

# model

February 23, 2025

## 1 Use Gurobi Optimizer

```
[2]: import gurobipy as gp

try:
    m = gp.Model("test")
    print("Gurobi license is working!")
except gp.GurobiError as e:
    print(f"Error: {e}")
```

Gurobi license is working!

## 2 Centralized Optimization Model

The centralized model assumes global access to network information and uses Integer Linear Programming (ILP) to allocate resources optimally.

### 2.1 Objective Function

The objective function maximizes overall QoE:

$$\text{Maximize: } \sum_i w_i \cdot \sum_j q_j \cdot x_{ij}$$

where: -  $w_i$  is the weight of user  $i$  -  $q_j$  is the quality score of resolution  $j$  -  $x_{ij} \in \{0,1\}$  indicates whether user  $i$  is assigned resolution  $j$

### 2.2 Constraints

1. Each user is assigned at most one resolution:

$$\sum_j x_{ij} \leq 1, \quad \forall i$$

2. Resolutions cannot exceed user bandwidth capability:

$$x_{ij} = 0, \quad \forall j \text{ such that } B(j) > \text{max\_bitrate}(i)$$

3. Bandwidth usage on each link must not exceed capacity:

$$b_{lk} = \sum_j m_{lk,j} \cdot B(j), \quad b_{lk} \leq C_{lk}, \quad \forall (l, k)$$

4. If user  $i$  is assigned resolution  $j$ , all links along its path must transmit that resolution:

$$m_{lk,j} \geq x_{ij}, \quad \forall i, \forall j, \forall (l, k) \in \text{Path}(i)$$

## 2.3 Complexity Analysis

The centralized model must solve a global ILP problem. The complexity grows with the number of variables (V) and constraints (C):

$$O(2^V \cdot \text{poly}(C))$$

where V and C are determined as follows:

Variables:

$$\text{Variables: } n \cdot k + L + L \cdot k$$

Constraints:

$$\text{Constraints: } n + n \cdot k + 2L + n \cdot k \cdot P$$

where: - n is the number of users - k is the number of resolutions - L is the number of links - P is the average path length

This exponential complexity makes the centralized model impractical for large-scale systems.

```
[43]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    """
    - L:
    - B_l:
    - Q_l:
    - L_u:      (dict)
    - C_{i,j}:  (i->j)
    - w_u:
    """

    #      (Mbps)
    B_l = {
        '8K': 200,
```

```

        '4K': 45,
        '2K': 16,
        '1K': 8
    }

    #
    Q_1 = {
        '8K': 4,
        '4K': 3,
        '2K': 2,
        '1K': 1
    }

    #          L_u
    L_u = {
        'A': ['1K'],
        'B': ['1K', '2K', '4K'],
        'C': ['1K', '2K', '4K', '8K'],
        'D': ['1K', '2K', '4K', '8K'],
        'E': ['1K', '2K'],
        'F': ['1K', '2K', '4K'],
        'G': ['1K', '2K', '4K', '8K'],
        'H': ['1K', '2K', '4K', '8K'],
        'I': ['1K', '2K', '4K'],
        'J': ['1K', '2K', '4K']
    }

    #      C_{i,j} (i->j)
    C_ij = {
        ('Server', 'Core_Forwarder_1'): 500,
        ('Server', 'Core_Forwarder_2'): 500,
        ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
        ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
        ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
        ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
        ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
        ('Edge_Forwarder_A', 'Client_A'): 150,
        ('Edge_Forwarder_A', 'Client_B'): 150,
        ('Edge_Forwarder_B', 'Client_C'): 200,
        ('Edge_Forwarder_B', 'Client_D'): 200,
        ('Edge_Forwarder_C', 'Client_E'): 100,
        ('Edge_Forwarder_C', 'Client_F'): 100,
        ('Edge_Forwarder_C', 'Client_G'): 100,
        ('Edge_Forwarder_A', 'Client_H'): 100,
        ('Edge_Forwarder_B', 'Client_I'): 150,
        ('Edge_Forwarder_C', 'Client_J'): 200
    }

```

```

#       $w_u$  ( 1)
w_u = {
    'A': 1,
    'B': 1,
    'C': 1,
    'D': 1,
    'E': 1,
    'F': 1,
    'G': 1,
    'H': 1,
    'I': 1,
    'J': 1
}

return B_l, Q_l, L_u, C_ij, w_u

class ContinuousCentralizedModel:
    def __init__(self):
        #
        self.B_l, self.Q_l, self.L_u, self.C_ij, self.w_u = _
    define_problem_data()

    #       $U$  ( 'A', 'B', 'C'...)
    self.U = list(self.L_u.keys())
    #       $L$  ( '8K', '4K', '2K', '1K')
    self.L = list(self.B_l.keys())

    #      /      /
    self.S = {'Server'}
    self.F = {'Core_Forwarder_1', 'Core_Forwarder_2',
              'Edge_Forwarder_A', 'Edge_Forwarder_B', 'Edge_Forwarder_C'}
    #      "Client_X"
    self.U_nodes = {f"Client_{u}" for u in self.U}

    #       $V = S \cup F \cup U\_nodes$ 
    self.V = self.S | self.F | self.U_nodes

    #       $E$  ( $i \rightarrow j$ )
    self.E = list(self.C_ij.keys())

    #
    self.model = None
    self.x = None #  $x[i, l] \in [0, 1]$ 
    self.y = None #  $y[i, j, l] \in [0, 1]$ 
    self.b = None #  $b[i, j] = (l) B_l * y[i, j, l]$ 

```

```

self._create_model()

def _create_model(self):
    """ Gurobi (continuous in [0,1]) """
    self.model = gp.Model("ContinuousCentralizedModel")

    #  $x[i, l] \in [0, 1]$ 
    self.x = self.model.addVars(
        self.V, self.L,
        vtype=GRB.BINARY,
        name="x"
    )

    #  $y[i, j, l] \in [0, 1]$ 
    self.y = self.model.addVars(
        self.E, self.L,
        vtype=GRB.BINARY,
        name="y"
    )

    #  $b[i, j] = (i \rightarrow j)$ 
    self.b = self.model.addVars(
        self.E,
        vtype=GRB.CONTINUOUS,
        name="b"
    )

def set_objective(self):
    """ : max (u in U)  $w_u * (l \in L) Q_l * x[u, l]$  """
    obj_expr = gp.quicksum(
        self.w_u[u] * self.Q_l[l] * self.x[f"Client_{u}", l], 1
        for u in self.U
        for l in self.L
    )
    self.model.setObjective(obj_expr, GRB.MAXIMIZE)

def add_constraints(self):
    """ """
    # 1) :  $x[s, l] = 1$ 
    for s in self.S:
        for l in self.L:
            self.model.addConstr(
                self.x[s, l] == 1,
                name=f"server_owns_{s}_{l}"
            )

```

```

# 2a)      :  l   L_u[u],   x[u,l] = 0
for u in self.U:
    allowed_set = set(self.L_u[u]) #
    for l in self.L:
        if l not in allowed_set:
            self.model.addConstr(
                self.x[f"Client_{u}", l] == 0,
                name=f"user_cap_{u}_{l}"
            )

# 2b)      1   :  (l in L) x[u,l]   1
for u in self.U:
    self.model.addConstr(
        gp.quicksum(self.x[f"Client_{u}", l] for l in self.L) <= 1,
        name=f"user_single_res_{u}"
    )

# 3a)      :  y[i,j,l]   x[i,l]
for (i, j) in self.E:
    for l in self.L:
        self.model.addConstr(
            self.y[i, j, l] <= self.x[i, l],
            name=f"trans_logic_1_{i}_{j}_{l}"
        )

# 3b)      l,      :  x[i,l]   (p->i) y[p,i,l]
for node in (self.F | self.U_nodes):
    in_edges = [(p, node) for (p, q) in self.E if q == node]
    for l in self.L:
        self.model.addConstr(
            self.x[node, l] <= gp.quicksum(self.y[p, node, l] for (p,
↪_) in in_edges),
            name=f"trans_logic_2_{node}_{l}"
        )

# 4)      :  (l) B_l * y[i,j,l]   C_{i,j}
for (i, j), cap in self.C_ij.items():
    self.model.addConstr(
        gp.quicksum(self.B_l[l] * self.y[i, j, l] for l in self.L) <=
↪cap,
        name=f"capacity_{i}_{j}"
    )

# 5) b[i,j] = (l) B_l * y[i,j,l]
for (i, j) in self.E:
    self.model.addConstr(

```

```

        self.b[i, j] == gp.quicksum(self.B_l[l] * self.y[i, j, l] for l_
↪in self.L),
        name=f"bandwidth_usage_{i}_{j}"
    )

    def optimize(self):
        """ """
        self.model.optimize()

    def print_results(self):
        """ """
        if self.model.status == GRB.OPTIMAL:
            print("\nOptimal solution found!")
            print(f"Optimal Objective Value = {self.model.ObjVal}\n")

            # x[u,l]
            results = []
            for u in self.U:
                for l in self.L:
                    val = self.x[f"Client_{u}", l].X
                    if val > 1e-6:
                        results.append({
                            'User': u,
                            'Resolution(l)': l,
                            'x[u,l]': round(val, 4),
                            'Bandwidth(B_l)': self.B_l[l]
                        })
            df_results = pd.DataFrame(results)
            print("User resolution selection (continuous x):")
            print(df_results)

            #
            link_usage = []
            for (i, j) in self.E:
                used = self.b[i, j].X
                cap = self.C_ij[(i, j)]
                link_usage.append({
                    'From': i,
                    'To': j,
                    'Usage(Mbps)': round(used, 2),
                    'Capacity(Mbps)': cap,
                    'Utilization(%)': round(used / cap * 100, 2) if cap > 1e-9_
↪else 0
                })
            df_links = pd.DataFrame(link_usage)
            print("\nLink bandwidth usage:")
            print(df_links)

```

```

        self.plot_network(df_links)
    else:
        print("No optimal solution found. Status:", self.model.status)

def plot_network(self, df_links):
    fig, ax = plt.subplots(figsize=(15, 10))

    pos = {
        'Server': (0.2, 0.8),
        'Core_Forwarder_1': (0.4, 0.7),
        'Core_Forwarder_2': (0.6, 0.7),
        'Edge_Forwarder_A': (0.3, 0.5),
        'Edge_Forwarder_B': (0.5, 0.5),
        'Edge_Forwarder_C': (0.7, 0.5),
        'Client_A': (0.1, 0.3),
        'Client_B': (0.2, 0.3),
        'Client_C': (0.4, 0.3),
        'Client_D': (0.5, 0.3),
        'Client_E': (0.6, 0.3),
        'Client_F': (0.7, 0.3),
        'Client_G': (0.8, 0.3),
        'Client_H': (0.3, 0.3),
        'Client_I': (0.45, 0.2),
        'Client_J': (0.65, 0.2),
    }

    user_choices = {}
    for u in self.U:
        chosen_info = []
        for l in self.L:
            val = self.x[f"Client_{u}", l].X
            if val > 1e-6:
                chosen_info.append(f"{l}({val:.2f})")
        user_choices[u] = chosen_info

    utilization_dict = {
        (row['From'], row['To']): row['Utilization(%)']
        for _, row in df_links.iterrows()
    }

    def get_color(util_pct):
        return plt.cm.RdYlGn_r(min(util_pct, 100) / 100.0)

    # Draw nodes
    for node, (x_c, y_c) in pos.items():
        if node.startswith("Client_"):

