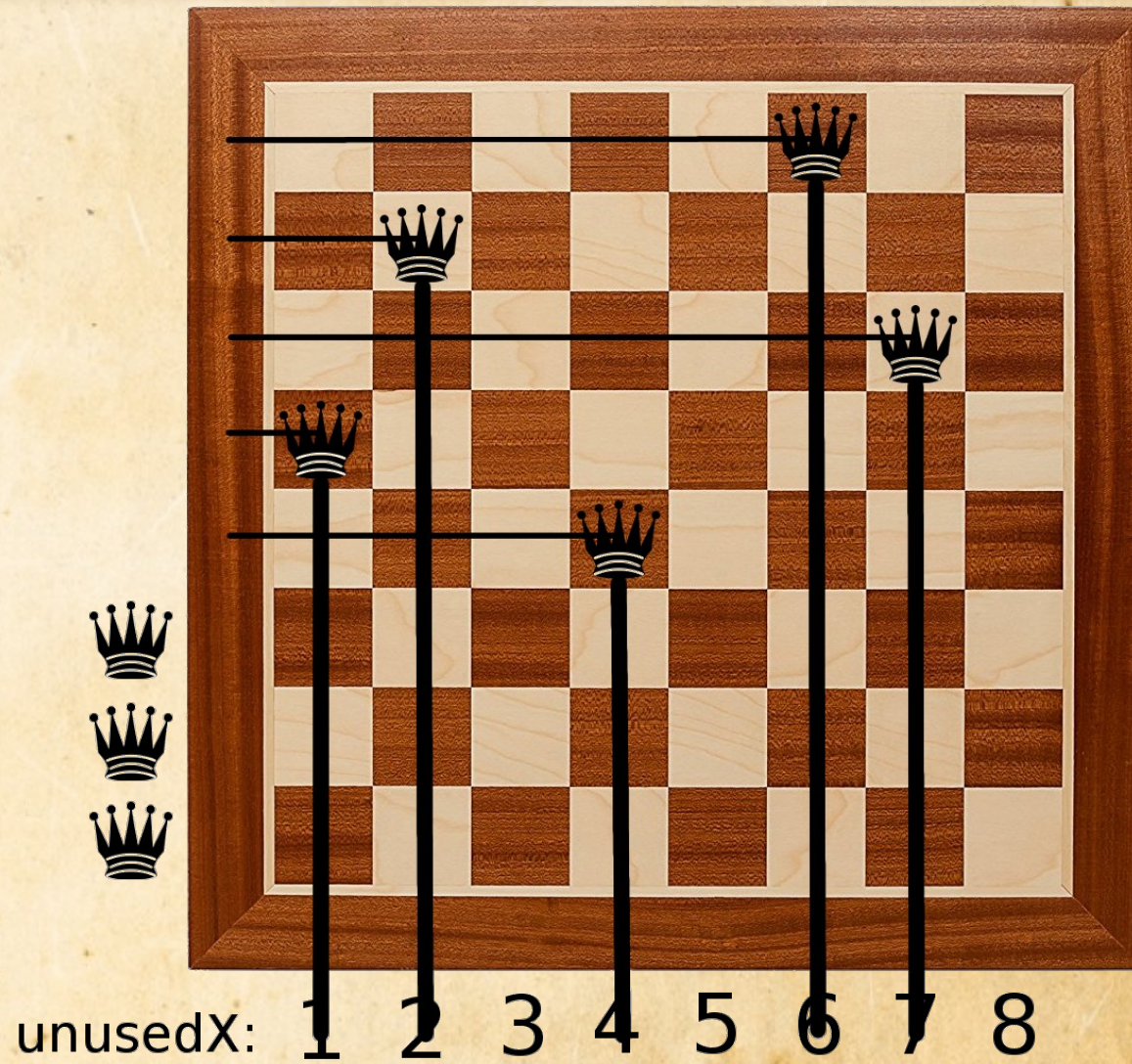
- Een verkoper zoekt de kortste route langs N steden en terug naar huis:



- Youtube Video:
  - <https://www.youtube.com/watch?v=SC5CX8drAtU>
    - Greedy
    - Random
    - Hill Climber
    - Simulated Annealing



# N-Queens





# N-Queens

=== N:10

\_queensX: [1, 3, 5, 7, 9, 0, 2, 4, 6, 8]

```
. Q . . . . . . .  
. . . Q . . . . .  
. . . . . Q . . .  
. . . . . . . Q .  
. . . . . . . . Q  
Q . . . . . . . .  
. . Q . . . . . .  
. . . . Q . . . .  
. . . . . Q . . .  
. . . . . . . Q .
```

Pattern works for even N that are not  $6K + 2$

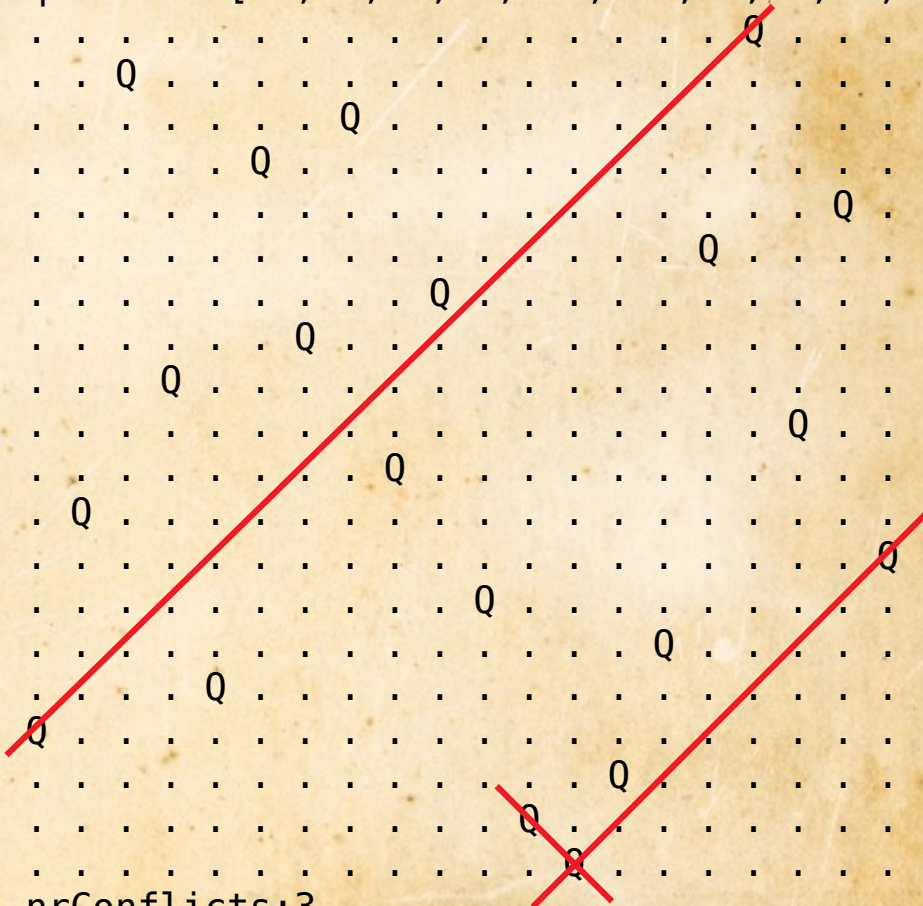
N	not working
4	
6	
8	*
10	
12	
14	*
16	
18	
20	*
22	
24	
26	*
28	
..	..



