

A large, abstract graphic composed of many thin, overlapping, wavy lines in shades of teal and green, creating a sense of depth and movement. The lines are arranged in a way that suggests a complex, organic shape, possibly resembling a stylized letter or a natural form like a leaf or a wave. The background is a light, pale blue with subtle, larger-scale wavy patterns.

Gurobi OptiMods

Painless Optimization Templates

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CO@Work 2024 - ZIB, Berlin

Why OptiMods?

Diving head-first into Optimization can be daunting!



You should be proficient in:

- maths
- modeling
- programming
- data science
- ...

enter Gurobi OptiMods!

What are OptiMods?

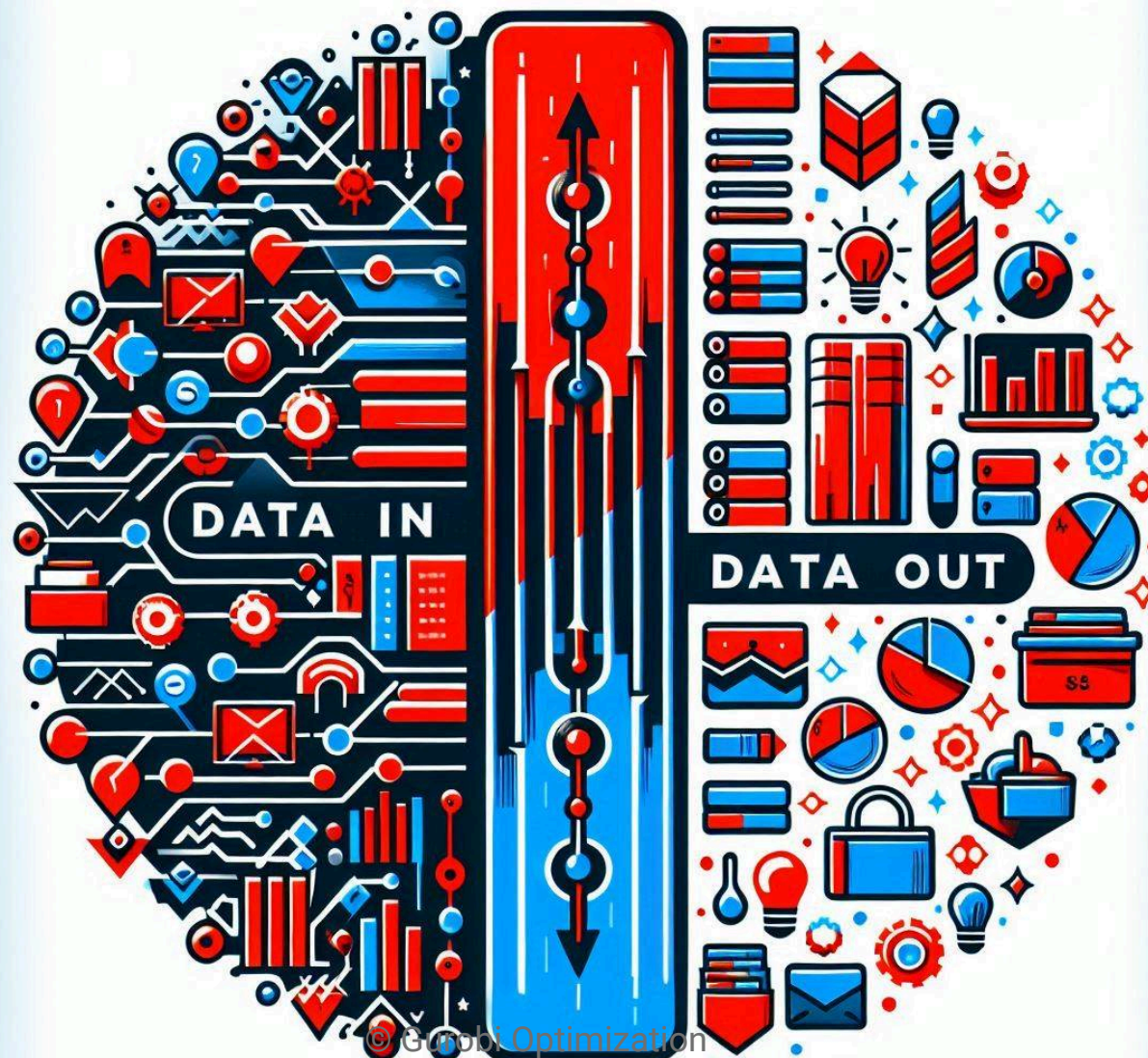
Each OptiMod solves a single, specific problem



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What are OptiMods?