```



```

ax.plot(x_c, y_c, 'gs', markersize=20)
ax.text(x_c, y_c - 0.02, node, ha='center')

uid = node.split('_')[1]
allowed = self.L_u[uid]
chosen = user_choices[uid]
chosen_str = ", ".join(chosen) if chosen else "N/A"

ax.text(
    x_c, y_c - 0.05,
    f"Allowed: {'', '.join(allowed)}\nChosen: {chosen_str}",
    ha='center', fontsize=8
)
elif "Forwarder" in node:
    ax.plot(x_c, y_c, 'bo', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')
else:
    ax.plot(x_c, y_c, 'rd', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')

# Draw links with arrows
for (i, j), util_pct in utilization_dict.items():
    if i not in pos or j not in pos:
        continue
    x1, y1 = pos[i]
    x2, y2 = pos[j]

    # Calculate arrow position (70% along the line)
    arrow_pos = 0.7
    arrow_x = x1 + arrow_pos * (x2 - x1)
    arrow_y = y1 + arrow_pos * (y2 - y1)

    # Draw line
    ax.plot([x1, x2], [y1, y2], '-', color=get_color(util_pct),
    ↪linewidth=2)

    # Add arrow
    dx = x2 - x1
    dy = y2 - y1
    ax.arrow(arrow_x, arrow_y, dx*0.15, dy*0.15,
             head_width=0.02, head_length=0.02,
             fc=get_color(util_pct), ec=get_color(util_pct))

    cap_ij = self.C_ij[(i, j)]
    label = f"{round(util_pct,1)}%\n({cap_ij} Mbps)"

    # Adjust label positions for specific edges to avoid overlap

```

```

        if (i, j) == ('Core_Forwarder_1', 'Edge_Forwarder_B'):
            label_offset = -0.03 # Move label down
        elif (i, j) == ('Core_Forwarder_2', 'Edge_Forwarder_A'):
            label_offset = 0.05 # Move label up
        else:
            label_offset = 0.02 # Default offset

        ax.text((x1 + x2)/2, (y1 + y2)/2 + label_offset, label)

    # Color bar
    norm = plt.Normalize(0, 100)
    sm = plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
    sm.set_array([])
    plt.colorbar(sm, ax=ax, label='Link Utilization(%)')

    # Resolution legend
    ax.text(0.2, 0.85, 'Resolutions:', ha='left')
    for idx, l in enumerate(self.L):
        ax.text(0.2, 0.83 - idx * 0.02, f"{l} -- {self.B_l[l]} Mbps",
        ↪ha='left')

    ax.set_title('Centralized Model: Network Topology and Link Utilization')
    ax.axis('off')
    plt.tight_layout()
    plt.show()

def run_optimization():
    model = ContinuousCentralizedModel()
    model.set_objective()
    model.add_constraints()
    model.optimize()
    model.print_results()

if __name__ == "__main__":
    run_optimization()

```

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 185 rows, 149 columns and 470 nonzeros

Model fingerprint: 0xab05749e

Variable types: 17 continuous, 132 integer (132 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 5e+02]

Found heuristic solution: objective -0.0000000

Presolve removed 185 rows and 149 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)

Thread count was 1 (of 64 available processors)

Solution count 2: 29 -0

Optimal solution found (tolerance 1.00e-04)

Best objective 2.9000000000000e+01, best bound 2.9000000000000e+01, gap 0.0000%

Optimal solution found!

Optimal Objective Value = 29.0

User resolution selection (continuous x):

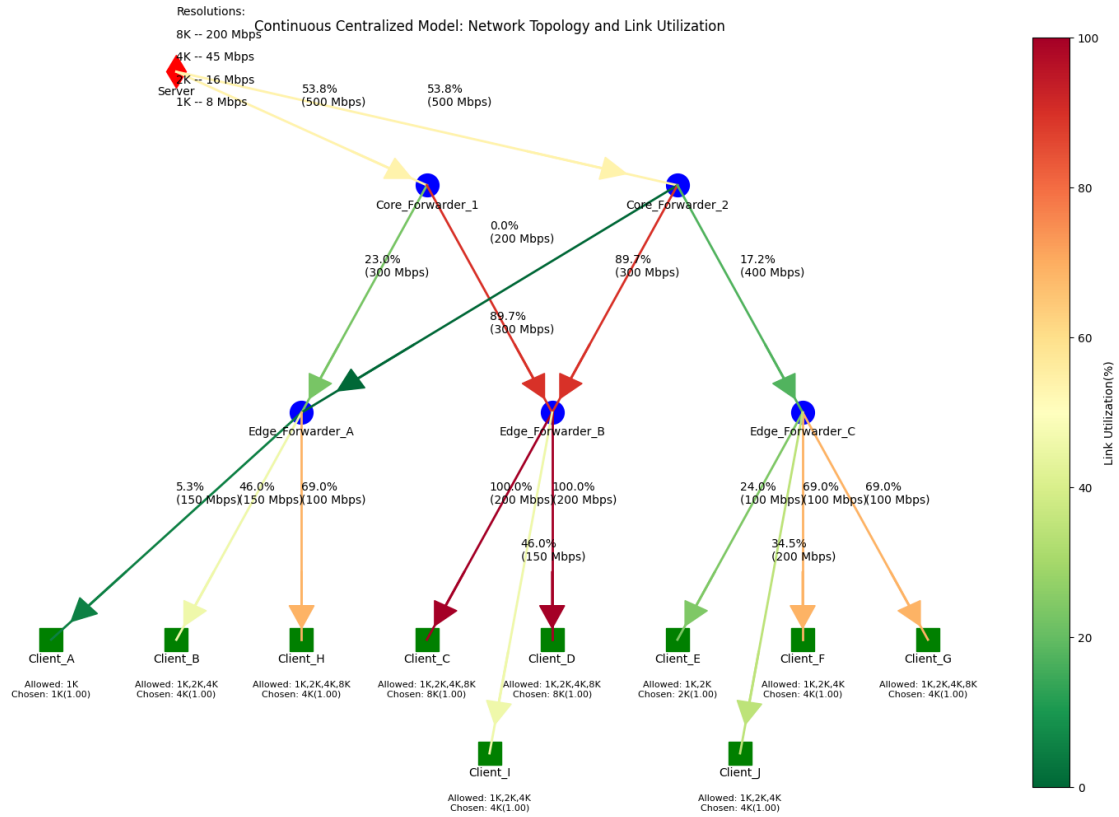
	User Resolution(l)	x[u,l]	Bandwidth(B_l)
0	A	1K	1.0
1	B	4K	1.0
2	C	8K	1.0
3	D	8K	1.0
4	E	2K	1.0
5	F	4K	1.0
6	G	4K	1.0
7	H	4K	1.0
8	I	4K	1.0
9	J	4K	1.0

Link bandwidth usage:

	From	To	Usage(Mbps)	Capacity(Mbps)	\
0	Server	Core_Forwarder_1	269.0	500	
1	Server	Core_Forwarder_2	269.0	500	
2	Core_Forwarder_1	Edge_Forwarder_A	69.0	300	
3	Core_Forwarder_1	Edge_Forwarder_B	269.0	300	
4	Core_Forwarder_2	Edge_Forwarder_A	0.0	200	
5	Core_Forwarder_2	Edge_Forwarder_B	269.0	300	
6	Core_Forwarder_2	Edge_Forwarder_C	69.0	400	
7	Edge_Forwarder_A	Client_A	8.0	150	
8	Edge_Forwarder_A	Client_B	69.0	150	
9	Edge_Forwarder_B	Client_C	200.0	200	
10	Edge_Forwarder_B	Client_D	200.0	200	

11	Edge_Forwarder_C	Client_E	24.0	100
12	Edge_Forwarder_C	Client_F	69.0	100
13	Edge_Forwarder_C	Client_G	69.0	100
14	Edge_Forwarder_A	Client_H	69.0	100
15	Edge_Forwarder_B	Client_I	69.0	150
16	Edge_Forwarder_C	Client_J	69.0	200

	Utilization(%)
0	53.80
1	53.80
2	23.00
3	89.67
4	0.00
5	89.67
6	17.25
7	5.33
8	46.00
9	100.00
10	100.00
11	24.00
12	69.00
13	69.00
14	69.00
15	46.00
16	34.50



```
[1]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    """
    Define the data required for the system model:
    - L: Set of resolutions
    - B_l: Bandwidth demand for each resolution
    - Q_l: Quality score for each resolution
    - L_u: Set of resolutions selectable by users (dict)
    - C_{i,j}: Capacity of link (i->j)
    - w_u: User weights
    """

    # Resolution bandwidth demand (Mbps)
    B_l = {
        '8K': 200,
        '4K': 45,
```

```

    '2K': 16,
    '1K': 8
}

# Resolution quality scores
Q_l = {
    '8K': 4,
    '4K': 3,
    '2K': 2,
    '1K': 1
}

# Set of resolutions selectable by users L_u
L_u = {
    'A': ['1K'],
    'B': ['1K', '2K', '4K'],
    'C': ['1K', '2K', '4K', '8K'],
    'D': ['1K', '2K', '4K', '8K'],
    'E': ['1K', '2K'],
    'F': ['1K', '2K', '4K'],
    'G': ['1K', '2K', '4K', '8K'],
    'H': ['1K', '2K', '4K', '8K'],
    'I': ['1K', '2K', '4K'],
    'J': ['1K', '2K', '4K']
}

# Link capacity C_{i,j} (i->j)
C_ij = {
    ('Server', 'Core_Forwarder_1'): 500,
    ('Server', 'Core_Forwarder_2'): 500,
    ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
    ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
    ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
    ('Edge_Forwarder_A', 'Client_A'): 150,
    ('Edge_Forwarder_A', 'Client_B'): 150,
    ('Edge_Forwarder_B', 'Client_C'): 200,
    ('Edge_Forwarder_B', 'Client_D'): 200,
    ('Edge_Forwarder_C', 'Client_E'): 100,
    ('Edge_Forwarder_C', 'Client_F'): 100,
    ('Edge_Forwarder_C', 'Client_G'): 100,
    ('Edge_Forwarder_A', 'Client_H'): 100,
    ('Edge_Forwarder_B', 'Client_I'): 150,
    ('Edge_Forwarder_C', 'Client_J'): 200
}