# N-Queens, Demo

GitHub: <https://github.com/bterwijn/NQueens>

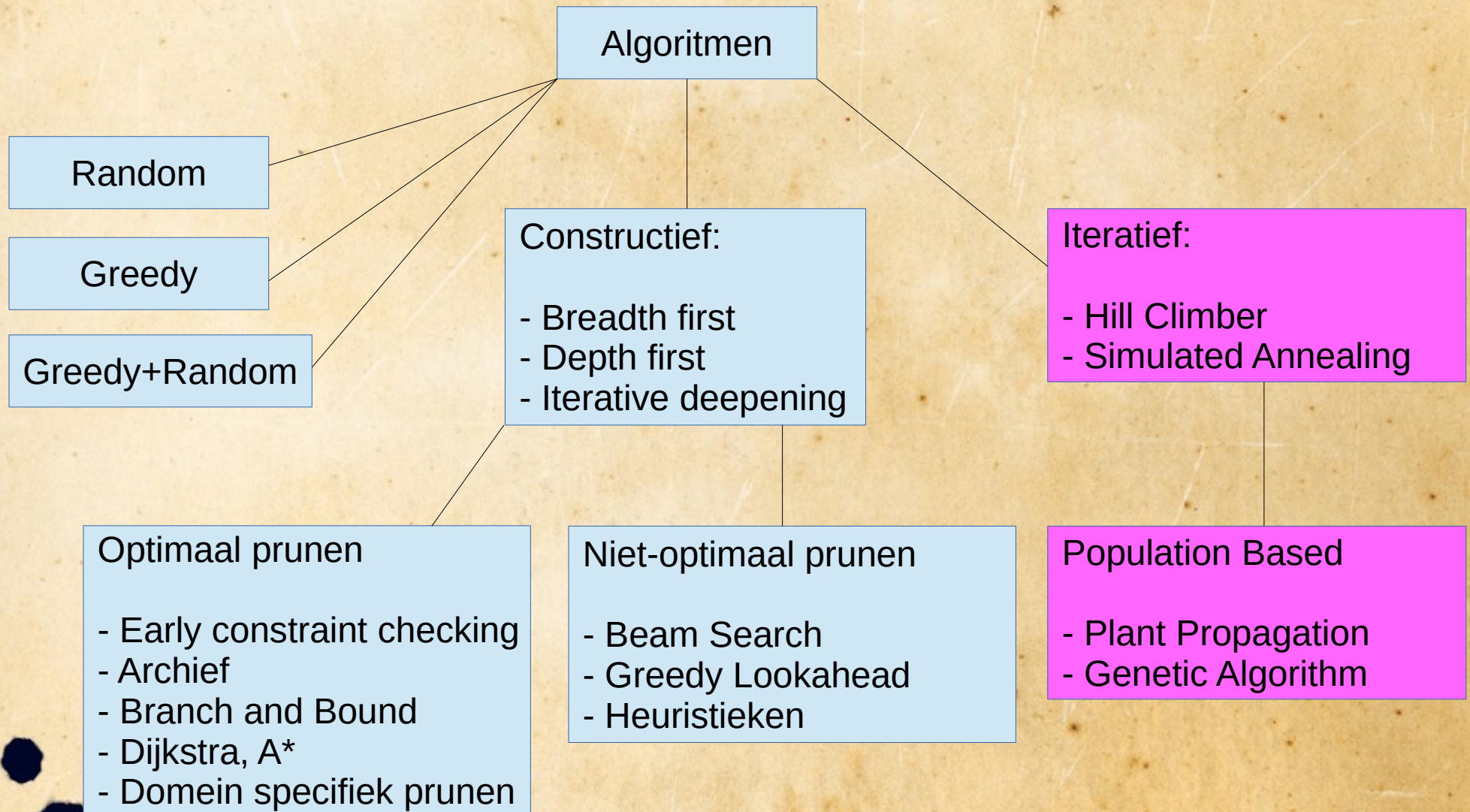
```
python3 -m NQueens.ConstructiveAlgorithms.Random 20 p  
queensX: [16, 2, 7, 5, 18, 15, 9, 6, 3, 17, 8, 1, 19, 10, 14, 4, 0, 13, 11, 12]
```