Take input data in natural form, return solutions in natural form



What are OptiMods?

Using gurobipy and common Python packages (pandas, scipy, networkx, ...)



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What are OptiMods?

Solves an optimization problem using Gurobi without explicit modeling

The image features a glowing lightbulb at the center, surrounded by a complex network of nodes and edges. The background is filled with various mathematical formulas, including matrix operations, vector equations, and scalar expressions. The formulas are scattered around the lightbulb, creating a sense of intellectual activity and problem-solving.

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What are OptiMods?

Each OptiMod is comprehensively documented



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Drawbacks

Are OptiMods the solution to all your problems?



... unfortunately not

! **Simplicity comes at the price of flexibility!**



START

Installation

Usage

USER GUIDE

The OptiMods Gallery ^

 Maximum Bipartite
Matching

 Least Absolute Deviation
Regression

 Line Optimization in Public
Transport

 Metro Map: Computing an
Octilinear Graph
Representation

Minimum-Cost Flow

 Maximum Flow / Minimum
Cut

 Maximum Weighted
Independent Set/Clique

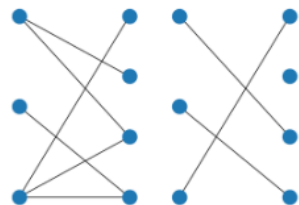
Optimal Power Flow v

Mean-Variance Portfolio

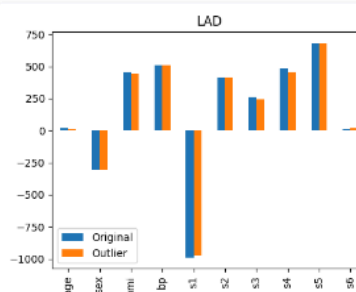
 Quadratic Unconstrained
Binary Optimization (QUBO)

Maximum Sharpe Ratio

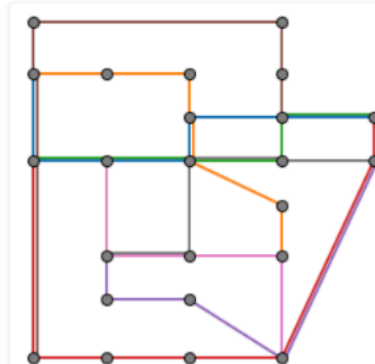
The OptiMods Gallery



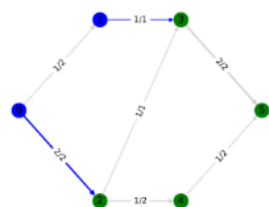
**Maximum Bipartite
Matching**



**Least Absolute Deviation
Regression**



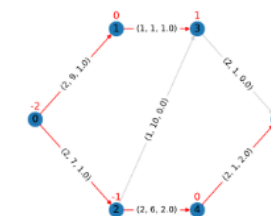
Line Optimization



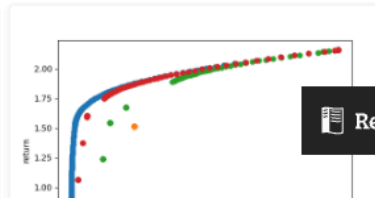
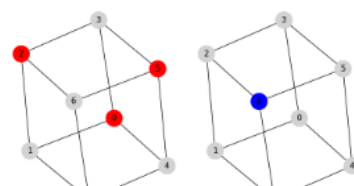
**Maximum-
Flow/Minimum-Cut**



Metro Map



Minimum-Cost Flow



Read the Docs

stable

Quick installation and usage instructions


1. Installation via PyPI:

```
1 pip install gurobi-optimods
```

2. Documentation on your selected Mod:

<https://gurobi-optimods.readthedocs.io>

3. Bring your data into the required format and run the Mod

 Documentation is a major part of the project

Demo: Min-Cost-Flow

Mathematical description

For a given graph G with set of vertices V and edges E . Each edge $(i, j) \in E$ has the following pair of attributes:

- cost: $c_{i,j} \in \mathbb{R}$
- capacity: $B_{i,j} \in \mathbb{R}$

Each vertex $i \in V$ has a demand $d_i \in \mathbb{R}$ that can be positive (requesting flow), negative (supplying flow), or zero (a transshipment node).