```

```

# User weights w_u (default is 1)
w_u = {
    'A': 1,
    'B': 1,
    'C': 1,
    'D': 1,
    'E': 1,
    'F': 1,
    'G': 1,
    'H': 1,
    'I': 1,
    'J': 1
}

return B_l, Q_l, L_u, C_ij, w_u

class ContinuousCentralizedModel:
    def __init__(self):
        # Load data
        self.B_l, self.Q_l, self.L_u, self.C_ij, self.w_u = _
    ↪ define_problem_data()

        # User set U (e.g., 'A', 'B', 'C'...)
        self.U = list(self.L_u.keys())
        # Resolution set L (e.g., '8K', '4K', '2K', '1K')
        self.L = list(self.B_l.keys())

        # Define server / forwarder / user node sets
        self.S = {'Server'}
        self.F = {'Core_Forwarder_1', 'Core_Forwarder_2',
                  'Edge_Forwarder_A', 'Edge_Forwarder_B', 'Edge_Forwarder_C'}
        # User nodes written as "Client_X"
        self.U_nodes = {f"Client_{u}" for u in self.U}

        # Entire node set V = S ∪ F ∪ U_nodes
        self.V = self.S | self.F | self.U_nodes

        # Link set E (i→j)
        self.E = list(self.C_ij.keys())

        # Prepare model and variables
        self.model = None
        self.x = None # x[i,l] [0,1]
        self.y = None # y[i,j,l] [0,1]
        self.b = None # b[i,j] = (l) B_l * y[i,j,l]

```

```

self._create_model()

def _create_model(self):
    """Create Gurobi model and variables (continuous in [0,1])"""
    self.model = gp.Model("ContinuousCentralizedModel")

    #  $x[i, l] \in [0, 1]$ 
    self.x = self.model.addVars(
        self.V, self.L,
        vtype=GRB.BINARY,
        name="x"
    )

    #  $y[i, j, l] \in [0, 1]$ 
    self.y = self.model.addVars(
        self.E, self.L,
        vtype=GRB.BINARY,
        name="y"
    )

    #  $b[i, j]$  = actual bandwidth usage of link ( $i \rightarrow j$ )
    self.b = self.model.addVars(
        self.E,
        vtype=GRB.CONTINUOUS,
        name="b"
    )

def set_objective(self):
    """Objective function:  $\max (u \in U) w_u * (l \in L) Q_l * x[u, l]$ ."""
    obj_expr = gp.quicksum(
        self.w_u[u] * self.Q_l[l] * self.x[f"Client_{u}", l]
        for u in self.U
        for l in self.L
    )
    self.model.setObjective(obj_expr, GRB.MAXIMIZE)

def add_constraints(self):
    """Add all linear constraints corresponding to the system model."""
    # 1) Server nodes possess all resolutions:  $x[s, l] = 1$ 
    for s in self.S:
        for l in self.L:
            self.model.addConstr(
                self.x[s, l] == 1,
                name=f"server_owns_{s}_{l}"
            )

    # 2a) User capability limit: if  $l$  is not in  $L_u[u]$ , then  $x[u, l] = 0$ 

```



```

    for u in self.U:
        allowed_set = set(self.L_u[u]) # Resolutions selectable by this_
↪user
        for l in self.L:
            if l not in allowed_set:
                self.model.addConstr(
                    self.x[f"Client_{u}", l] == 0,
                    name=f"user_cap_{u}_{l}"
                )

# 2b) Each user selects at most 1 resolution: (l in L) x[u,l] 1
for u in self.U:
    self.model.addConstr(
        gp.quicksum(self.x[f"Client_{u}", l] for l in self.L) <= 1,
        name=f"user_single_res_{u}"
    )

# 3a) Transmission logic: y[i,j,l]  x[i,l]
for (i, j) in self.E:
    for l in self.L:
        self.model.addConstr(
            self.y[i, j, l] <= self.x[i, l],
            name=f"trans_logic_1_{i}_{j}_{l}"
        )

# 3b) Non-server nodes need at least one upstream transmission to_
↪possess l: x[i,l] (p->i) y[p,i,l]
for node in (self.F | self.U_nodes):
    in_edges = [(p, node) for (p, q) in self.E if q == node]
    for l in self.L:
        self.model.addConstr(
            self.x[node, l] <= gp.quicksum(self.y[p, node, l] for (p,
↪_) in in_edges),
            name=f"trans_logic_2_{node}_{l}"
        )

# 4) Bandwidth capacity: (l) B_l * y[i,j,l] C_{i,j}
for (i, j), cap in self.C_ij.items():
    self.model.addConstr(
        gp.quicksum(self.B_l[l] * self.y[i, j, l] for l in self.L) <=
↪cap,
        name=f"capacity_{i}_{j}"
    )

# 5) b[i,j] = (l) B_l * y[i,j,l]
for (i, j) in self.E:
    self.model.addConstr(

```

```

        self.b[i, j] == gp.quicksum(self.B_l[l] * self.y[i, j, l] for l in self.L),
        name=f"bandwidth_usage_{i}_{j}"
    )

    def optimize(self):
        """Perform optimization"""
        self.model.optimize()

    def print_results(self):
        """Print and visualize results"""
        if self.model.status == GRB.OPTIMAL:
            print("\nOptimal solution found!")
            print(f"Optimal Objective Value = {self.model.ObjVal}\n")

            # Check the values of x[u,l] for users
            results = []
            for u in self.U:
                for l in self.L:
                    val = self.x[f"Client_{u}", l].X
                    if val > 1e-6:
                        results.append({
                            'User': u,
                            'Resolution(l)': l,
                            'x[u,l]': round(val, 4),
                            'Bandwidth(B_l)': self.B_l[l]
                        })
            df_results = pd.DataFrame(results)
            print("User resolution selection (continuous x):")
            print(df_results)

            # Link bandwidth usage
            link_usage = []
            for (i, j) in self.E:
                used = self.b[i, j].X
                cap = self.C_ij[(i, j)]
                link_usage.append({
                    'From': i,
                    'To': j,
                    'Usage(Mbps)': round(used, 2),
                    'Capacity(Mbps)': cap,
                    'Utilization(%)': round(used / cap * 100, 2) if cap > 1e-9
                })
            df_links = pd.DataFrame(link_usage)
            print("\nLink bandwidth usage:")
            print(df_links)

```

```

        self.plot_network(df_links)
    else:
        print("No optimal solution found. Status:", self.model.status)

def plot_network(self, df_links):
    fig, ax = plt.subplots(figsize=(15, 10))

    pos = {
        'Server': (0.2, 0.8),
        'Core_Forwarder_1': (0.4, 0.7),
        'Core_Forwarder_2': (0.6, 0.7),
        'Edge_Forwarder_A': (0.3, 0.5),
        'Edge_Forwarder_B': (0.5, 0.5),
        'Edge_Forwarder_C': (0.7, 0.5),
        'Client_A': (0.1, 0.3),
        'Client_B': (0.2, 0.3),
        'Client_C': (0.4, 0.3),
        'Client_D': (0.5, 0.3),
        'Client_E': (0.6, 0.3),
        'Client_F': (0.7, 0.3),
        'Client_G': (0.8, 0.3),
        'Client_H': (0.3, 0.3),
        'Client_I': (0.45, 0.2),
        'Client_J': (0.65, 0.2),
    }

    user_choices = {}
    for u in self.U:
        chosen_info = []
        for l in self.L:
            val = self.x[f"Client_{u}", l].X
            if val > 1e-6:
                chosen_info.append(f"{l}({val:.2f})")
        user_choices[u] = chosen_info

    utilization_dict = {
        (row['From'], row['To']): row['Utilization(%)']
        for _, row in df_links.iterrows()
    }

    def get_color(util_pct):
        return plt.cm.RdYlGn_r(min(util_pct, 100) / 100.0)

    # Draw nodes
    for node, (x_c, y_c) in pos.items():
        if node.startswith("Client_"):

```

```

ax.plot(x_c, y_c, 'gs', markersize=20)
ax.text(x_c, y_c - 0.02, node, ha='center')

uid = node.split('_')[1]
allowed = self.L_u[uid]
chosen = user_choices[uid]
chosen_str = ", ".join(chosen) if chosen else "N/A"

ax.text(
    x_c, y_c - 0.05,
    f"Allowed: {'', '.join(allowed)}\nChosen: {chosen_str}",
    ha='center', fontsize=8
)
elif "Forwarder" in node:
    ax.plot(x_c, y_c, 'bo', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')
else:
    ax.plot(x_c, y_c, 'rd', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')

# Draw links with arrows
for (i, j), util_pct in utilization_dict.items():
    if i not in pos or j not in pos:
        continue
    x1, y1 = pos[i]
    x2, y2 = pos[j]

    # Calculate arrow position (70% along the line)
    arrow_pos = 0.7
    arrow_x = x1 + arrow_pos * (x2 - x1)
    arrow_y = y1 + arrow_pos * (y2 - y1)

    # Draw line
    ax.plot([x1, x2], [y1, y2], '-', color=get_color(util_pct),
    ↪linewidth=2)

    # Add arrow
    dx = x2 - x1
    dy = y2 - y1
    ax.arrow(arrow_x, arrow_y, dx*0.15, dy*0.15,
             head_width=0.02, head_length=0.02,
             fc=get_color(util_pct), ec=get_color(util_pct))

    cap_ij = self.C_ij[(i, j)]
    label = f"{round(util_pct,1)}%\n({cap_ij} Mbps)"

    # Adjust label positions for specific edges to avoid overlap

```

```

        if (i, j) == ('Core_Forwarder_1', 'Edge_Forwarder_B'):
            label_offset = -0.03 # Move label down
        elif (i, j) == ('Core_Forwarder_2', 'Edge_Forwarder_A'):
            label_offset = 0.05 # Move label up
        else:
            label_offset = 0.02 # Default offset

        ax.text((x1 + x2)/2, (y1 + y2)/2 + label_offset, label)

    # Color bar
    norm = plt.Normalize(0, 100)
    sm = plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
    sm.set_array([])
    plt.colorbar(sm, ax=ax, label='Link Utilization(%)')

    # Resolution legend
    ax.text(0.2, 0.85, 'Resolutions:', ha='left')
    for idx, l in enumerate(self.L):
        ax.text(0.2, 0.83 - idx * 0.02, f"{l} -- {self.B_l[l]} Mbps",
        ↪ha='left')

    ax.set_title('Centralized Model: Network Topology and Link Utilization')
    ax.axis('off')
    plt.tight_layout()
    plt.show()

def run_optimization():
    model = ContinuousCentralizedModel()
    model.set_objective()
    model.add_constraints()
    model.optimize()
    model.print_results()

if __name__ == "__main__":
    run_optimization()

```

Set parameter LicenseID to value 2601434

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 185 rows, 149 columns and 470 nonzeros

Model fingerprint: 0xced192d

Variable types: 17 continuous, 132 integer (132 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 5e+02]

Found heuristic solution: objective -0.0000000

Presolve removed 185 rows and 149 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)

Thread count was 1 (of 64 available processors)

Solution count 2: 29 -0

Optimal solution found (tolerance 1.00e-04)

Best objective 2.9000000000000e+01, best bound 2.9000000000000e+01, gap 0.0000%

Optimal solution found!