nrConflicts:3

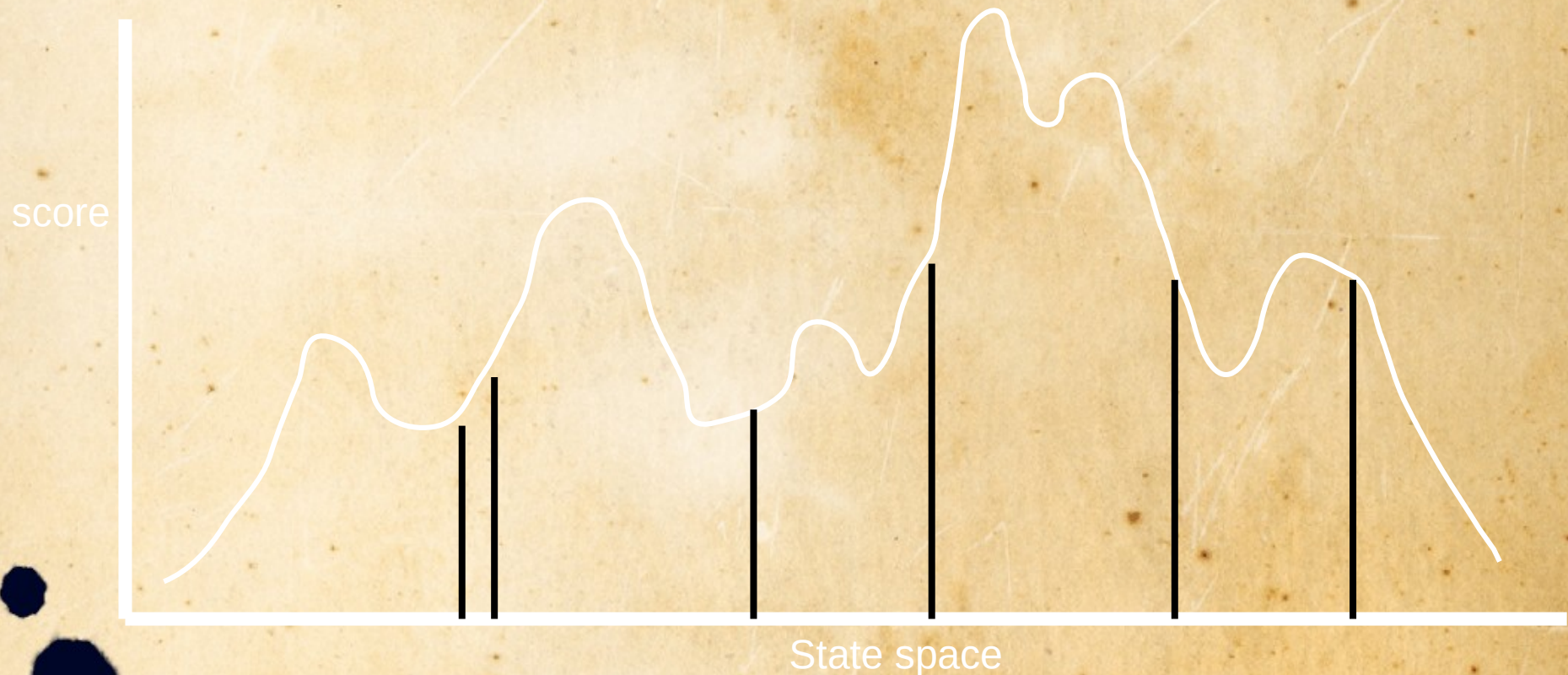


# Algoritmen





# Population based algorithms

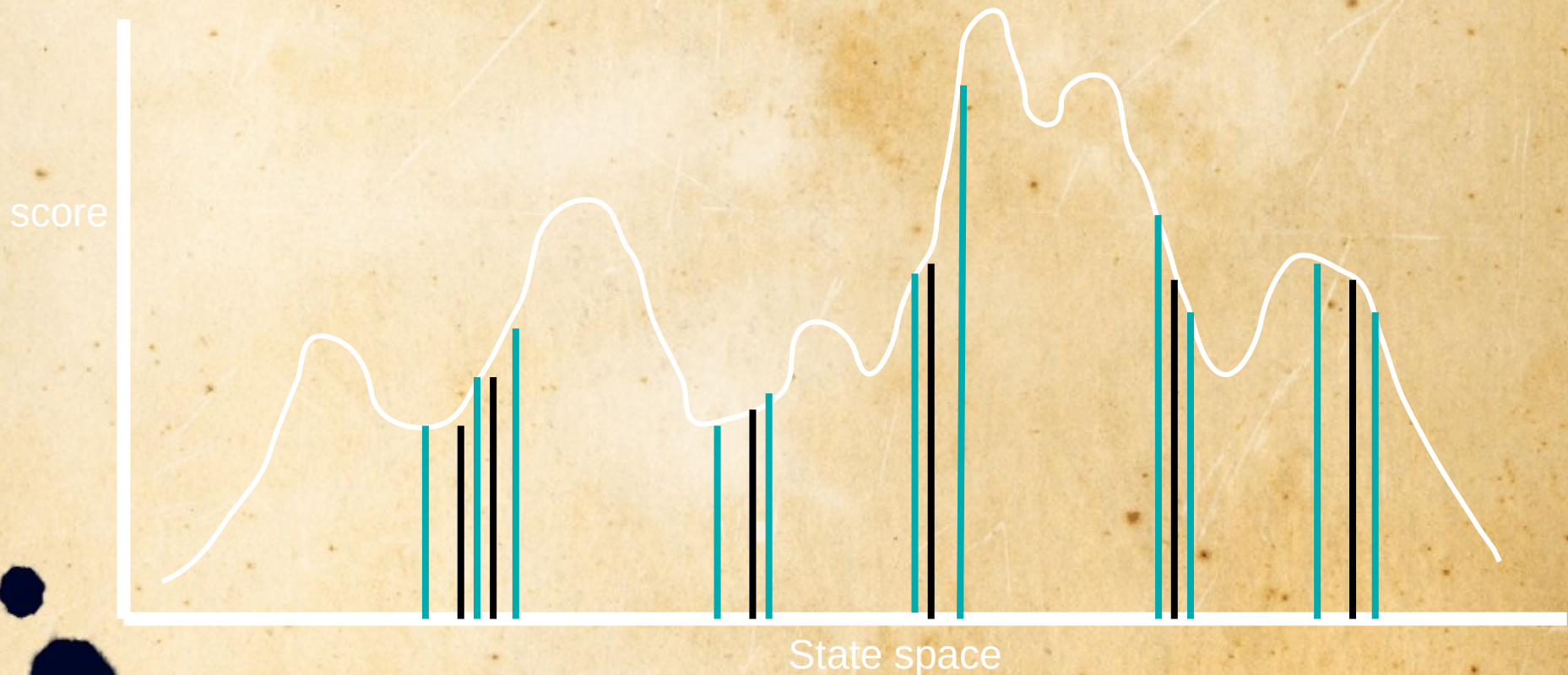


Populatie van Hill Climbers: 6



# Population based algorithms

Populatie krijgt nieuwe kinderen

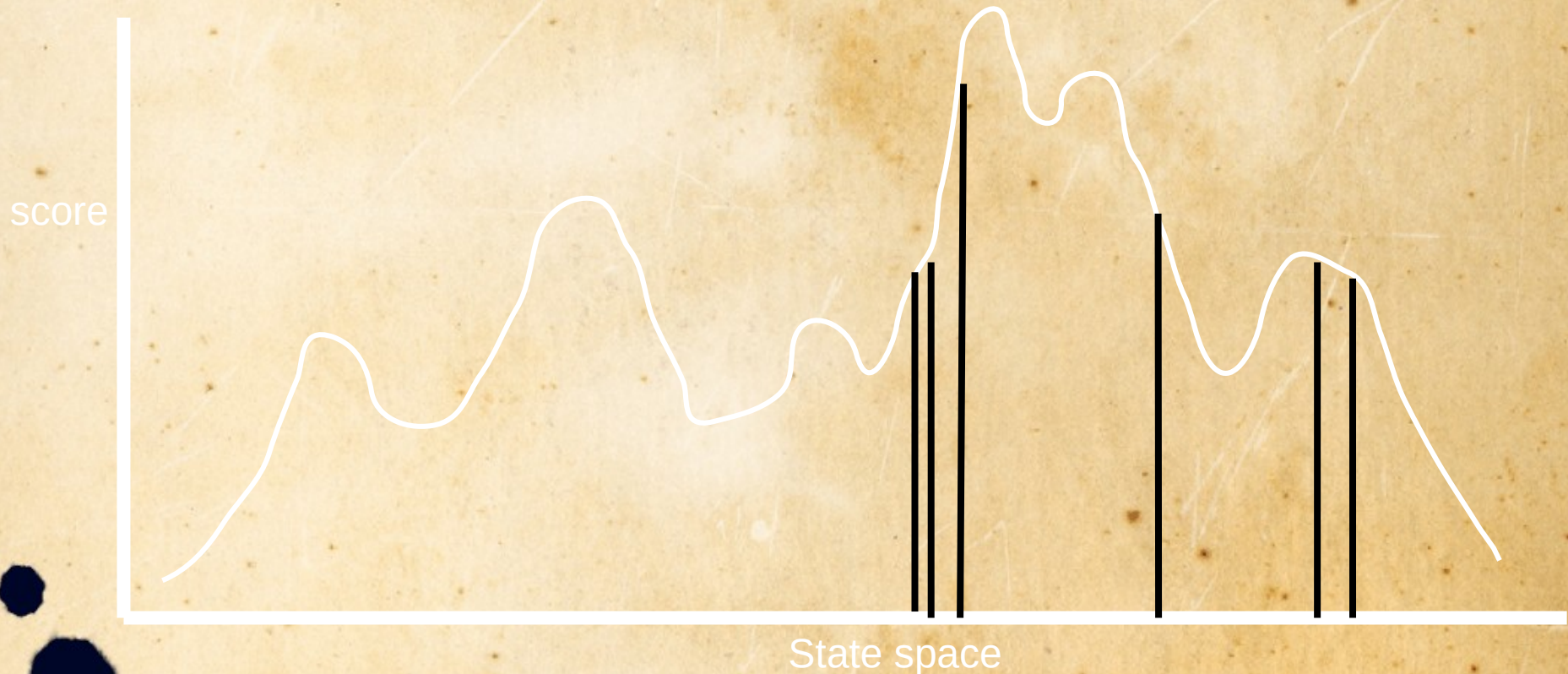


Populatie van Hill Climbers: 6



# Population based algorithms

Survival of the fittest



Populatie van Hill Climbers: 6

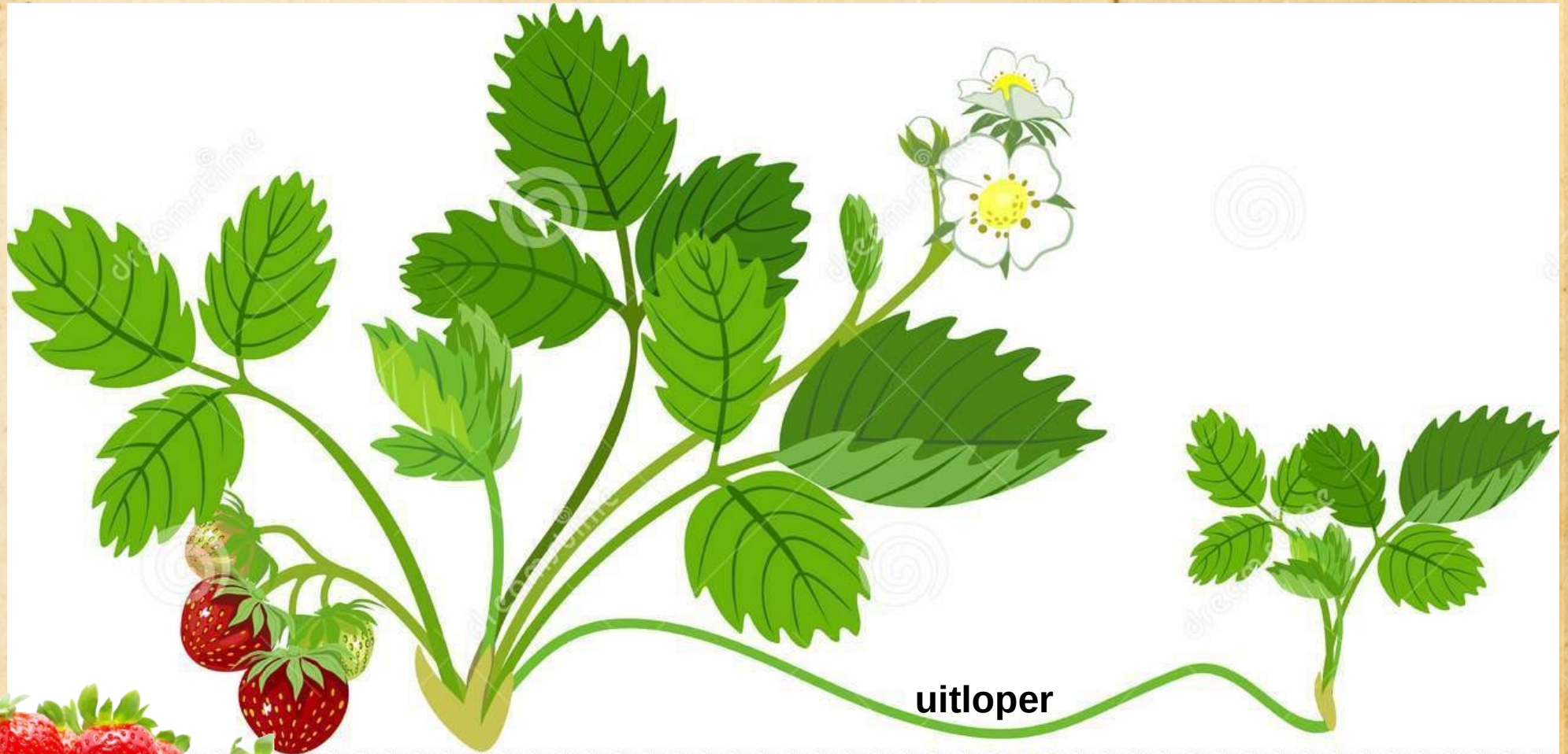


# Plant Propagation Algorithm

- Paper:
  - **Nature-Inspired Optimisation Approaches and the New Plant Propagation Algorithm**
    - Abdellah Salhi, Eric S Fraga