The problem can be stated as finding the flow with minimal total cost such that:

- demand at each vertex is met exactly and
- flow respects the capacity limit

Demo: Min-Cost-Flow

Input data

```

1 from gurobi_optimods import datasets
2
3 edge_data, node_data = datasets.simple_graph_pandas()
4 edge_data
5 node_data

```

		capacity		cost	demand	
source	target				0	-2
0	1	2	9		1	0
	2	2	7		2	-1
1	3	1	1		3	1
2	3	1	10		4	0
	4	2	6		5	2
3	5	2	1			
4	5	2	1			

Demo: Min-Cost-Flow Solution

```

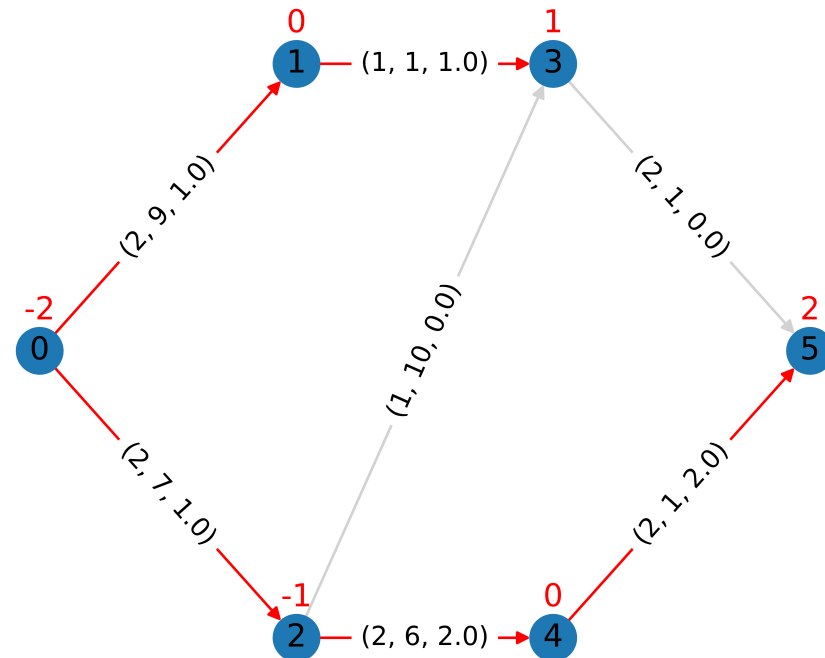
1 from gurobi_optimods import datasets
2 from gurobi_optimods.min_cost_flow import min_cost_flow_pandas
3 edge_data, node_data = datasets.simple_graph_pandas()
4 obj, sol = min_cost_flow_pandas(edge_data, node_data, verbose=False)
5 obj
6 sol

```

```

31.0      source  target
      0         1         1.0
      0         2         1.0
      1         3         1.0
      2         3         0.0
      2         4         2.0
      3         5         0.0
      4         5         2.0
Name: flow, dtype:
float64

```



Demo: Line Optimization in Public Transport

Abstract description

The line optimization problem in public transport is to choose a set of lines (routes in the transportation network) with associated frequencies (how often the lines are operated) such that a given transportation demand can be satisfied.

The total operational costs of the computed line plan has to be minimized.

Demo: Line Optimization in Public Transport

Input data

```
1 from gurobi_optimods import datasets
2
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
```

```
1 node_data.head(4)
```

	number	posx	posy
0	1	50000.0	510000.0
1	2	320000.0	510000.0
2	3	50000.0	440000.0
3	4	130000.0	440000.0

Demo: Line Optimization in Public Transport

Input data

```
1 from gurobi_optimods import datasets
2
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
```

```
1 edge_data.head(4)
```

	source	target	length	time
0	1	2	0.010	360
1	2	1	0.010	360
2	1	3	0.006	240
3	3	1	0.006	240

Demo: Line Optimization in Public Transport

Input data

```
1 from gurobi_optimods import datasets
2
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
```

```
1 line_data.head(4)
```

	linename	capacity	fix_cost	operating_cost
0	new7_B	600	15	3
1	new15_B	600	15	2
2	new23_B	600	15	6
3	new31_B	600	15	6

