

# 1 Approximation of point equilibria in MATSim

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4 Two MATSim contribs accompanying the hEART 2023 presentation “Improved  
5 precision in a heuristic for particle-based and stochastic dynamic traffic assignment”  
6 are presented. The underlying heuristic approximates point equilibria, i.e. reproducible  
7 “relaxed” states with minimal randomness in simulation setups that comprise the itera-  
8 tive “mobsim/scoring/replanning” logic of MATSim. An (already somewhat outdated)  
9 implementation of the method is on <https://github.com/gunnarfloetteroed>, it will be  
10 replaced with the version presented at hEART 2023 by the time of the conference.

11 The assignment method relies on the (approximate) minimization of an *equilibrium*  
12 *gap* that measures, for every agent, how much that agent could improve its score by  
13 changing its plan if all other agents kept their plans. The gap is always non-negative  
14 and zero at a (possibly non-existing) Nash equilibrium. The assignment method aims  
15 to reduce the gap in subsequent MATSim iterations, steering the simulation process  
16 towards a state that has a low gap, and hence may be deemed to be near an equilibrium  
17 point.

18 Apart from computing reproducible low-variability results, this approach provides  
19 a clear statement of what the simulation attempts to achieve when iterating the “MAT-  
20 Sim loop”. This may help to argue more convincingly in favor of an agent-based ap-  
21 proach rather than a traditional four-step/static-assignment model system: The single  
22 most important advantage of the latter is its ability to reproducibly compute point equi-  
23 libria.

24 The method is encoded in two contribs.

25 The `emulation` contrib is a successor of the `ier` (iterative emulated replanning) con-  
26 trib, which in turn succeeds the `psim` (pseudo simulation) contrib. All of these execute  
27 MATSim plans in a simplified mobility simulation where all movement-related perfor-  
28 mance measures are fixed (including exogenously set travel times). This allows to  
29 evaluate new plans in (possibly hypothetical) network conditions in a computationally  
30 efficient manner, prior to their computationally more involved execution in the mobil-  
31 ity simulation. This capability is central for computing the aforementioned equilibrium  
32 gap, in that it allows an agent to myopically optimize its travel plan in an otherwise un-  
33 changed physical world. (We also use it in the simulation-based optimization of vehicle  
34 sharing stations, where it allows to anticipate demand responses without permanently

35 re-running the simulation with new station configurations.) Specifications of new trans-  
36 port modes can be rather straightforwardly injected into the `emulation` contrib.

37 The `greedo` contrib depends on the `emulation` contrib and turns the MATSim loop  
38 into a point equilibrium approximation procedure. Plugging this contrib into MATSim  
39 requires minimally three lines of code:

```
40 Greedo greedo = new Greedo();  
41 ...  
42 Config config = ...;  
43 greedo.meet(config);  
44 ...  
45 Controler controler = ...;  
46 greedo.meet(controler);  
47 ...  
48 controler.run();  
49
```

50  
51 The method is illustrated in a multi-modal Greater Stockholm scenario that, in its  
52 maximal setup, comprises all-day choice of route, departure time, and mode. A uni-  
53 modal car-only version is used to benchmark the method in a more conventional setup.