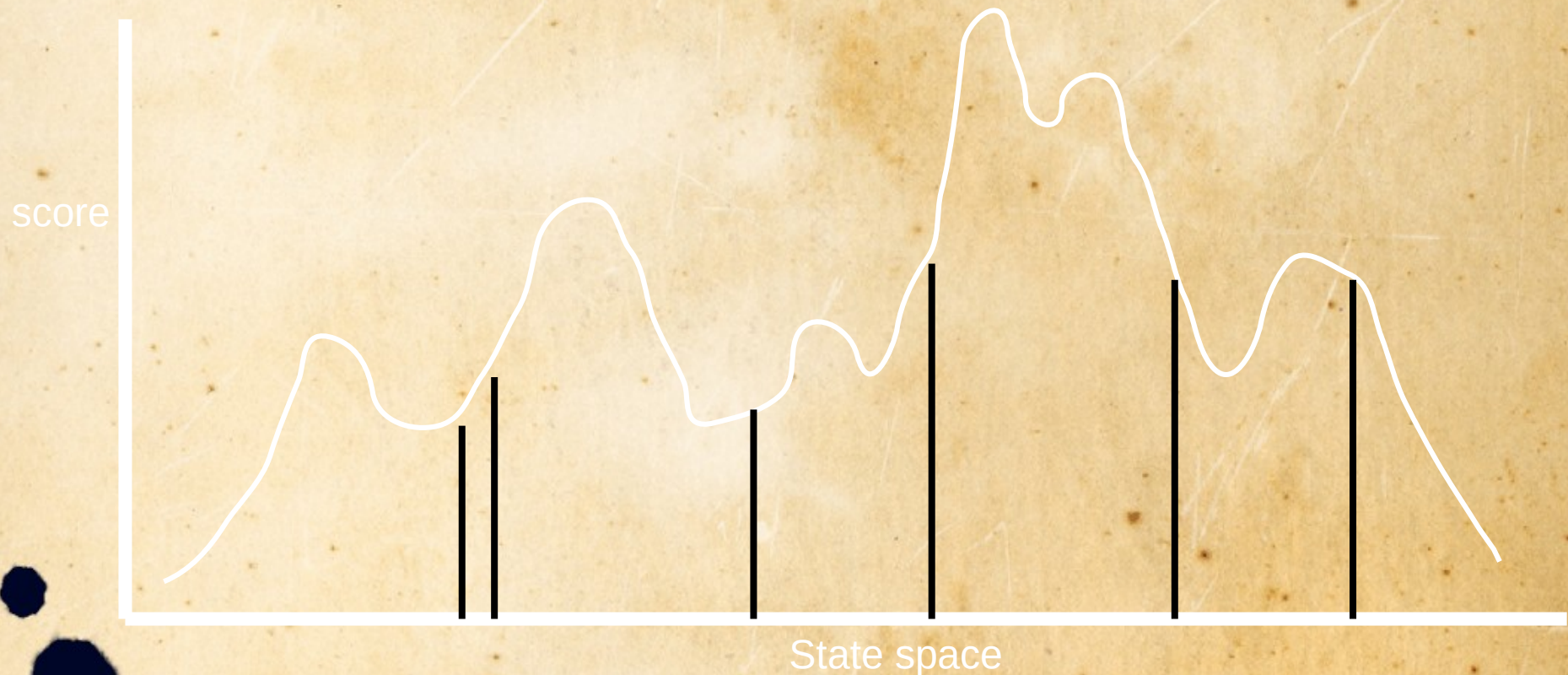


# Plant Propagation Algorithm





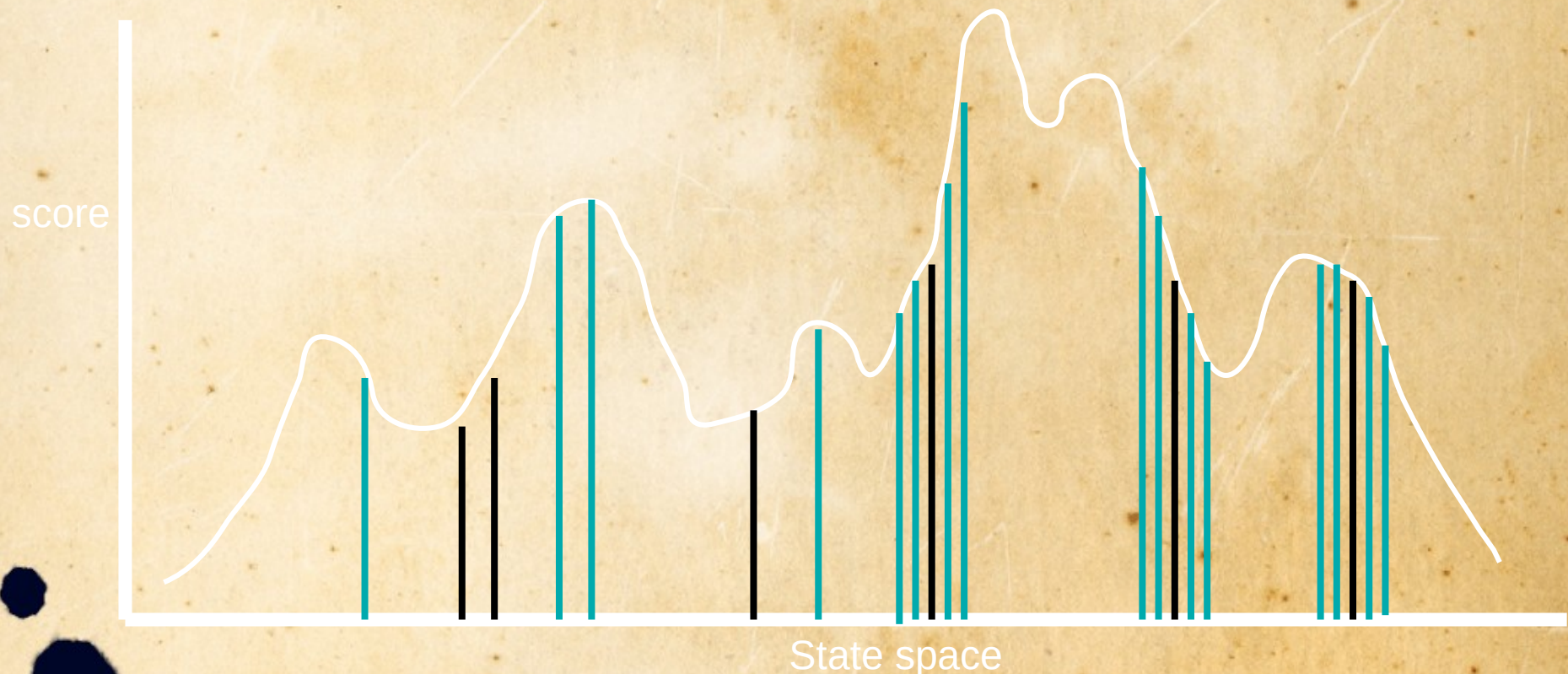
# Plant Propagation Algorithm



Populatie van Hill Climbers: 6



# Plant Propagation Algorithm

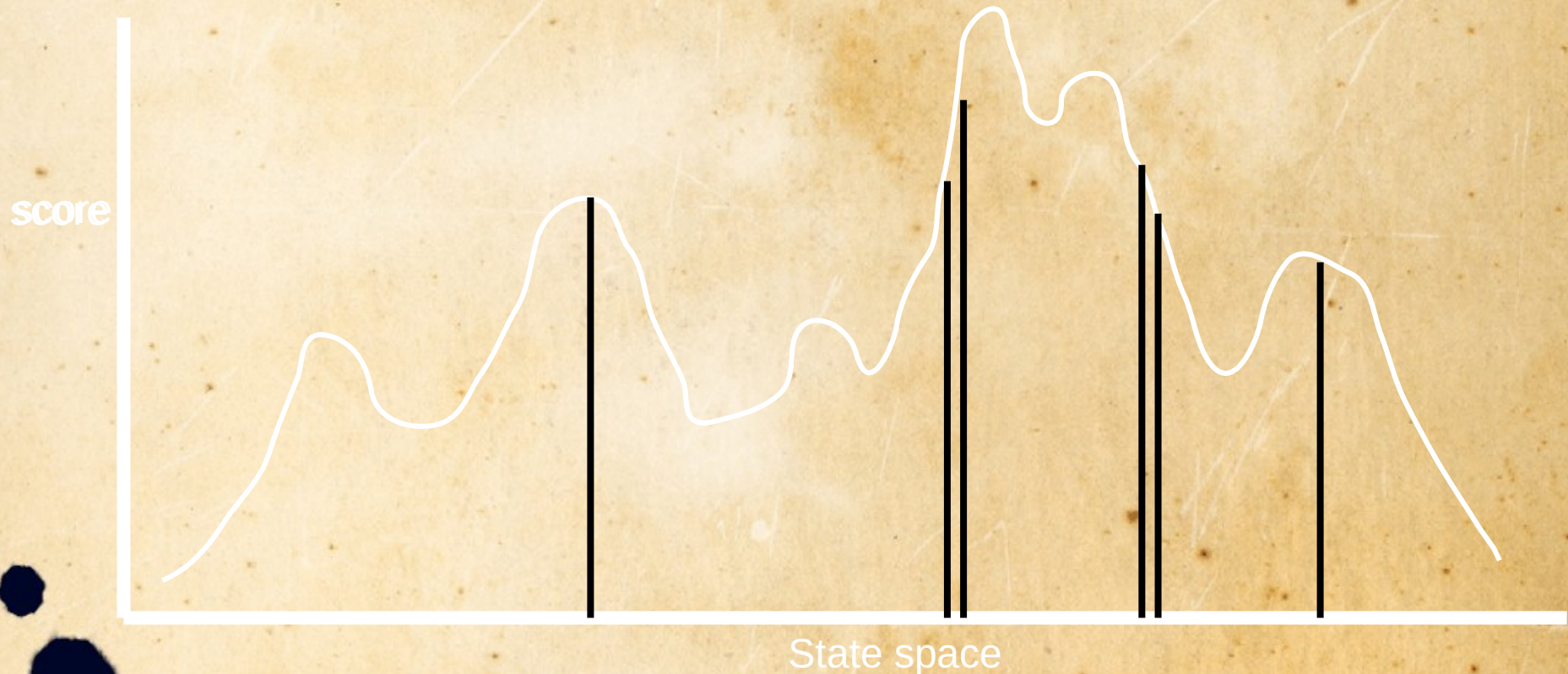


Populatie van Hill Climbers: 6



# Plant Propagation Algorithm

Selecteer beste 6



Populatie van Hill Climbers: 6



# Plant Propagation Algorithm

**Require:** objective  $f(x)$ ,  $x \in \mathcal{R}^n$

Generate a population  $P = \{p_i, i = 1, \dots, m\}$

$g \leftarrow 1$

**for**  $g \leftarrow 1$  **to**  $g_{\max}$  **do**

    compute  $N_i = f(p_i), \forall p_i \in P$

    sort  $P$  in descending order of  $N$

    create new population  $\phi$

**for** each  $p_i, i = 1, \dots, m$  **do** {best  $m$  only}

$r_i \leftarrow$  set of runners where both the size of the set and the distance for each runner (individually) is proportional to the fitness  $N_i$

$\phi \leftarrow \phi \cup r_i$  {append to population; death occurs by omission above}