Demo: Line Optimization in Public Transport

Input data

```
1 from gurobi_optimods import datasets
2
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
```

```
1 linepath_data.head(4)
```

	linename	edge_source	edge_target
0	new7_B	1	2
1	new7_B	2	6
2	new7_B	6	8
3	new7_B	8	6

Demo: Line Optimization in Public Transport

Input data

```
1 from gurobi_optimods import datasets
2
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
```

```
1 demand_data.head(4)
```

	source	target	demand
0	1	2	5
1	1	3	5
2	1	4	25
3	1	5	10

Demo: Line Optimization in Public Transport Solution

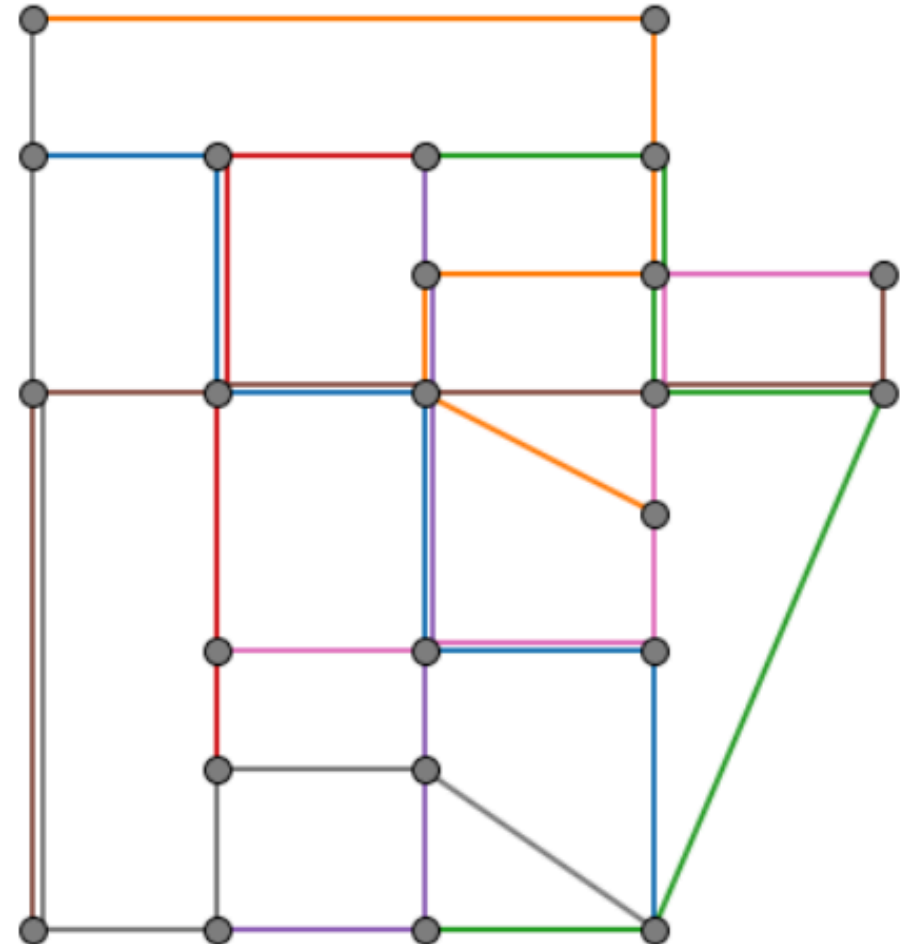
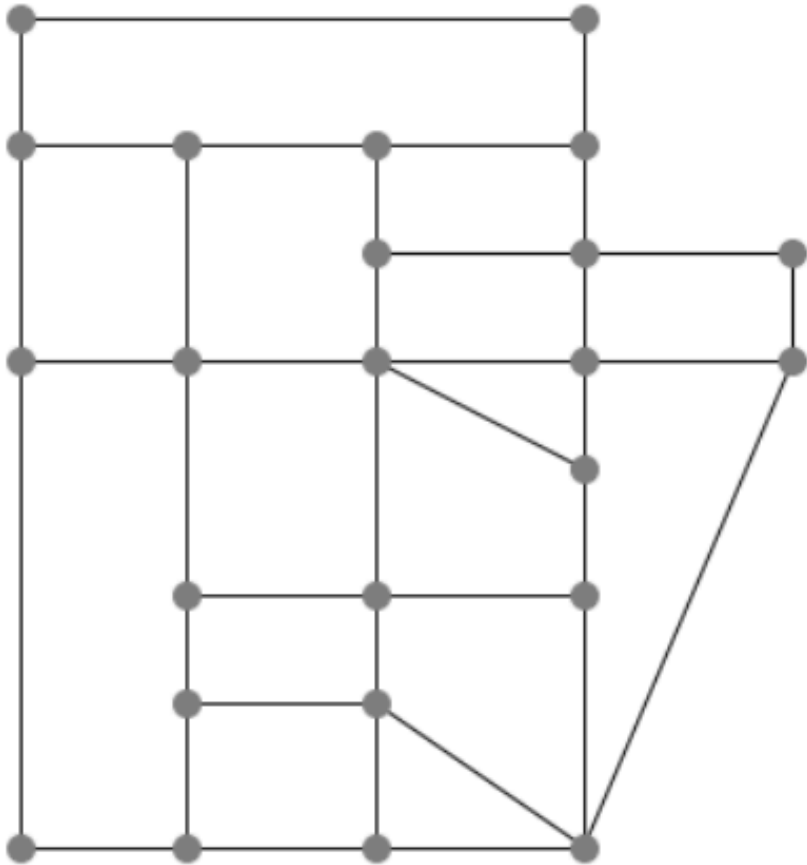
```
1 from gurobi_optimods import datasets
2 from gurobi_optimods.line_optimization import line_optimization
3 node_data, edge_data, line_data, linepath_data, demand_data = (
4     datasets.load_siouxfalls_network_data()
5 )
6 frequencies = [1, 3]
7 obj_cost, final_lines = line_optimization(
8     node_data,
9     edge_data,
10    line_data,
11    linepath_data,
12    demand_data,
13    frequencies,
14    verbose=False,
15 )
16 obj_cost
```