Optimal Objective Value = 29.0

User resolution selection (continuous x):

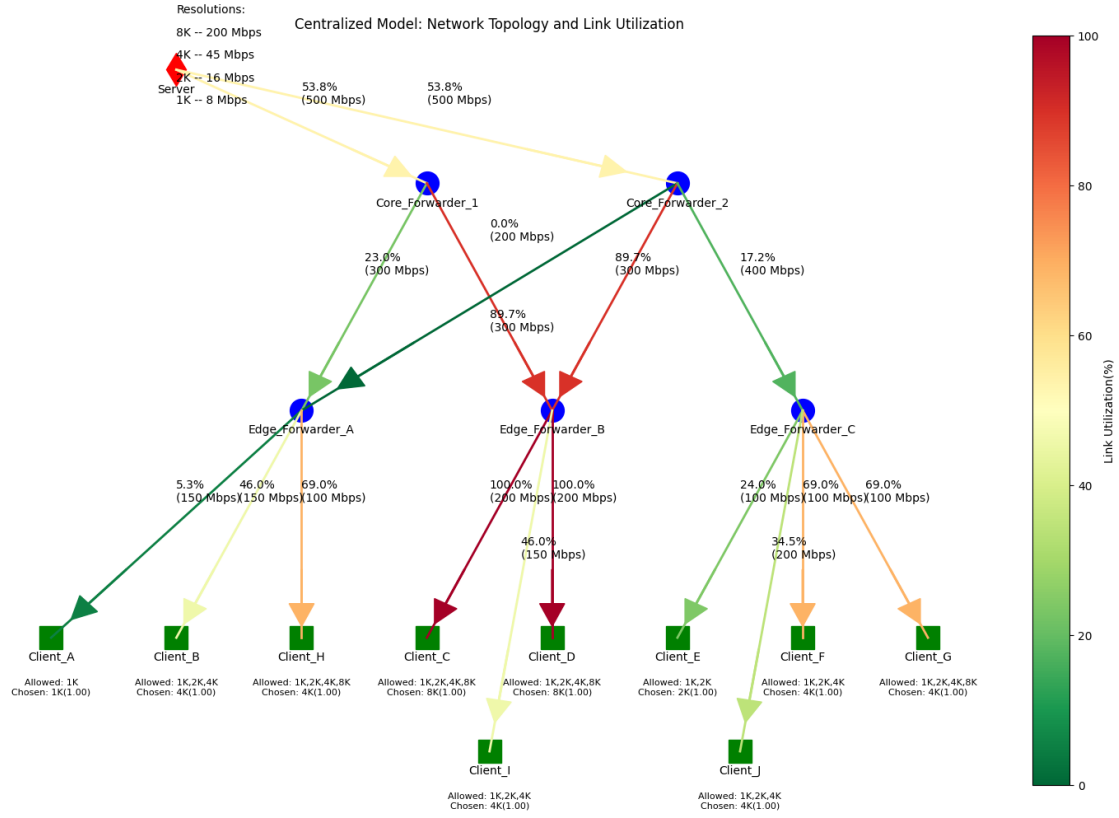
	User Resolution(l)	x[u,l]	Bandwidth(B_l)
0	A	1K	1.0
1	B	4K	1.0
2	C	8K	1.0
3	D	8K	1.0
4	E	2K	1.0
5	F	4K	1.0
6	G	4K	1.0
7	H	4K	1.0
8	I	4K	1.0
9	J	4K	1.0

Link bandwidth usage:

	From	To	Usage(Mbps)	Capacity(Mbps)	\
0	Server	Core_Forwarder_1	269.0	500	
1	Server	Core_Forwarder_2	269.0	500	
2	Core_Forwarder_1	Edge_Forwarder_A	69.0	300	
3	Core_Forwarder_1	Edge_Forwarder_B	269.0	300	
4	Core_Forwarder_2	Edge_Forwarder_A	0.0	200	
5	Core_Forwarder_2	Edge_Forwarder_B	269.0	300	
6	Core_Forwarder_2	Edge_Forwarder_C	69.0	400	
7	Edge_Forwarder_A	Client_A	8.0	150	
8	Edge_Forwarder_A	Client_B	69.0	150	
9	Edge_Forwarder_B	Client_C	200.0	200	

10	Edge_Forwarder_B	Client_D	200.0	200
11	Edge_Forwarder_C	Client_E	24.0	100
12	Edge_Forwarder_C	Client_F	69.0	100
13	Edge_Forwarder_C	Client_G	69.0	100
14	Edge_Forwarder_A	Client_H	69.0	100
15	Edge_Forwarder_B	Client_I	69.0	150
16	Edge_Forwarder_C	Client_J	69.0	200

Utilization(%)	
0	53.80
1	53.80
2	23.00
3	89.67
4	0.00
5	89.67
6	17.25
7	5.33
8	46.00
9	100.00
10	100.00
11	24.00
12	69.00
13	69.00
14	69.00
15	46.00
16	34.50



```
[45]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    """
    - L      B_l
    - Q_l
    - U      L_u
    - C_ij
    - w_u
    """
    # (Mbps)
    B_l = {
        '8K': 200,
        '4K': 45,
        '2K': 16,
        '1K': 8
    }
```



```

#
Q_1 = {
    '8K': 4,
    '4K': 3,
    '2K': 2,
    '1K': 1
}

#          L_u
L_u = {
    'A': ['1K'],
    'B': ['1K', '2K', '4K'],
    'C': ['1K', '2K', '4K', '8K'],
    'D': ['1K', '2K', '4K', '8K'],
    'E': ['1K', '2K'],
    'F': ['1K', '2K', '4K'],
    'G': ['1K', '2K', '4K', '8K'],
    'H': ['1K', '2K', '4K', '8K'],
    'I': ['1K', '2K', '4K'],
    'J': ['1K', '2K', '4K']
}

#          C_{i,j} (i->j)
C_ij = {
    ('Server', 'Core_Forwarder_1'): 500,
    ('Server', 'Core_Forwarder_2'): 500,
    ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
    ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
    ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
    ('Edge_Forwarder_A', 'Client_A'): 150,
    ('Edge_Forwarder_A', 'Client_B'): 150,
    ('Edge_Forwarder_B', 'Client_C'): 200,
    ('Edge_Forwarder_B', 'Client_D'): 200,
    ('Edge_Forwarder_C', 'Client_E'): 100,
    ('Edge_Forwarder_C', 'Client_F'): 100,
    ('Edge_Forwarder_C', 'Client_G'): 100,
    ('Edge_Forwarder_A', 'Client_H'): 100,
    ('Edge_Forwarder_B', 'Client_I'): 150,
    ('Edge_Forwarder_C', 'Client_J'): 200
}

#          w_u ( 1)
w_u = {
    'A': 1,

```

```

        'B': 1,
        'C': 1,
        'D': 1,
        'E': 1,
        'F': 1,
        'G': 1,
        'H': 1,
        'I': 1,
        'J': 1
    }

    return B_l, Q_l, L_u, C_ij, w_u

class ContinuousCentralizedModel:
    def __init__(self):
        #
        self.B_l, self.Q_l, self.L_u, self.C_ij, self.w_u = _
    ↪ define_problem_data()

        #    U ( 'A', 'B', 'C', ... )
        self.U = list(self.L_u.keys())
        #    L ( '8K', '4K', '2K', '1K' )
        self.L = list(self.B_l.keys())

        #    /    /
        self.S = {'Server'}
        self.F = {'Core_Forwarder_1', 'Core_Forwarder_2',
                  'Edge_Forwarder_A', 'Edge_Forwarder_B', 'Edge_Forwarder_C'}
        #    "Client_X"
        self.U_nodes = {f"Client_{u}" for u in self.U}

        #    V = S ∪ F ∪ U_nodes
        self.V = self.S | self.F | self.U_nodes

        #    E (i→j)
        self.E = list(self.C_ij.keys())

        #
        self.model = None
        self.x = None #  $x[i, l]$  in  $[0, 1]$ 
        self.y = None #  $y[i, j, l]$  in  $[0, 1]$ 
        self.b = None #  $b[i, j] = (l) B_l * y[i, j, l]$ 

        self._create_model()

    def _create_model(self):

```

```

""" Gurobi (continuous in [0,1]) """
self.model = gp.Model("ContinuousCentralizedModel")

#  $x[i,l], y[i,j,l] \in [0,1]$ 
self.x = self.model.addVars(
    self.V, self.L,
    vtype=GRB.CONTINUOUS, lb=0, ub=1, # <--  $[0,1]$ 
    name="x"
)

self.y = self.model.addVars(
    self.E, self.L,
    vtype=GRB.CONTINUOUS, lb=0, ub=1, # <--
    name="y"
)

#  $b[i,j], (i \rightarrow j)$ 
self.b = self.model.addVars(
    self.E,
    vtype=GRB.CONTINUOUS,
    lb=0, #  $\geq 0$ 
    name="b"
)

def set_objective(self):
    """
    : max (u in U) w_u * (l in L) Q_l * x[u,l].
    """
    obj_expr = gp.quicksum(
        self.w_u[u] * self.Q_l[l] * self.x[f"Client_{u}", l]
        for u in self.U
        for l in self.L
    )
    self.model.setObjective(obj_expr, GRB.MAXIMIZE)

def add_constraints(self):
    """ """
    # 1)  $x[s,l] = 1$ 
    for s in self.S:
        for l in self.L:
            self.model.addConstr(
                self.x[s, l] == 1,
                name=f"server_owns_{s}_{l}"
            )

    # 2a)  $l \in L_u[u], x[u,l] = 0$ 
    for u in self.U:

```

```

        allowed_set = set(self.L_u[u]) #
        for l in self.L:
            if l not in allowed_set:
                self.model.addConstr(
                    self.x[f"Client_{u}", l] == 0,
                    name=f"user_cap_{u}_{l}"
                )

# 2b)          1      : (l in L) x[u, l]      1
for u in self.U:
    self.model.addConstr(
        gp.quicksum(self.x[f"Client_{u}", l] for l in self.L) <= 1,
        name=f"user_single_res_{u}"
    )

# 3a)          : y[i, j, l]      x[i, l]
for (i, j) in self.E:
    for l in self.L:
        self.model.addConstr(
            self.y[i, j, l] <= self.x[i, l],
            name=f"trans_logic_1_{i}_{j}_{l}"
        )

# 3b)          l,          : x[i, l]      (p->i) y[p, i, l]
#          F ( )          U_nodes
for node in (self.F | self.U_nodes):
    in_edges = [(p, node) for (p, q) in self.E if q == node]
    for l in self.L:
        self.model.addConstr(
            self.x[node, l] <= gp.quicksum(self.y[p, node, l] for (p,
↪_) in in_edges),
            name=f"trans_logic_2_{node}_{l}"
        )

# 4)          : (l) B_l * y[i, j, l]      C_{i, j}
for (i, j), cap in self.C_ij.items():
    self.model.addConstr(
        gp.quicksum(self.B_l[l] * self.y[i, j, l] for l in self.L) <=
↪cap,
        name=f"capacity_{i}_{j}"
    )

# 5) b[i, j] = (l) B_l * y[i, j, l]
for (i, j) in self.E:
    self.model.addConstr(
        self.b[i, j] == gp.quicksum(self.B_l[l] * self.y[i, j, l] for l
↪in self.L),

```

```

        name=f"bandwidth_usage_{i}_{j}"
    )

def optimize(self):
    """ """
    self.model.optimize()

def print_results(self):
    """ """
    if self.model.status == GRB.OPTIMAL:
        print("\nOptimal solution found!")
        print(f"Optimal Objective Value = {self.model.ObjVal}\n")

        # x[u,l]
        results = []
        for u in self.U:
            for l in self.L:
                val = self.x[f"Client_{u}", l].X
                if val > 1e-6: #
                    results.append({
                        'User': u,
                        'Resolution(l)': l,
                        'x[u,l]': round(val, 4),
                        'Bandwidth(B_l)': self.B_l[l]
                    })
        df_results = pd.DataFrame(results)
        print("User resolution selection (continuous x):")
        print(df_results)

        #
        link_usage = []
        for (i, j) in self.E:
            used = self.b[i, j].X
            cap = self.C_ij[(i, j)]
            link_usage.append({
                'From': i,
                'To': j,
                'Usage(Mbps)': round(used, 2),
                'Capacity(Mbps)': cap,
                'Utilization(%)': round(used / cap * 100, 2) if cap > 1e-9
            })
        else 0
        df_links = pd.DataFrame(link_usage)
        print("\nLink bandwidth usage:")
        print(df_links)

        self.plot_network(df_links)

```

```

else:
    print("No optimal solution found. Status:", self.model.status)

def plot_network(self, df_links):
    fig, ax = plt.subplots(figsize=(15, 10))

    pos = {
        'Server': (0.2, 0.8),
        'Core_Forwarder_1': (0.4, 0.7),
        'Core_Forwarder_2': (0.6, 0.7),
        'Edge_Forwarder_A': (0.3, 0.5),
        'Edge_Forwarder_B': (0.5, 0.5),
        'Edge_Forwarder_C': (0.7, 0.5),
        'Client_A': (0.1, 0.3),
        'Client_B': (0.2, 0.3),
        'Client_C': (0.4, 0.3),
        'Client_D': (0.5, 0.3),
        'Client_E': (0.6, 0.3),
        'Client_F': (0.7, 0.3),
        'Client_G': (0.8, 0.3),
        'Client_H': (0.3, 0.3),
        'Client_I': (0.45, 0.2),
        'Client_J': (0.65, 0.2),
    }

    user_choices = {}
    for u in self.U:
        chosen_info = []
        for l in self.L:
            val = self.x[f"Client_{u}", l].X
            if val > 1e-6:
                chosen_info.append(f"{l}({val:.2f})")
        user_choices[u] = chosen_info

    utilization_dict = {
        (row['From'], row['To']): row['Utilization(%)']
        for _, row in df_links.iterrows()
    }

    def get_color(util_pct):
        return plt.cm.RdYlGn_r(min(util_pct, 100) / 100.0)

    # Draw nodes
    for node, (x_c, y_c) in pos.items():
        if node.startswith("Client_"):
            ax.plot(x_c, y_c, 'gs', markersize=20)
            ax.text(x_c, y_c - 0.02, node, ha='center')

```

```

uid = node.split('_')[1]
allowed = self.L_u[uid]
chosen = user_choices[uid]
chosen_str = ", ".join(chosen) if chosen else "N/A"

ax.text(
    x_c, y_c - 0.05,
    f"Allowed: {'', '.join(allowed)}\nChosen: {chosen_str}",
    ha='center', fontsize=8
)
elif "Forwarder" in node:
    ax.plot(x_c, y_c, 'bo', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')
else:
    ax.plot(x_c, y_c, 'rd', markersize=20)
    ax.text(x_c, y_c - 0.02, node, ha='center')

# Draw links with arrows
for (i, j), util_pct in utilization_dict.items():
    if i not in pos or j not in pos:
        continue
    x1, y1 = pos[i]
    x2, y2 = pos[j]

    # Calculate arrow position (70% along the line)
    arrow_pos = 0.7
    arrow_x = x1 + arrow_pos * (x2 - x1)
    arrow_y = y1 + arrow_pos * (y2 - y1)

    # Draw line
    ax.plot([x1, x2], [y1, y2], '-', color=get_color(util_pct),
    ↪linewidth=2)

    # Add arrow
    dx = x2 - x1
    dy = y2 - y1
    ax.arrow(arrow_x, arrow_y, dx*0.15, dy*0.15,
             head_width=0.02, head_length=0.02,
             fc=get_color(util_pct), ec=get_color(util_pct))

    cap_ij = self.C_ij[(i, j)]
    label = f"{round(util_pct,1)}%\n({cap_ij} Mbps)"

    # Adjust label positions for specific edges to avoid overlap
    if (i, j) == ('Core_Forwarder_1', 'Edge_Forwarder_B'):
        label_offset = -0.03 # Move label down

```

```

        elif (i, j) == ('Core_Forwarder_2', 'Edge_Forwarder_A'):
            label_offset = 0.05    # Move label up
        else:
            label_offset = 0.02    # Default offset

        ax.text((x1 + x2)/2, (y1 + y2)/2 + label_offset, label)

    # Color bar
    norm = plt.Normalize(0, 100)
    sm = plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
    sm.set_array([])
    plt.colorbar(sm, ax=ax, label='Link Utilization(%)')

    # Resolution legend
    ax.text(0.2, 0.85, 'Resolutions:', ha='left')
    for idx, l in enumerate(self.L):
        ax.text(0.2, 0.83 - idx * 0.02, f"{l} -- {self.B_l[l]} Mbps",
        ↪ha='left')

    ax.set_title('Continuous Model')
    ax.axis('off')
    plt.tight_layout()
    plt.show()

def run_optimization():
    """ """
    model = ContinuousCentralizedModel()
    model.set_objective()
    model.add_constraints()
    model.optimize()
    model.print_results()

if __name__ == "__main__":
    run_optimization()

```

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 185 rows, 149 columns and 470 nonzeros

Model fingerprint: 0xd0ca3e06

Coefficient statistics:



Matrix range [1e+00, 2e+02]  
 Objective range [1e+00, 4e+00]  
 Bounds range [1e+00, 1e+00]  
 RHS range [1e+00, 5e+02]  
 Presolve removed 177 rows and 133 columns  
 Presolve time: 0.00s  
 Presolved: 8 rows, 16 columns, 32 nonzeros

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	3.1000000e+01	2.845200e+01	0.000000e+00	0s

Extra simplex iterations after uncrush: 5

9	2.9709677e+01	0.000000e+00	0.000000e+00	0s
---	---------------	--------------	--------------	----

Solved in 9 iterations and 0.01 seconds (0.00 work units)  
 Optimal objective 2.970967742e+01

Optimal solution found!  
 Optimal Objective Value = 29.70967741935484

User resolution selection (continuous x):

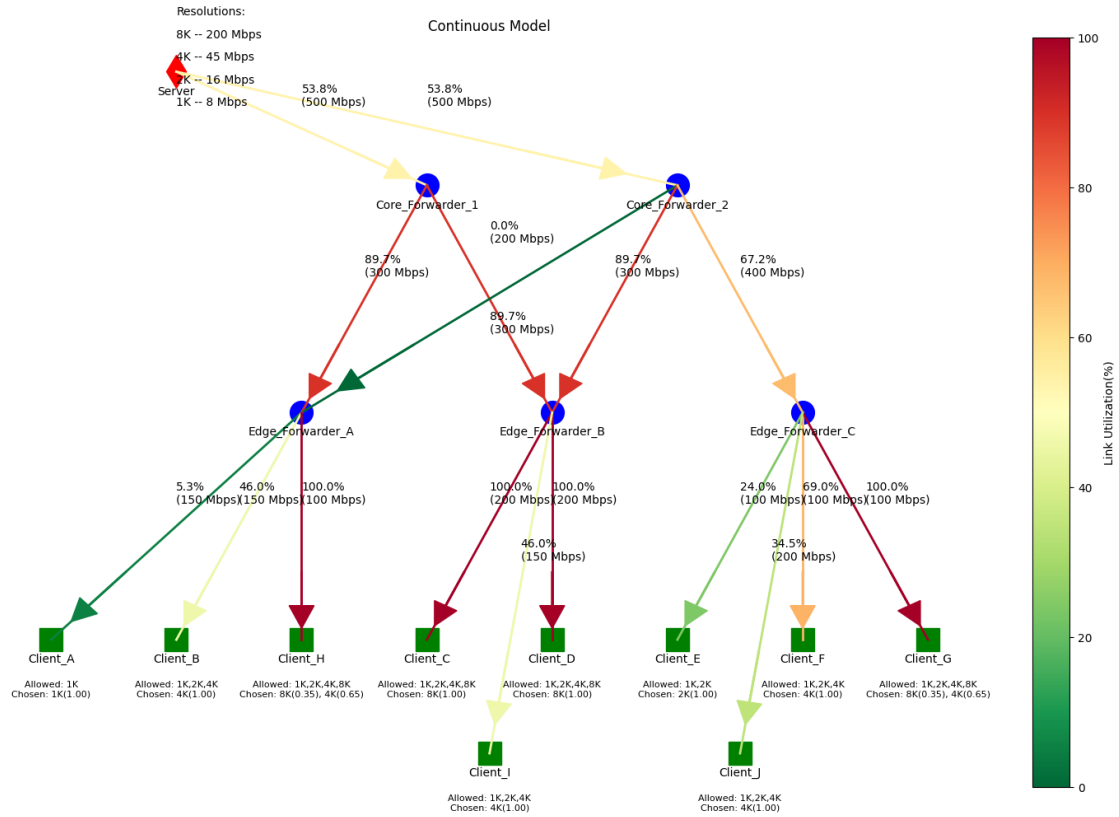
	User Resolution(l)	x[u,l]	Bandwidth(B_l)
0	A	1K 1.0000	8
1	B	4K 1.0000	45
2	C	8K 1.0000	200
3	D	8K 1.0000	200
4	E	2K 1.0000	16
5	F	4K 1.0000	45
6	G	8K 0.3548	200
7	G	4K 0.6452	45
8	H	8K 0.3548	200
9	H	4K 0.6452	45
10	I	4K 1.0000	45
11	J	4K 1.0000	45