**end for**

$P \leftarrow \phi$  {new population}

**end for**

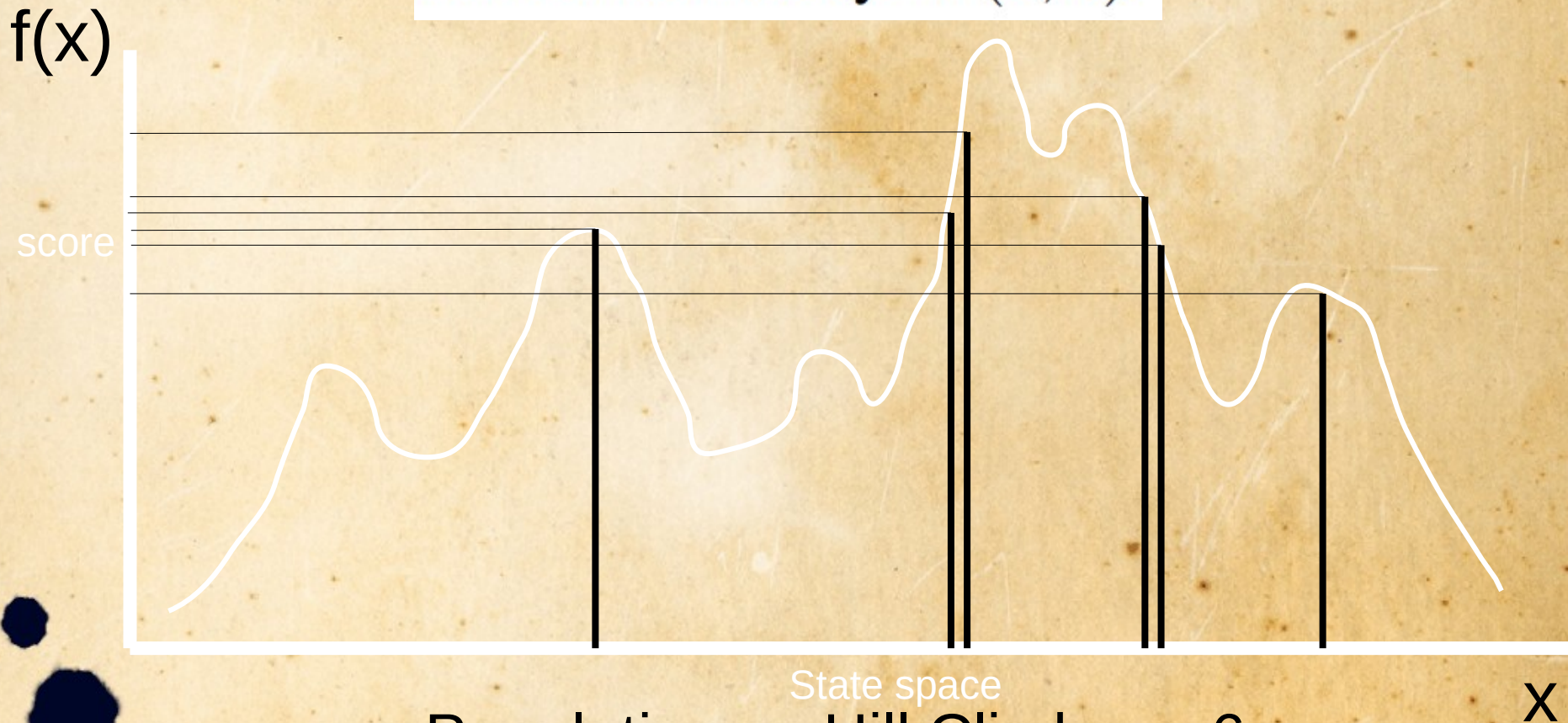
**return**  $P$ , the population of solutions



# Plant Propagation Algorithm

$$N(x) = \frac{1}{2} (\tanh(4f(x)) - 2) + 1$$

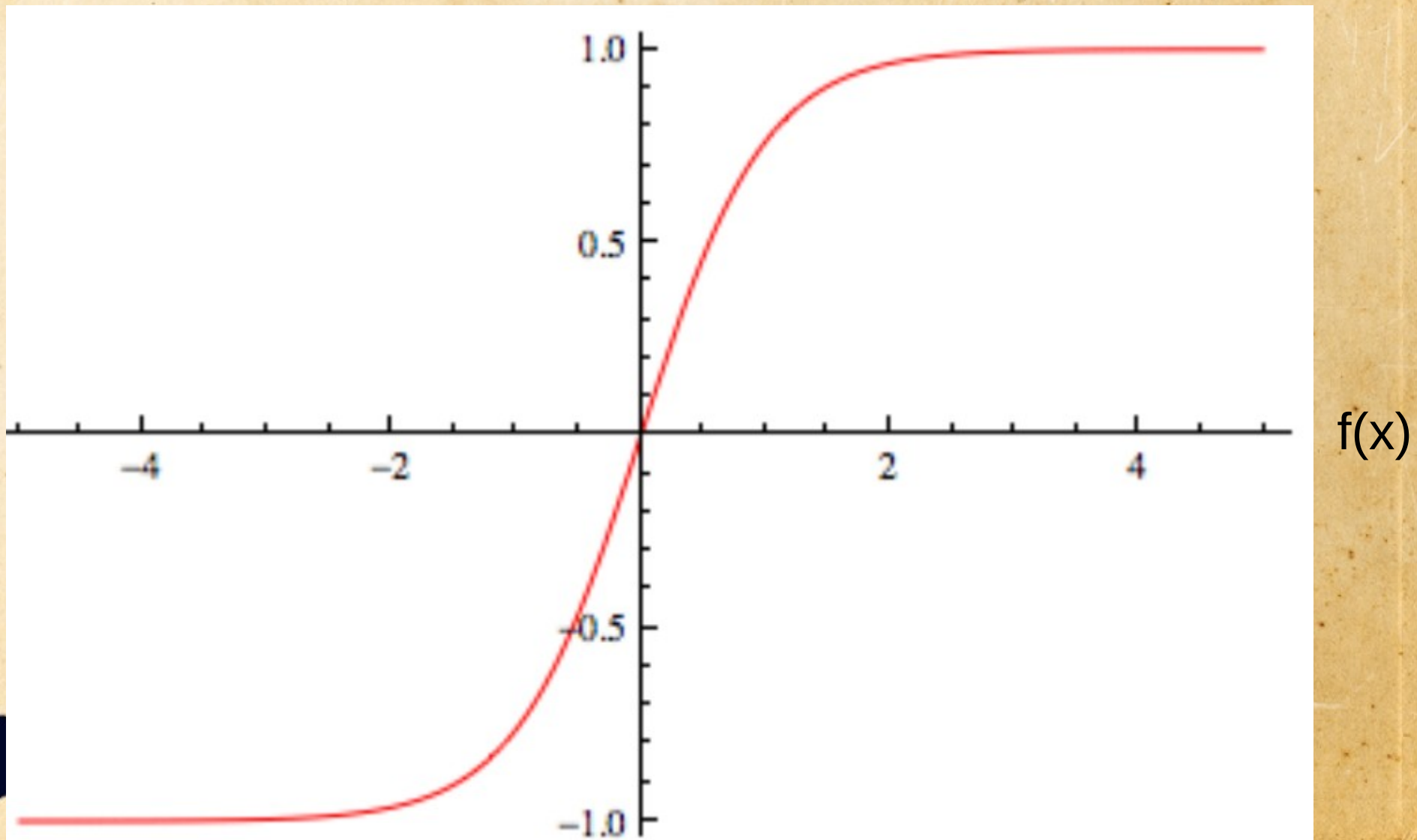
values lie strictly in  $(0, 1)$ .



Populatie van Hill Climbers: 6

# Plant Propagation Algorithm

$\tanh( f(x) )$





# Plant Propagation Algorithm

Number of runners :

$$n_r = \lceil n_{\max} N_i r \rceil$$

$n_{\max}$  is the maximum number of runners to generate in this paper,  $n_{\max} = 5$ .

$r \in [0, 1]$  is a randomly chosen number



# Plant Propagation Algorithm

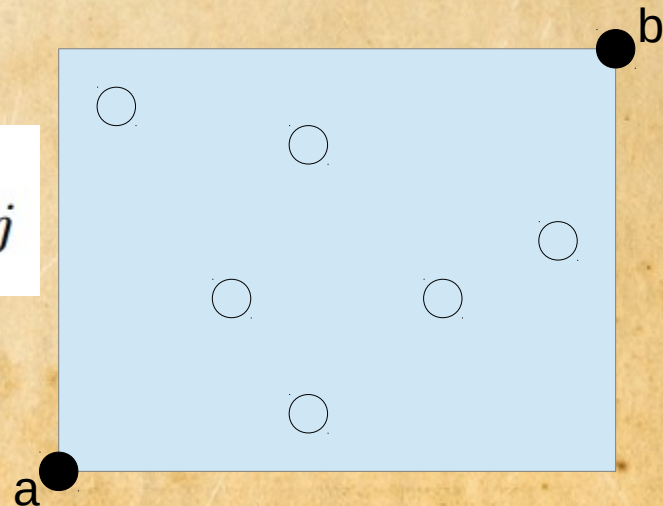
Distance of runner:

$$d_{r,j} = 2(1 - N_i)(r - 0.5)$$

Each  $d_{r,j}$  will be in  $(-1,1)$ .

for  $j = 1, \dots, n$ , where  $n$  is the dimension of the search space.