211.0

Demo: Line Optimization in Public Transport

Visualization

```
1 from gurobi_optimods.line_optimization import plot_lineplan
2 plot_lineplan(node_data, edge_data, linepath_data, final_lines)
```



Our latest Addition: Metro Map OptiMod


Computing an Octilinear Graph Representation



Summary

<https://github.com/Gurobi/gurobi-optimods>

- OptiMods provide data-driven APIs to solve common optimization problems
- Great way to make optimization more accessible
- Many optimization topics are still waiting to be turned into a new OptiMod
- Gurobi's other open-source Python packages:
 - [gurobipy-pandas](#): directly connect DataFrames with gurobipy
 - [gurobi-logtools](#): parse logfiles into DataFrames, e.g., for plotting
 - [gurobi-machinelearning](#): use trained models within optimization problems
 - [gurobi-modelanalyzer](#): compute numerical features of LPs

 Get involved and share your feedback and ideas!

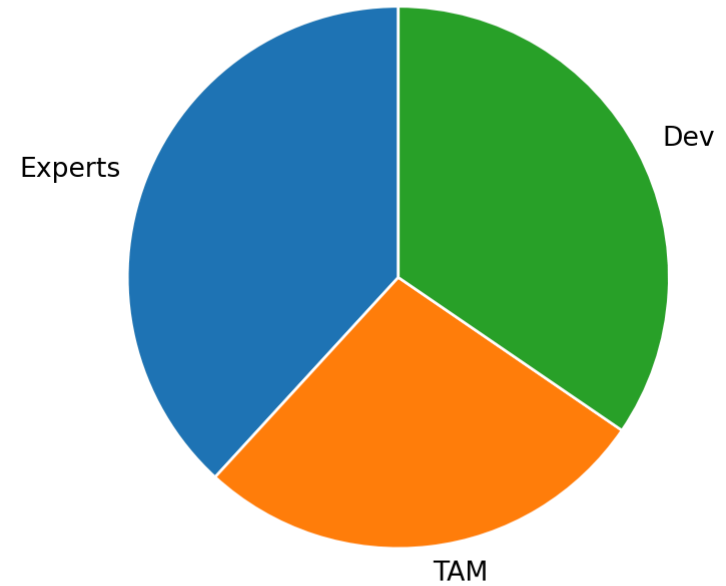
Gurobi as a Company

170 people all around the world working remotely

Distribution of Departments



Tech Teams

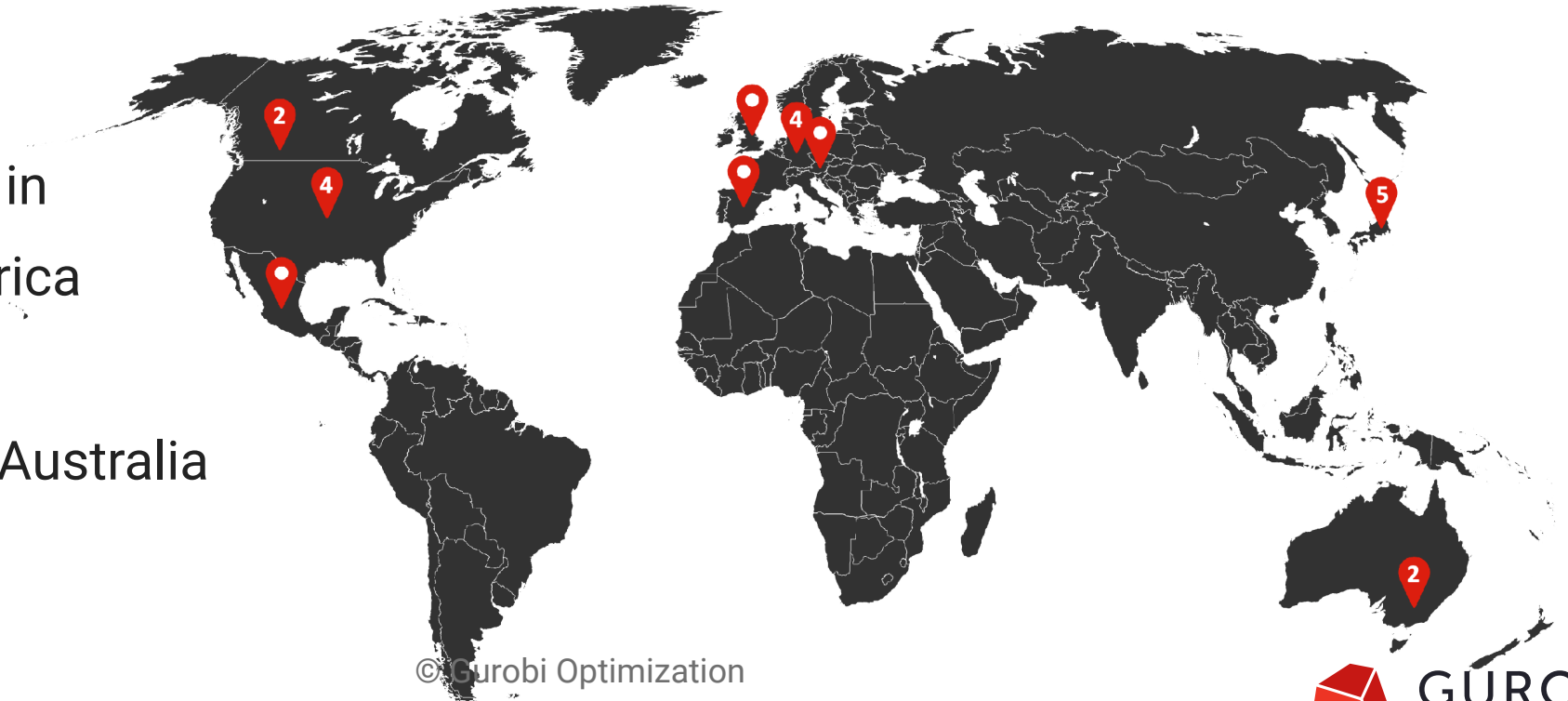


The Gurobi Experts Team

Supporting all our users around the clock and driving innovation



- Global team
- 7 people each in
 - North America
 - Europe
 - Japan and Australia



Recent Success Story

- Customer opens a ticket and asks for a parameter tuning:
 - starting point: **5% gap in 30 minutes**
- Goals:
 - 🏅 1% gap in 10 minutes would be great
 - 🏆 1% gap in 30 minutes would be good
 - 🕒 5% gap in 5 minutes would also be helpful
- After automatic parameter tuning for 35 hours on 12 machines:
 - 🏆 2% gap in 30 minutes
- **After Simran's reformulation idea:**
 - 🚀 **0.03% gap in 60 seconds!**