Link bandwidth usage:

	From	To	Usage(Mbps)	Capacity(Mbps)	\
0	Server	Core_Forwarder_1	269.0	500	
1	Server	Core_Forwarder_2	269.0	500	
2	Core_Forwarder_1	Edge_Forwarder_A	269.0	300	
3	Core_Forwarder_1	Edge_Forwarder_B	269.0	300	
4	Core_Forwarder_2	Edge_Forwarder_A	0.0	200	
5	Core_Forwarder_2	Edge_Forwarder_B	269.0	300	
6	Core_Forwarder_2	Edge_Forwarder_C	269.0	400	
7	Edge_Forwarder_A	Client_A	8.0	150	
8	Edge_Forwarder_A	Client_B	69.0	150	
9	Edge_Forwarder_B	Client_C	200.0	200	
10	Edge_Forwarder_B	Client_D	200.0	200	

11	Edge_Forwarder_C	Client_E	24.0	100
12	Edge_Forwarder_C	Client_F	69.0	100
13	Edge_Forwarder_C	Client_G	100.0	100
14	Edge_Forwarder_A	Client_H	100.0	100
15	Edge_Forwarder_B	Client_I	69.0	150
16	Edge_Forwarder_C	Client_J	69.0	200

	Utilization(%)
0	53.80
1	53.80
2	89.67
3	89.67
4	0.00
5	89.67
6	67.25
7	5.33
8	46.00
9	100.00
10	100.00
11	24.00
12	69.00
13	100.00
14	100.00
15	46.00
16	34.50



```
[34]: #!/usr/bin/env python3
# -*- coding: utf-8 -*-

import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    """
    // :
    Edge Forwarder 3 :
    ('Edge_Forwarder_A', 'Edge_Forwarder_B'),
    ('Edge_Forwarder_B', 'Edge_Forwarder_C'),
    ('Edge_Forwarder_A', 'Edge_Forwarder_C').
    """
    #
    B_1 = {
        '8K': 200,
        '4K': 45,
        '2K': 16,
```

```

    '1K': 8
}
#
Q_1 = {
    '8K': 4,
    '4K': 3,
    '2K': 2,
    '1K': 1
}
#
L_u = {
    'A': ['1K'],
    'B': ['1K', '2K', '4K'],
    'C': ['1K', '2K', '4K', '8K'],
    'D': ['1K', '2K', '4K', '8K'],
    'E': ['1K', '2K'],
    'F': ['1K', '2K', '4K'],
    'G': ['1K', '2K', '4K', '8K'],
    'H': ['1K', '2K', '4K', '8K'],
    'I': ['1K', '2K', '4K'],
    'J': ['1K', '2K', '4K']
}
# ( Edge<->Edge)
C_ij = {
    ('Server', 'Core_Forwarder_1'): 500,
    ('Server', 'Core_Forwarder_2'): 500,
    ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
    ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
    ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,

    ('Edge_Forwarder_A', 'Client_A'): 150,
    ('Edge_Forwarder_A', 'Client_B'): 150,
    ('Edge_Forwarder_B', 'Client_C'): 200,
    ('Edge_Forwarder_B', 'Client_D'): 200,
    ('Edge_Forwarder_C', 'Client_E'): 100,
    ('Edge_Forwarder_C', 'Client_F'): 100,
    ('Edge_Forwarder_C', 'Client_G'): 100,
    ('Edge_Forwarder_A', 'Client_H'): 100,
    ('Edge_Forwarder_B', 'Client_I'): 150,
    ('Edge_Forwarder_C', 'Client_J'): 200
    # ('Edge_Forwarder_A', 'Edge_Forwarder_B'):100, removed
    # ('Edge_Forwarder_B', 'Edge_Forwarder_C'):150, removed
    # ('Edge_Forwarder_A', 'Edge_Forwarder_C'):120, removed
}
#

```

```

w_u = {
    'A': 1,
    'B': 1,
    'C': 1,
    'D': 1,
    'E': 1,
    'F': 1,
    'G': 1,
    'H': 1,
    'I': 1,
    'J': 1
}
return B_l, Q_l, L_u, C_ij, w_u

# ----- Server subproblem -----
class ServerSubproblem:
    """
     $x[s,l]=1( )$   $y[s,f,l]$  in  $[0,1]$ ,  $=0$ ,  $\leq C_{\{s,f\}}$ 
    """
    def __init__(self, s_name, L, out_edges, B_l, C_ij):
        self.s_name=s_name
        self.L=L
        self.out_edges=out_edges
        self.B_l=B_l
        self.C_ij=C_ij

        self.model=None
        self.yvar={}

    def build_model(self):
        self.model=gp.Model(f"Server_{self.s_name}")
        for (s,f) in self.out_edges:
            for l in self.L:
                self.yvar[(s,f,l)] = self.model.addVar(lb=0,ub=1,vtype=GRB.
↪CONTINUOUS,
                                                                    name=f"y_{s}_{f}_{l}")

        #
        for (s,f) in self.out_edges:
            cap= self.C_ij[(s,f)]
            self.model.addConstr(
                gp.quicksum(self.B_l[l]* self.yvar[(s,f,l)] for l in self.L)
                <= cap
            )
        self.model.setObjective(0.0,GRB.MINIMIZE)
        self.model.update()

    def solve(self):

```

```

        self.model.optimize()

def update_solution(self, global_x, global_y):
    #  $x[s,l]=1$ 
    for l in self.L:
        global_x[(self.s_name,l)] = 1.0
    if self.model.status in (GRB.OPTIMAL,GRB.SUBOPTIMAL,GRB.TIME_LIMIT):
        for (s,f) in self.out_edges:
            for l in self.L:
                global_y[(s,f,l)] = self.yvar[(s,f,l)].X
    else:
        print(f"[WARN] server {self.s_name}, infeasible st={self.model.
↪status}")

# ----- Forwarder subproblem -----
class ForwarderSubproblem:
    """
    :  $x[f,l], y[f,k,l], : y[f,k,l] \leq x[f,l], \sum_l B_l y[f,k,l] \leq C_{\{f,k\}}$ 
    :  $\sum_l \alpha_{\{f,l\}} x[f,l] - \sum_{\{k,l\}} \alpha_{\{k,l\}} y[f,k,l]$ 
    """
    def __init__(self, f_name, L, in_edges, out_edges,
                  B_l,C_ij, alpha_f, alpha_down):
        self.f_name=f_name
        self.L=L
        self.in_edges=in_edges
        self.out_edges=out_edges
        self.B_l=B_l
        self.C_ij=C_ij

        self.alpha_f=alpha_f #  $\alpha_{\{f,l\}}$ 
        self.alpha_down= alpha_down #  $\alpha_{\{k,l\}}$  for each out edge k

        self.model=None
        self.xvar={}
        self.yvar={}

    def build_model(self):
        self.model= gp.Model(f"Fwd_{self.f_name}")

        for l in self.L:
            self.xvar[l] = self.model.addVar(lb=0,ub=1,vtype=GRB.CONTINUOUS,
                                              name=f"x_{self.f_name}_{l}")

        for (f,k) in self.out_edges:
            for l in self.L:
                self.yvar[(f,k,l)] = self.model.addVar(lb=0,ub=1,vtype=GRB.
↪CONTINUOUS,

```

```

name=f"y_{f}_{k}_{l}")

#
for (f,k) in self.out_edges:
    cap= self.C_ij.get((f,k),0)
    self.model.addConstr(
        gp.quicksum(self.B_l[l]* self.yvar[(f,k,l)] for l in self.L)
        <= cap
    )

#  $y[f,k,l] \leq x[f,l]$ 
for (f,k) in self.out_edges:
    for l in self.L:
        self.model.addConstr(
            self.yvar[(f,k,l)] <= self.xvar[l]
        )

#  $= \sum_l \alpha_f[l] * x[f,l] - \sum_{\{(f \rightarrow k), l\}} \alpha_{k,l} y[f,k,l]$ 
obj= gp.quicksum(self.alpha_f[l]* self.xvar[l] for l in self.L)
for (f2,k) in self.out_edges:
    for l in self.L:
        if (k,l) in self.alpha_down:
            obj-= self.alpha_down[(k,l)]* self.yvar[(f2,k,l)]
self.model.setObjective(obj, GRB.MINIMIZE)
self.model.update()

def solve(self):
    self.model.optimize()

def update_solution(self, global_x, global_y):
    st= self.model.status
    if st not in (GRB.OPTIMAL,GRB.SUBOPTIMAL,GRB.TIME_LIMIT):
        print(f"[WARN] forwarder {self.f_name}, st={st}")
        return
    # x
    for l in self.L:
        global_x[(self.f_name,l)] = self.xvar[l].X
    # y
    for (f,k) in self.out_edges:
        for l in self.L:
            val= self.yvar[(f,k,l)].X
            global_y[(f,k,l)] = val

# ----- User subproblem -----
class UserSubproblem:
    """

```

```

:  $x[u,l]$  in  $[0,1]$ ,  $\sum_l x[u,l] \leq 1$ ,  $x[u,l]=0$  if  $l$  not in  $L_u[u]$ .
:  $-w_u \sum_l Q_l x[u,l] + \sum_l \alpha_{\{u,l\}} x[u,l]$ 
"""
def __init__(self, u_name, L, in_edges,
              w_u, Q_l, L_u_allows,
              alpha_u):
    self.u_name=u_name
    self.L=L
    self.in_edges=in_edges
    self.w_u=w_u
    self.Q_l=Q_l
    self.L_u_allows= L_u_allows
    self.alpha_u= alpha_u

    self.model=None
    self.xvar={}

def build_model(self):
    self.model= gp.Model(f"User_{self.u_name}")
    for l in self.L:
        self.xvar[l] = self.model.addVar(lb=0,ub=1,vtype=GRB.CONTINUOUS,
                                          name=f"x_{self.u_name}_{l}")

    #  $x[u,l]=0$  if not allowed
    for l in self.L:
        if l not in self.L_u_allows:
            self.model.addConstr(self.xvar[l]==0)
    #  $\sum_l x[u,l] \leq 1$ 
    self.model.addConstr(
        gp.quicksum(self.xvar[l] for l in self.L) <= 1
    )

    #  $= -w_u \sum_l Q_l x[u,l] + \sum_l \alpha_{\{u,l\}} x[u,l]$ 
    obj= gp.quicksum(
        (-self.w_u*self.Q_l[l] + self.alpha_u[l]) * self.xvar[l]
        for l in self.L
    )
    self.model.setObjective(obj, GRB.MINIMIZE)
    self.model.update()

def solve(self):
    self.model.optimize()

def update_solution(self, global_x):
    st= self.model.status
    if st not in (GRB.OPTIMAL,GRB.SUBOPTIMAL,GRB.TIME_LIMIT):
        print(f"[WARN] user {self.u_name}, st={st}")
    return