$$x_j^* = x_j + (b_j - a_j)d_{r,j}$$



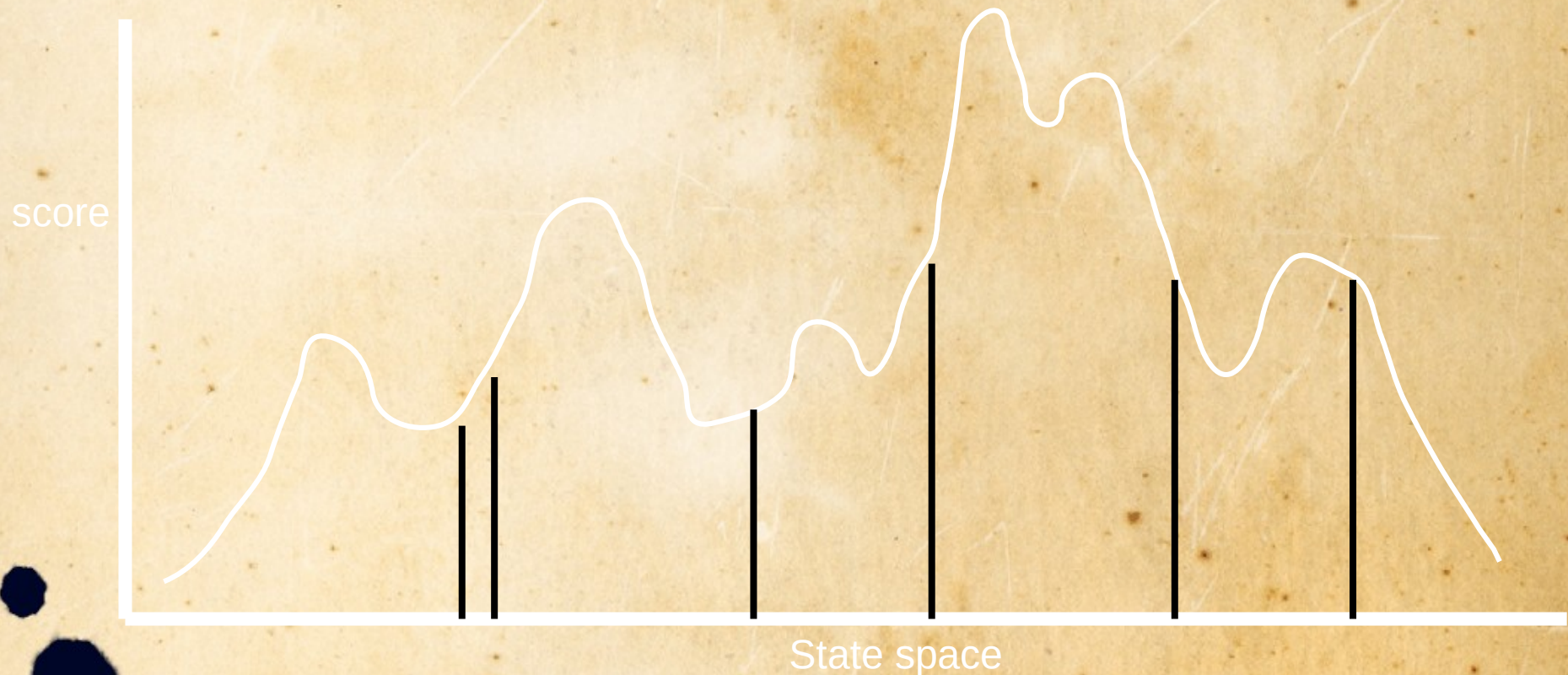


# Genetic Algorithm

- Website:
  - [https://www.tutorialspoint.com/genetic\\_algorithms](https://www.tutorialspoint.com/genetic_algorithms)



