Introduction Scenario/Program Optimization Algorithm Experiments/Results Summary

Simulation And Optimization Of Traffic Light

Öztürk Emrah, Nicolai Schwartze, Jan Wenger

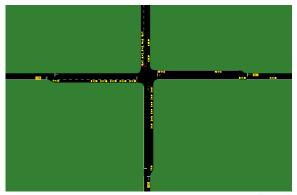
January 28, 2020



- Introduction
- Scenario/Program
- Optimization Algorithm
- 4 Experiments/Results
- Summary

optimization of fairness, stop time and number of stops

$$\mathit{func} = \frac{\mathit{g}_1\mathit{fairness} + \mathit{g}_2\mathit{stop}_\mathit{time} + \mathit{g}_3\mathit{number}_\mathit{stops}}{3} \to \mathsf{min}$$

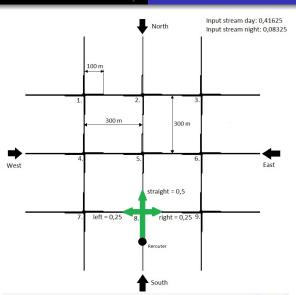


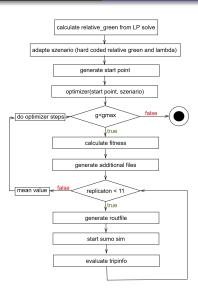
Simplifications

- simplifications of software SUMO
- velocity, gap between cars and acceleration same
- one constant input stream from every geographic direction (N,S,W,E)
- perfection = 1, same turn probabilities (0.5 straight ahead, 0.25 left and right)
- one type of car, non intelligent, dynamic rerouteing
- exponential distribution for input streams
- stop simulation after 3600 simulation steps



- Introduction
- 2 Scenario/Program
- Optimization Algorithm
- 4 Experiments/Results
- Summary





- Introduction
- 2 Scenario/Program
- Optimization Algorithm
- 4 Experiments/Results
- Summary

Differential Evolution

```
Differential evolution
```

```
1: population \leftarrow initialization
 2: while g < G_{max} do
        for individual x_i in population do
 3:
           v_i = mutation(x_i, population, F)
 4:
           u_i = \text{crossover}(x_i, v_i, CR)
5:
           if function(u_i) < function(x_i) then
6:
 7:
               x_i = u_i
8:
           end
9:
           g = g + 1
10:
        end
11: end
```

NSGA-II

```
1: population \leftarrow initialization
 2: while g < G_{max} do
        for i in \frac{population}{2} do
 3:
 4:
             p_1, p_2 = \text{tournament selection}(P_t)
             q_1, q_2 = \operatorname{crossover}(p_1, p_2)
 5:
             a_1 = mutations(a_1)
 6:
 7:
             q_2 = \text{mutations}(q_2)
             Q_t = Q_t \cap (q_1, q_2)
 8:
 g٠
        end
10:
        R_t = R_t \cap (Q_t)
        R_t = \text{fast non dominated sort}(R_t)
11:
        P_t = \text{crowding distance sorting}(R_t, P_t. \text{size})
12:
13:
        g = g + 1
14: end
```

Conjugate Gradient Descent

```
1: x,d \leftarrow initialization
 2: while ||\nabla f(x)|| > \epsilon or ||\hat{\eta}d|| < \epsilon do
         grad = numGrad(x)
 3:
      d = -grad + \frac{||grad||^2}{||grad_{old}||^2 d}
 4:
         if \frac{gradd}{||grad_{old}||||d||} > -\alpha then
 5:
              d = -grad
 6:
         end
 7:
      \hat{\eta} = linsearch(f(x, d))
 8:
 9:
     x_{old} = x
10:
    grad_{old} = grad
     x = x + \hat{\eta} d
11:
12: end
```

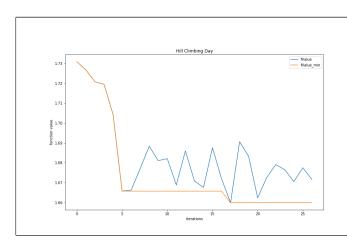
Hill Climbing

```
1: x_{start} \leftarrow \text{initialization}
 2: while fe < \#FE_{max} do
       for d in Dim do
 3:
 4:
       z[d] = step
 5:
          y_{1,2} = x \pm step
6:
       f = function(y)
       fe = fe + 1
7:
8:
      end
       if function(y_d)<function(x) then
 9:
10:
           x = y_d
11:
       end
       else
12:
           step = \frac{step}{2}
13:
14:
       end
15: end
```

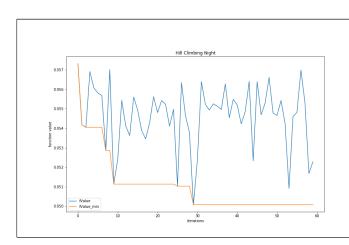
- Introduction
- 2 Scenario/Program
- Optimization Algorithm
- 4 Experiments/Results
- Summary

 $\begin{array}{c} \text{NSGA-II} \\ \text{parameter} \\ \text{epsilon} = 0.5 \\ \text{popsize} = 30 \\ \text{gen} = 20 \end{array}$

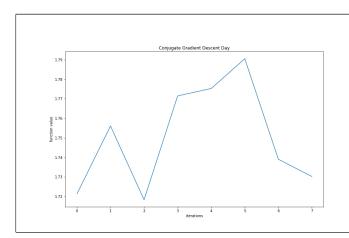
 $\begin{array}{c} {\sf NSGA-II} \\ {\sf parameter} \\ {\it epsilon} = 0.1 \\ {\it popsize} = 30 \\ {\it gen} = 15 \end{array}$



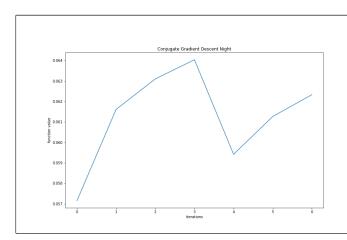
 $\begin{array}{c} \text{Hill-Climbing} \\ \text{parameter} \\ \text{\it epsilon} = 0.5 \end{array}$



 $\begin{array}{c} \text{Hill-Climbing} \\ \text{parameter} \\ \text{\it epsilon} = 0.1 \end{array}$



Cgd parameter epsilon = 0.5 lpha = 0.1 $\epsilon = 0.1$ $\eta = 10$ h = eps



 $\begin{array}{c} \mathsf{Cgd} \\ \mathsf{parameter} \\ \mathsf{epsilon} = 0.1 \\ \alpha = 0.1 \\ \epsilon = 0.1 \\ \eta = 10 \\ \mathsf{h} = \mathsf{eps} \end{array}$

- Introduction
- 2 Scenario/Program
- Optimization Algorithm
- 4 Experiments/Results
- **5** Summary

Summary

- Problems:
 - time consuming (calculation time)
 - program crash (sumo)
 - teleportation of cars
- Reduction of the cost function is observable
- No possibility to check for correctness
- ullet Less function evaluations o less calculation time needed
- Conjugate gradient descent problem with line-search (probability based rout file generation)