```



```

        for l in self.L:
            global_x[(self.u_name,l)] = self.xvar[l].X

# -----
class DistributedLagrangianSolver:
    def __init__(self,B_l,Q_l,L_u,C_ij,w_u,
                  max_iter=6, step_size=0.5):
        self.B_l=B_l
        self.Q_l=Q_l
        self.L_u=L_u
        self.C_ij=C_ij
        self.w_u=w_u

        #
        self.S= {"Server"}
        #
        self.F= {"Core_Forwarder_1","Core_Forwarder_2",
                 "Edge_Forwarder_A","Edge_Forwarder_B","Edge_Forwarder_C"}
        self.U= list(L_u.keys()) # e.g. A,B,C,D,E,F,G,H,I,J

        # self.U_nodes      "Client_A",      "A"
        #
        #      "Client_X" ,      :
        #      map: nodeName= f"Client_{u}"

        self.V= self.S| self.F # + user? We'll treat user as just string id
        self.E= list(C_ij.keys())
        self.L= list(B_l.keys())

        # alpha_{i,l}: i in (F or user)
        self.alpha={}
        for f in self.F:
            for l in self.L:
                self.alpha[(f,l)] = 0.0
        for u in self.U:
            for l in self.L:
                self.alpha[(u,l)] = 0.0

        self.max_iter=max_iter
        self.step_size=step_size

        self.x_sol={} # (node,l)-> fraction
        self.y_sol={} # ((i,j),l)-> fraction
        self.obj_history=[]

    def run(self):

```

```

for it in range(self.max_iter):
    print(f"\n=== Iteration {it} ===")
    # 1) server
    for s in self.S:
        out_e = [(i,j) for (i,j) in self.E if i==s]
        sp= ServerSubproblem(s,self.L,out_e,self.B_l,self.C_ij)
        sp.build_model()
        sp.solve()
        sp.update_solution(self.x_sol,self.y_sol)

    # 2) forwarder
    for f in self.F:
        in_e = [(p,f2) for (p,f2) in self.E if f2==f]
        out_e=[(f2,k) for (f2,k) in self.E if f2==f]
        alpha_f= {}
        alpha_down={}
        for l in self.L:
            alpha_f[l]= self.alpha[(f,l)]
        for (f2,k) in out_e:
            for l in self.L:
                alpha_down[(k,l)] = self.alpha.get((k,l),0.0)
        fprob= ForwarderSubproblem(f,self.L,in_e,out_e,
                                   self.B_l,self.C_ij,
                                   alpha_f, alpha_down)

        fprob.build_model()
        fprob.solve()
        fprob.update_solution(self.x_sol,self.y_sol)

    # 3) user
    for u in self.U:
        in_e= [(p,u2) for (p,u2) in self.E if u2==f"Client_{u}"]
        # E "Client_U"? Let's unify approach
        # Actually in your data "Client_{A}" is not in keys, you used
        ↪ (Edge_Forwarder_A, 'Client_A')?
        # So we do:
        node_u= f"Client_{u}"
        # but we see your E uses "Client_A" or "Client_B"? yes
        # so let's rename input data to (Edge_Forwarder_A, 'Client_A')
        ↪ => that is (i-> j)
        # so in_e= ???

        # For simplicity, let's just do "u" as is. Because in data
        ↪ 'Client_A' is j. We'll do:
        # in_e= (p,f"Client_{u}") if that is how we spelled j. We'll
        ↪ check carefully
        # we'll do:

```

```

# Actually let's match your data: they are "Client_A" not "A"
only

# So let's do: in_e= list of edges if j== f"Client_{u}"
node_u= f"Client_{u}"
in_e= [(p,j) for (p,j) in self.E if j== node_u]

alpha_u={}
for l in self.L:
    alpha_u[l]= self.alpha.get((u,l),0.0)
up= UserSubproblem(u,self.L,in_e,self.w_u[u],self.Q_l,
                    self.L_u[u], alpha_u)
up.build_model()
up.solve()
up.update_solution(self.x_sol)

# 4) subgradient
# alpha_{i,l} = max(0, alpha_{i,l}+ step( x[i,l] - sum_{(p->i)}
y[p,i,l] ) )
# i can be f or user
# for user i, we do i= u. For forwarder i= f
for f in self.F:
    in_e=[(p,f2) for (p,f2) in self.E if f2==f]
    for l in self.L:
        lhs= self.x_sol.get((f,l),0.0)
        rhs=0.0
        for (p,f2) in in_e:
            rhs+= self.y_sol.get((p,f,l),0.0)
        old= self.alpha[(f,l)]
        new= old+ self.step_size*(lhs-rhs)
        self.alpha[(f,l)] = max(0.0,new)

for u in self.U:
    node_u= f"Client_{u}"
    in_e=[(p,j) for (p,j) in self.E if j==node_u]
    for l in self.L:
        lhs= self.x_sol.get((u,l),0.0) # store user as key=(u,l)?
or (Client_u,l)?
        # depends on how user subproblem updated x
        # we said update_solution used self.xvar[l],
global_x[(u,l)] = ...
        # so let's do that
        lhs= self.x_sol.get((u,l),0.0)

        rhs=0.0
        for (p,j) in in_e:
            rhs+= self.y_sol.get((p,node_u,l),0.0)
        old= self.alpha[(u,l)]

```

```

        new= old+ self.step_size*(lhs-rhs)
        self.alpha[(u,l)] = max(0.0,new)

    # 5) estimate obj
    val= self.estimate_obj()
    self.obj_history.append(val)
    print(f"approx Lagrange obj= {val:.3f}")

def estimate_obj(self):
    #  $L = -\sum_{\{u,l\}} w_u Q_l x[u,l] + \sum_{\{i,l\}} \alpha_{\{i,l\}} (x[i,l] -$ 
    ↪  $\sum_{\{p \rightarrow i\}} y_{\{p,i,l\}})$ 
    sum_user=0
    for u in self.U:
        for l in self.L:
            sum_user+= self.w_u[u]* self.Q_l[l]* self.x_sol.get((u,l),0.0)
    sum_alpha=0
    for f in self.F:
        in_e= [(p,f2) for (p,f2) in self.E if f2==f]
        for l in self.L:
            xv= self.x_sol.get((f,l),0.0)
            sy= sum(self.y_sol.get((p,f,l),0.0) for (p,f2) in in_e)
            sum_alpha+= self.alpha[(f,l)]*(xv- sy)
    for u in self.U:
        node_u= f"Client_{u}"
        in_e= [(p,j) for (p,j) in self.E if j==node_u]
        for l in self.L:
            xv= self.x_sol.get((u,l),0.0)
            sy=0
            for (p,j) in in_e:
                sy+= self.y_sol.get((p,node_u,l),0.0)
            sum_alpha+= self.alpha[(u,l)]*(xv- sy)
    return - sum_user + sum_alpha

def get_final_solution(self):
    return self.x_sol,self.y_sol,self.alpha

def plot_network_and_solution(B_l, C_ij, L_u, x_sol, y_sol):
    """
    ,      fwd-fwd link.
    final usage & user choice
    """
    E=list(C_ij.keys())
    usage=[]
    for (i,j),cap in C_ij.items():
        used=0
        for l in B_l:
            frac= y_sol.get((i,j,l),0.0)

```

```

        used+= B_l[l]* frac
    usage.append({
        'From':i, 'To':j,
        'Used': round(used,2),
        'Cap': cap,
        'Util(%)': round(used/cap*100,2) if cap>1e-9 else 0
    })
df_links= pd.DataFrame(usage)

# user
user_choices={}
for u, allows in L_u.items():
    chosen=[]
    for l in B_l:
        val= x_sol.get((u,l),0.0)
        if val>1e-5:
            chosen.append(f"{l}({val:.2f})")
    user_choices[u]= chosen

# pos
pos={
    'Server':(0.2,0.8),
    'Core_Forwarder_1':(0.4,0.7),
    'Core_Forwarder_2':(0.6,0.7),
    'Edge_Forwarder_A':(0.3,0.5),
    'Edge_Forwarder_B':(0.5,0.5),
    'Edge_Forwarder_C':(0.7,0.5),
    'Client_A':(0.1,0.3),
    'Client_B':(0.2,0.3),
    'Client_C':(0.4,0.3),
    'Client_D':(0.5,0.3),
    'Client_E':(0.6,0.3),
    'Client_F':(0.7,0.3),
    'Client_G':(0.8,0.3),
    'Client_H':(0.3,0.3),
    'Client_I':(0.45,0.2),
    'Client_J':(0.65,0.2),
}

import matplotlib
fig,ax= plt.subplots(figsize=(14,8))
def get_color(u):
    return plt.cm.RdYlGn_r(min(u,100)/100.0)

for node,(xx,yy) in pos.items():
    if node.startswith("Client_"):
        ax.plot(xx,yy, 'bs', markersize=14)

```

```

        ax.text(xx,yy-0.02,node,ha='center',fontsize=9)
        # parse user ID
        uid=node.split('_')[1]
        allows= L_u.get(uid,[])
        chosen_str=', '.join(user_choices[uid]) if uid in user_choices else_
↪ "N/A"

        ax.text(xx,yy-0.06,f"Allowed:{', '.join(allows)}\nChosen:
↪ {chosen_str}",
                ha='center',fontsize=8)
    elif "Forwarder" in node:
        ax.plot(xx,yy,'go',markersize=14)
        ax.text(xx,yy-0.02,node,ha='center',fontsize=9)
    else:
        ax.plot(xx,yy,'r^',markersize=14)
        ax.text(xx,yy-0.02,node,ha='center',fontsize=9)

for idx,row in df_links.iterrows():
    i,j= row['From'], row['To']
    used, cap, util= row['Used'], row['Cap'], row['Util(%)']
    if i in pos and j in pos:
        (x1,y1)= pos[i]
        (x2,y2)= pos[j]
        ax.plot([x1,x2],[y1,y2], '--', color=get_color(util), linewidth=2)
        ax.text((x1+x2)/2, (y1+y2)/2+0.02, f"{util:.1f}%\n({cap}Mbps)",
                ha='center', fontsize=8)

norm= matplotlib.colors.Normalize(0,100)
sm= plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
sm.set_array([])
plt.colorbar(sm, ax=ax, fraction=0.03).set_label("Link Util(%)")
ax.set_title("Distributed Lagrangian (No Edge-Forwarder <-> Edge-Forwarder_
↪ link)")
ax.axis('off')
plt.tight_layout()
plt.show()

print("\nLink usage summary:\n", df_links)

def run_distributed_model():
    B_l, Q_l, L_u, C_ij, w_u= define_problem_data()
    solver= DistributedLagrangianSolver(B_l, Q_l, L_u, C_ij, w_u,
                                       max_iter=8, step_size=0.3)

    solver.run()

    x_sol, y_sol, alpha= solver.get_final_solution()