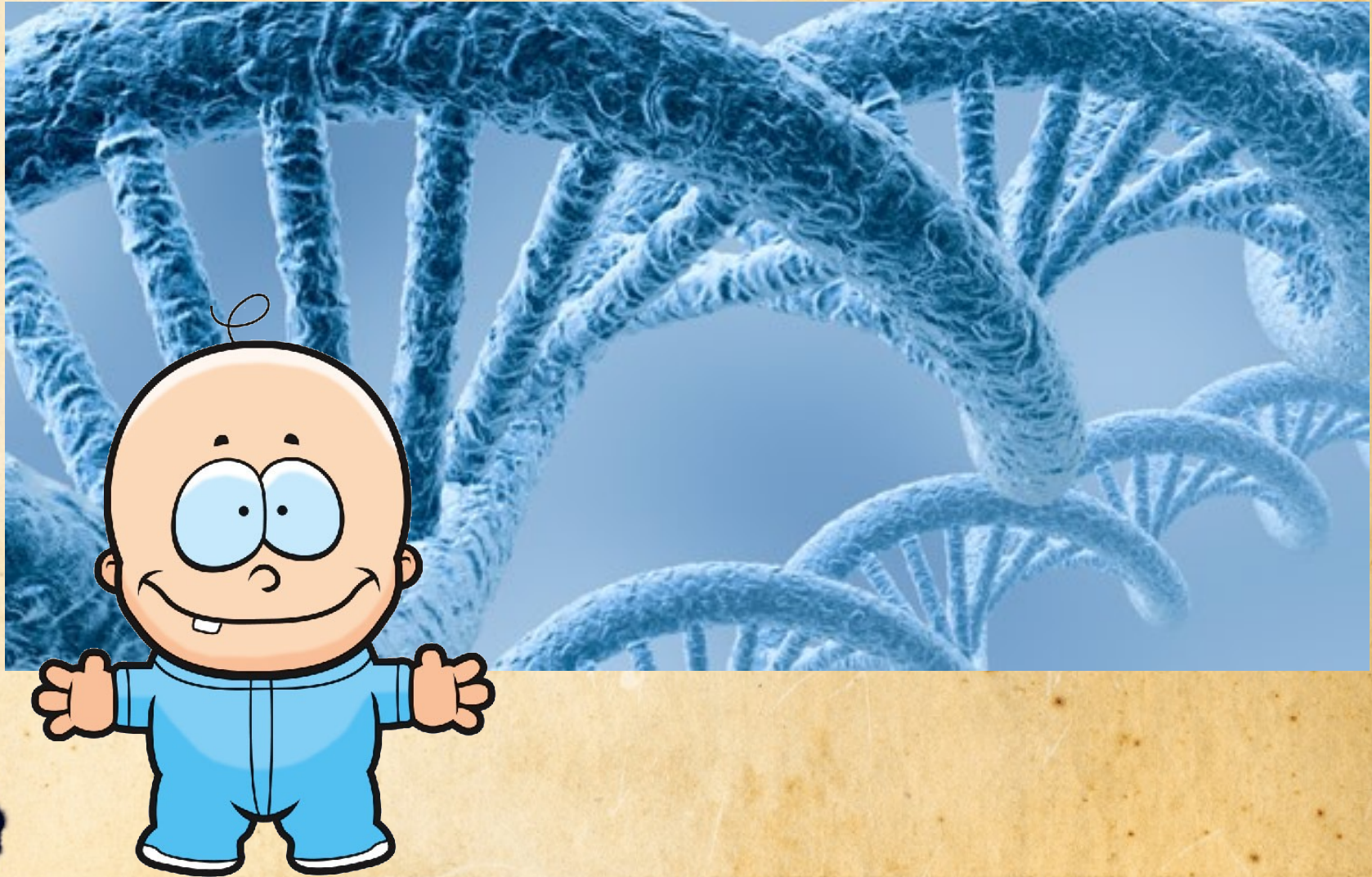
# Genetic Algorithm



Populatie van Hill Climber: 6

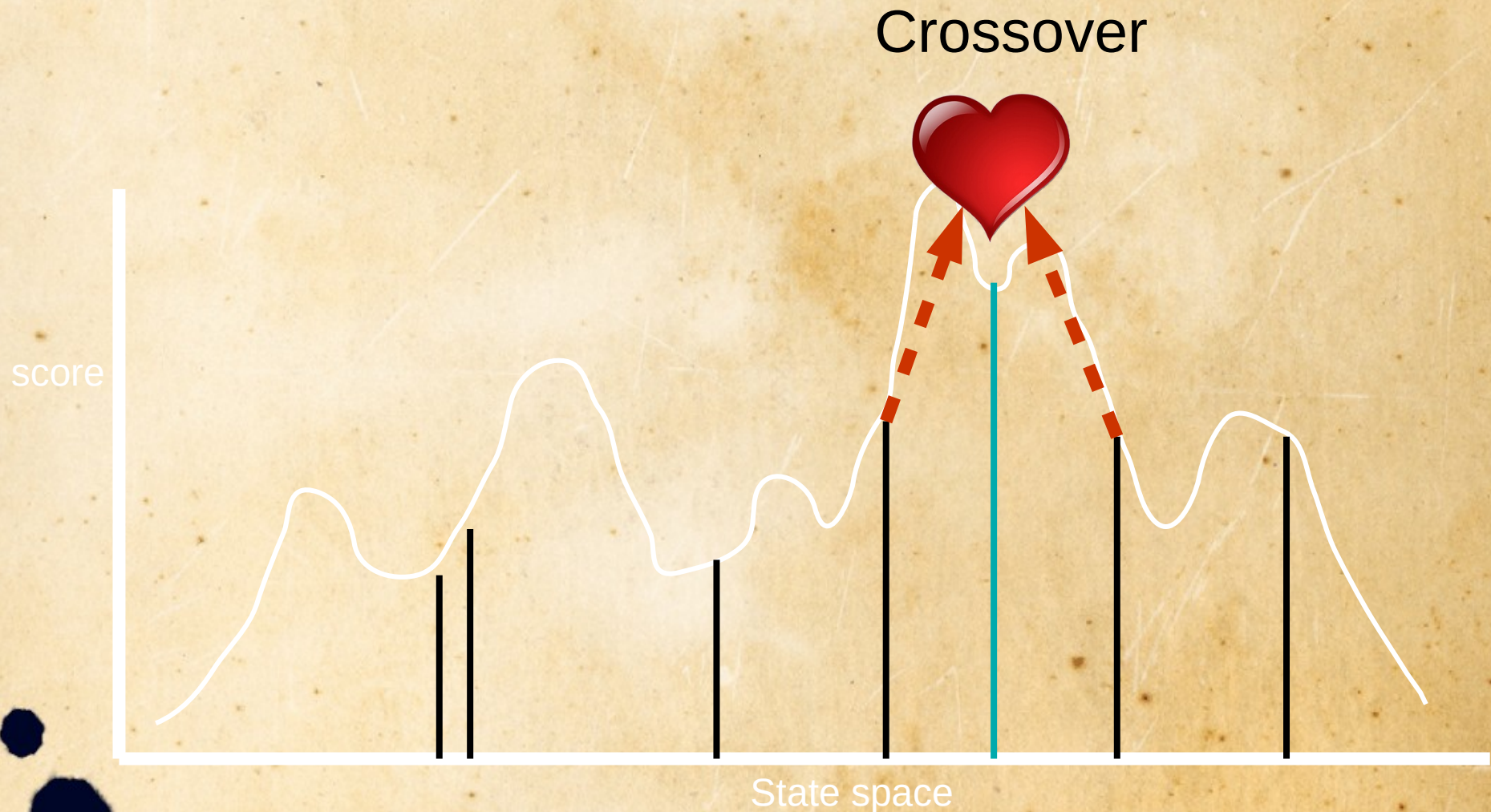


# Genetic Algorithm





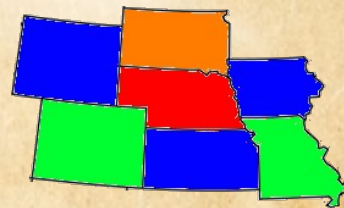
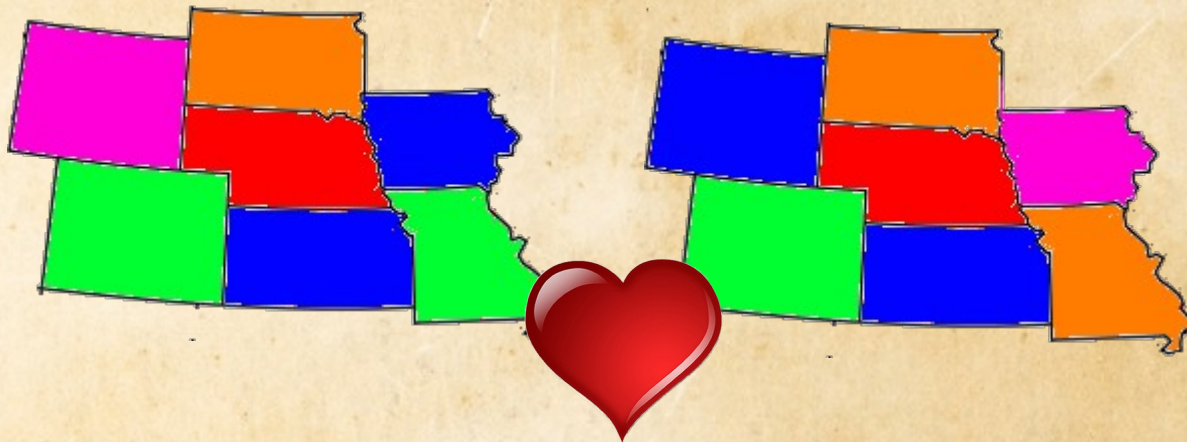
# Genetic Algorithm