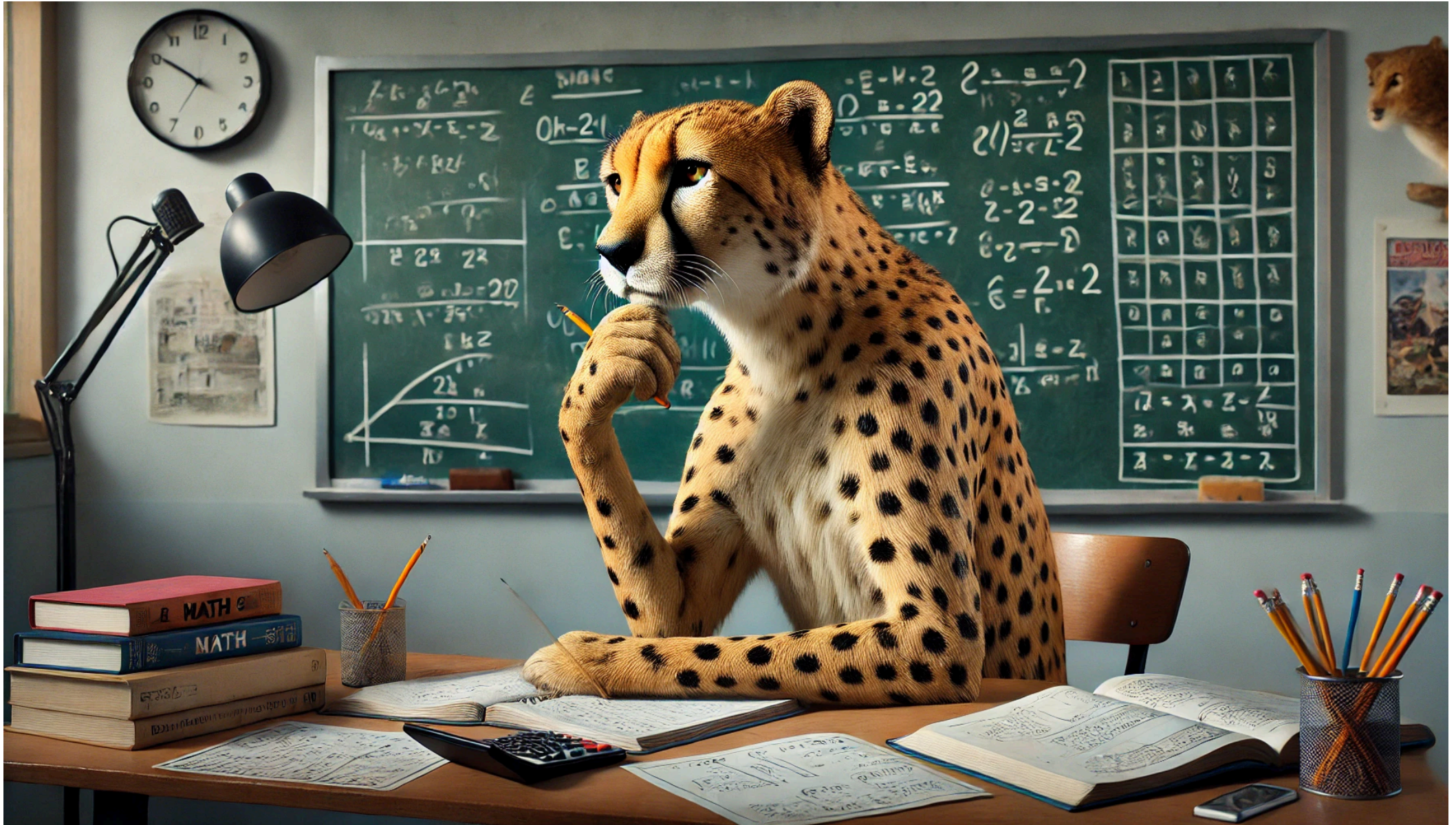


Raw Performance is good...



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... but don't forget about the Mathematics!



Happy Customers

- Many customers have a strong OR/math background
- We engage with the subject matter experts of the specific company
- Less tech-savvy customers are always super thankful for our help

We send out a quick survey after each ticket to ask how they liked our Support...

i For the last 2 years the rating has never been below 98% positive replies



How are you going to start your Journey?