Populatie van Hill Climber: 6

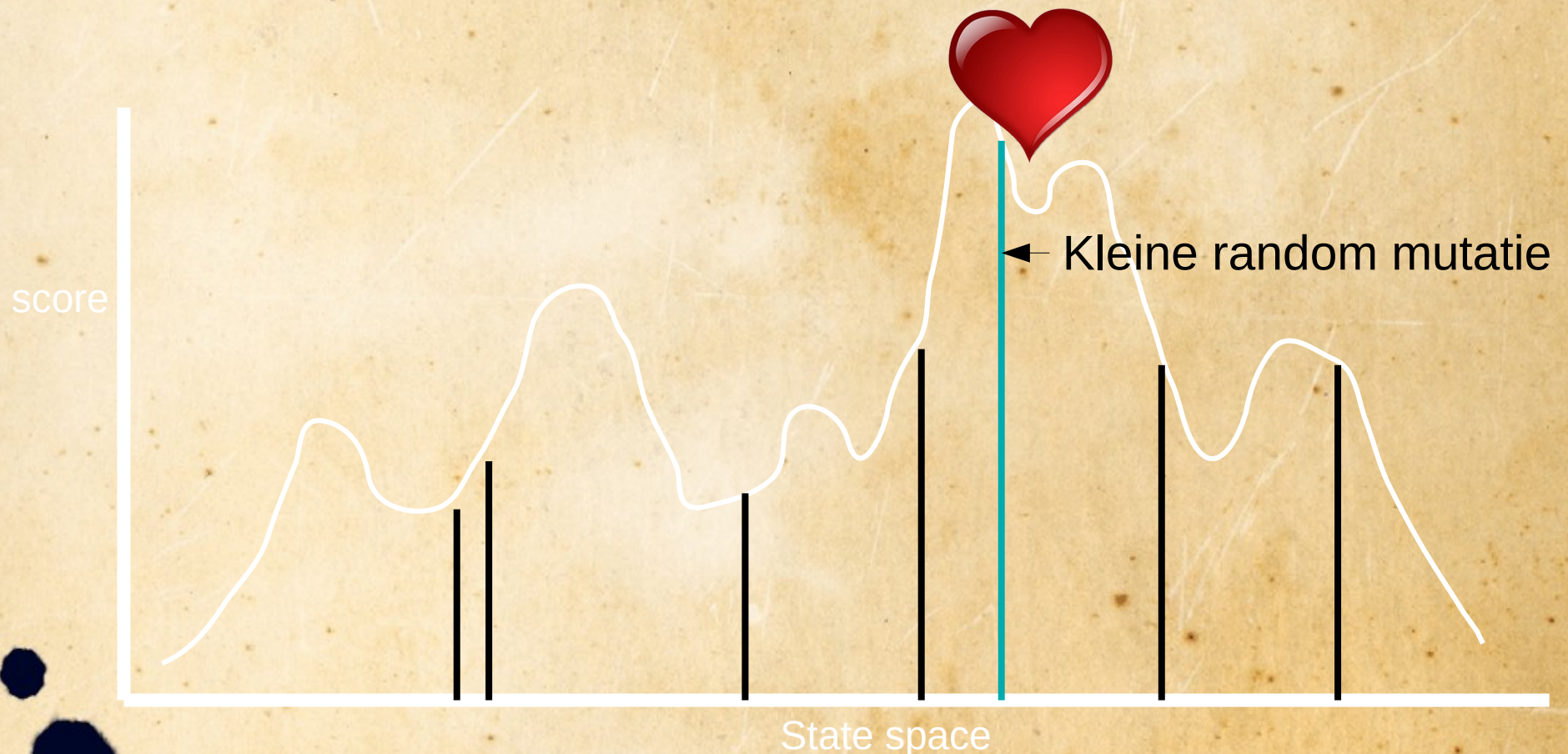


# Genetic Algorithm





# Genetic Algorithm



Populatie van Hill Climber



# Genetic Algorithm

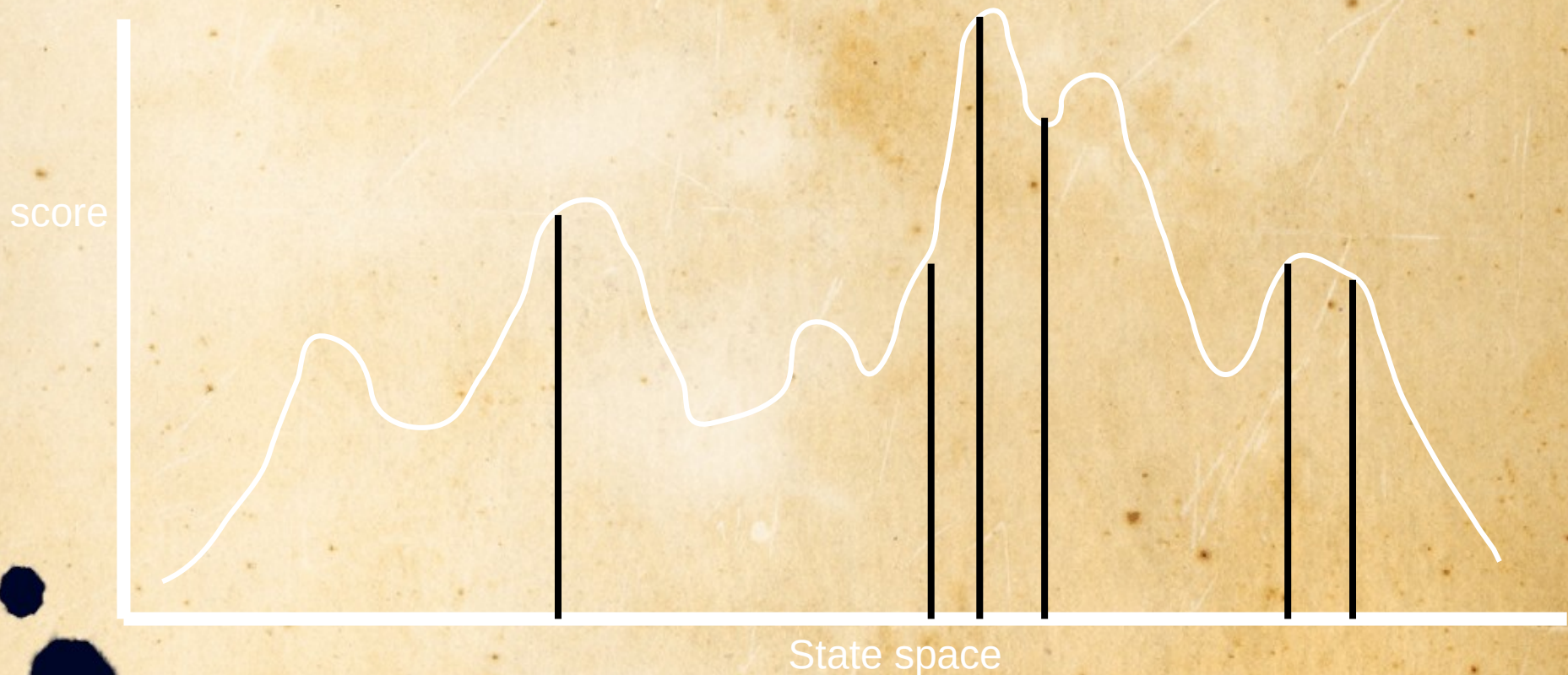


Populatie van Hill Climber



# Genetic Algorithm

Select the best 6



Populatie van Hill Climber: 6

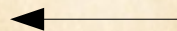
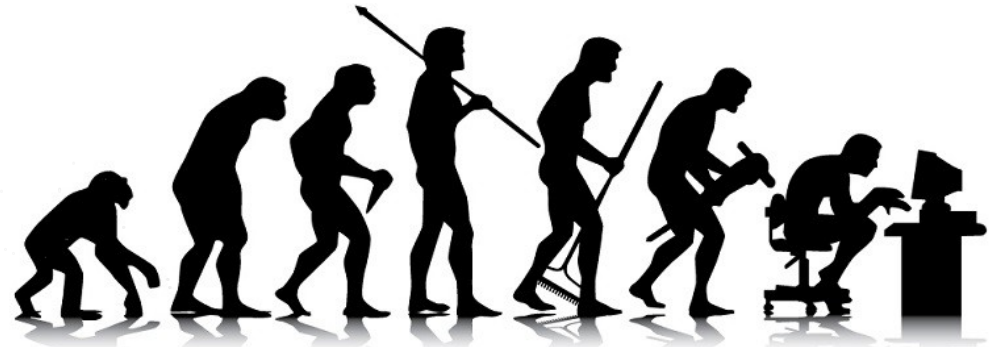
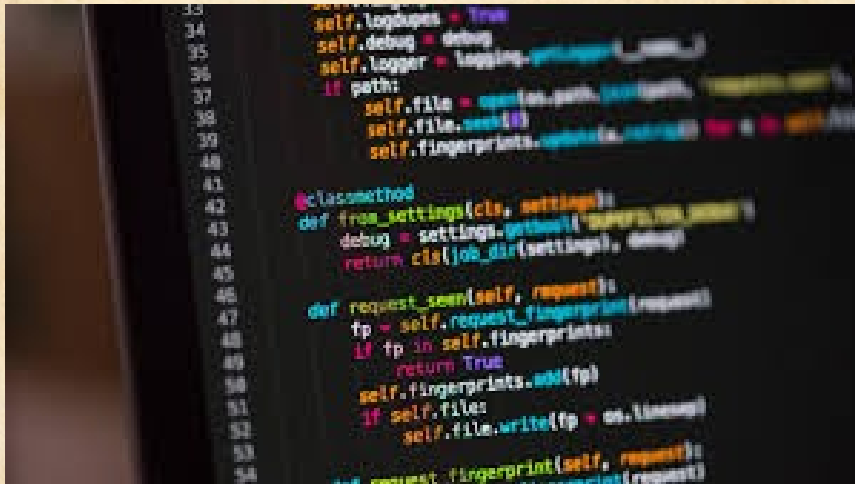


# Genetic Algorithm





# Genetic Algorithm



**Heuristieken  
Programmeertheorie**



# Simulation Hypothesis



IS THIS A  
SIMULATION?



# Andere population based algoritmen

- Ant colony optimization
- Artificial immune system
- Bee colony optimization
- Brain storm optimization
- Fireworks algorithm
- Particle swarm optimization
- ...

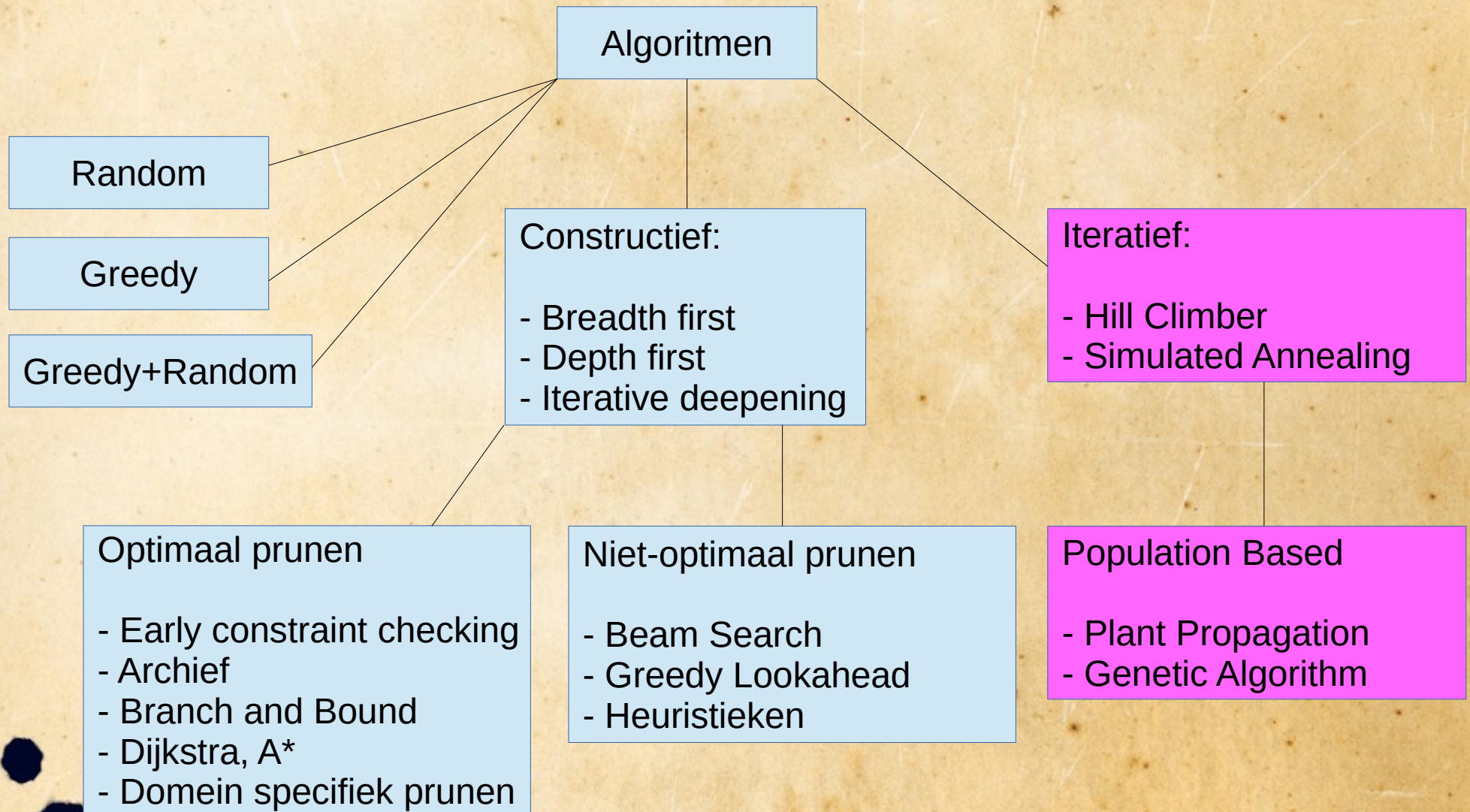


# Population based, Demo

- Learn to walk, Roberto Mior ([miorsoft.itch.io](http://miorsoft.itch.io))
  - Youtube video:
    - [www.youtube.com/watch?v=qtmG8mMGbpo](http://www.youtube.com/watch?v=qtmG8mMGbpo)



# Algoritmen





# Stuart J. Russell and Peter Norvig

## Hfdstk 6.3 Most Constraint Variable heuristic





# Lecture video 2020, terugkijken



- Youtube playlist, Iteratieve algoritmen:
  - <https://www.youtube.com/playlist?list=PLJBtJTYGPSzJaxroYW-6OH1NRuUFqpGER>



# Tips



- Eerst de basis algoritmen, daarna pas creatieve uitbreiding
- Houd je code zo simpel mogelijk
  - Johan Cruijff: “Voetbal is simpel. Wat moeilijk is, is simpel voetballen.”
- Maak een klein probleempje om mee te testen
- Los je case eerst met de hand op voor goede heuristische ideeën
- Doe leuke onderzoekjes, experimenteer
  - welke heuristiek/parameter-waarde werkt het best?
- Future work, wat zou ik gaan doen in vervolgonderzoek?