```

```

#
plot_network_and_solution(B_l,C_ij,L_u,x_sol,y_sol)

print("\nAlpha:\n", alpha)
print("\nObj history:\n", solver.obj_history)

if __name__=="__main__":
    run_distributed_model()

```

=== Iteration 0 ===

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range	[8e+00, 2e+02]
Objective range	[0e+00, 0e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0xc793cab2

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[0e+00, 0e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x1ce972db

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x3c52091a

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
-----------	-----------	-------------	-----------	------



0 0.0000000e+00 0.000000e+00 0.000000e+00 0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xfde94416

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xe4e69cae

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros

Model fingerprint: 0x18a411d8

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 4 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xcf2d623e

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-3.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x9f1e2cf6

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-4.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -4.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x9f1e2cf6

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-4.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -4.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0x2877f4a3

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.0000000e+00	0.000000e+00	1.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.0000000e+00	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xcf2d623e

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-3.0000000e+00	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x9f1e2cf6

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.01s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-4.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -4.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x9f1e2cf6

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-4.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -4.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xcfd2d623e

Coefficient statistics:

```

Matrix range      [1e+00, 1e+00]
Objective range   [1e+00, 4e+00]
Bounds range      [1e+00, 1e+00]
RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0   -3.0000000e+00   0.000000e+00   3.000000e+00    0s
Extra simplex iterations after uncrush: 1
           1   -3.0000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -3.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0xcf2d623e
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0   -3.0000000e+00   0.000000e+00   3.000000e+00    0s
Extra simplex iterations after uncrush: 1
           1   -3.0000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -3.000000000e+00
approx Lagrange obj= -22.000

=== Iteration 1 ===
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

```

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range [8e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0x3a4ac5ec

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.2000000e+00	0.000000e+00	1.200000e+00	0s

Extra simplex iterations after uncrush: 4

4	-1.2000000e+00	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 4 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.200000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x3dd5b98d

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x66658194

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xd7c9503e

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]



```

    RHS range      [2e+02, 4e+02]
Presolve removed 15 rows and 16 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0    -2.2965000e+00    0.000000e+00    1.200000e+00      0s
Extra simplex iterations after uncrush: 7
           7    -2.2965000e+00    0.000000e+00    0.000000e+00      0s

Solved in 7 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.296500000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros
Model fingerprint: 0x8349a3cf
Coefficient statistics:
  Matrix range      [1e+00, 2e+02]
  Objective range   [3e-01, 3e-01]
  Bounds range      [1e+00, 1e+00]
  RHS range         [2e+02, 2e+02]
Presolve removed 15 rows and 16 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0    0.0000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros
Model fingerprint: 0x764ef582
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [7e-01, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 4 rows and 4 columns

```

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-7.0000000e-01	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -7.000000000e-01

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xa3a0eaaf

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.7000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x79784bad

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -3.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x79784bad

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -3.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0x9761b8f5

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.7000000e+00	0.000000e+00	1.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.7000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)  
 Optimal objective -1.700000000e+00  
 Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
 Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros  
 Model fingerprint: 0xa3a0eaaf

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
 Objective range [1e+00, 4e+00]  
 Bounds range [1e+00, 1e+00]  
 RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.7000000e+00	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)  
 Optimal objective -2.700000000e+00  
 Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
 Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros  
 Model fingerprint: 0x79784bad

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
 Objective range [1e+00, 4e+00]  
 Bounds range [1e+00, 1e+00]  
 RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.700000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros  
Model fingerprint: 0x79784bad  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 1 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -3.700000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros  
Model fingerprint: 0xa3a0eaaf  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 2 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.7000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -2.700000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5

LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xa3a0eaaf

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	3.000000e+00	0s
Extra simplex iterations after uncrush: 1				
1	-2.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.700000000e+00

approx Lagrange obj= -20.200

=== Iteration 2 ===

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5  
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range [8e+00, 2e+02]  
Objective range [0e+00, 0e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0x350fd1c3

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[6e-01, 6e-01]
Bounds range	[1e+00, 1e+00]
RHS range	[3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x1ce972db

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[0e+00, 0e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x3c52091a

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x1e36fe5b

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [6e-01, 6e-01]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros



Model fingerprint: 0xe4e69cae

Coefficient statistics:

Matrix range [1e+00, 2e+02]  
Objective range [0e+00, 0e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros

Model fingerprint: 0x60bdec89

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [4e-01, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 4 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-4.0000000e-01	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -4.000000000e-01

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xbdaf8787

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]

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    Bounds range      [1e+00, 1e+00]
    RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0   -2.4000000e+00   0.000000e+00   3.000000e+00    0s
Extra simplex iterations after uncrush: 1
          1   -2.4000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.400000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x118cb53b
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 3e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0   -3.4000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)
Optimal objective -3.400000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x118cb53b
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 3e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]

```

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.4000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)

Optimal objective -3.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0x0664bb33

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.4000000e+00	0.000000e+00	1.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.4000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xbdaf8787

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.4000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.4000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x118cb53b

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.4000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x118cb53b

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
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0 -3.4000000e+00 0.000000e+00 0.000000e+00 0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xbdaf8787

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.4000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.4000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.400000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0xbdaf8787

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.4000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

```

1    -2.4000000e+00    0.000000e+00    0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.400000000e+00
approx Lagrange obj= -18.400

=== Iteration 3 ===
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros
Model fingerprint: 0xf98ed5cb
Coefficient statistics:
  Matrix range      [8e+00, 2e+02]
  Objective range   [0e+00, 0e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [5e+02, 5e+02]
Presolve removed 2 rows and 8 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0    0.0000000e+00    0.000000e+00    0.000000e+00    0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective 0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros
Model fingerprint: 0x6eb95e38
Coefficient statistics:
  Matrix range      [1e+00, 2e+02]
  Objective range   [3e-01, 6e-01]
  Bounds range      [1e+00, 1e+00]
  RHS range         [3e+02, 3e+02]
Presolve removed 10 rows and 12 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0    0.0000000e+00    0.000000e+00    0.000000e+00    0s

```

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros  
Model fingerprint: 0x3dd5b98d  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [3e-01, 3e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+02, 2e+02]  
Presolve removed 20 rows and 20 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros  
Model fingerprint: 0x66658194  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [3e-01, 3e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+02, 2e+02]  
Presolve removed 15 rows and 16 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5

LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x5e488ecc

Coefficient statistics:

Matrix range [1e+00, 2e+02]  
Objective range [3e-01, 6e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.0965000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.096500000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5  
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x8349a3cf

Coefficient statistics:

Matrix range [1e+00, 2e+02]  
Objective range [3e-01, 3e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5  
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]



Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros

Model fingerprint: 0x8bfbc65

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e-01, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 4 rows and 4 columns

Presolve time: 0.01s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.0000000e-01	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.000000000e-01

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x22c20213

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.1000000e+00	0.000000e+00	3.000000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.1000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.100000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x2b130a9d

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.1000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.100000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x2b130a9d

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.1000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.100000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0xaff795bf

Coefficient statistics:

Matrix range [1e+00, 1e+00]

```

Objective range [1e+00, 4e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Presolve removed 3 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -1.1000000e+00    0.000000e+00    1.000000e+00      0s
Extra simplex iterations after uncrush: 1
      1      -1.1000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -1.100000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x22c20213
Coefficient statistics:
  Matrix range [1e+00, 1e+00]
  Objective range [1e+00, 4e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -2.1000000e+00    0.000000e+00    3.000000e+00      0s
Extra simplex iterations after uncrush: 1
      1      -2.1000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.100000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x2b130a9d
Coefficient statistics:
  Matrix range [1e+00, 1e+00]

```

```

Objective range [1e+00, 3e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.01s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.      Dual Inf.      Time
          0      -3.1000000e+00      0.000000e+00      0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective -3.100000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x2b130a9d
Coefficient statistics:
  Matrix range [1e+00, 1e+00]
  Objective range [1e+00, 3e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.      Dual Inf.      Time
          0      -3.1000000e+00      0.000000e+00      0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective -3.100000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x22c20213
Coefficient statistics:
  Matrix range [1e+00, 1e+00]
  Objective range [1e+00, 4e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns

```

```

Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -2.1000000e+00    0.000000e+00    3.000000e+00      0s
Extra simplex iterations after uncrush: 1
      1      -2.1000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.100000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x22c20213
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -2.1000000e+00    0.000000e+00    3.000000e+00      0s
Extra simplex iterations after uncrush: 1
      1      -2.1000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.100000000e+00
approx Lagrange obj= -15.468

=== Iteration 4 ===
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros
Model fingerprint: 0xf98ed5cb
Coefficient statistics:
  Matrix range      [8e+00, 2e+02]
  Objective range   [0e+00, 0e+00]

```

```

    Bounds range      [1e+00, 1e+00]
    RHS range         [5e+02, 5e+02]
Presolve removed 2 rows and 8 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0    0.0000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros
Model fingerprint: 0x555e43a4
Coefficient statistics:
  Matrix range      [1e+00, 2e+02]
  Objective range   [1e-01, 6e-01]
  Bounds range      [1e+00, 1e+00]
  RHS range         [3e+02, 3e+02]
Presolve removed 10 rows and 12 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
          0    0.0000000e+00   0.000000e+00   6.468750e-03    0s
Extra simplex iterations after uncrush: 1
          1    0.0000000e+00   0.000000e+00   0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros
Model fingerprint: 0x1ce972db
Coefficient statistics:
  Matrix range      [1e+00, 2e+02]
  Objective range   [0e+00, 0e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+02, 2e+02]

```

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xe21cbc05

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[1e-01, 1e-01]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xc090aaa0

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[1e-01, 9e-01]
Bounds range	[1e+00, 1e+00]
RHS range	[2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
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0	0.0000000e+00	0.0000000e+00	6.468750e-03	0s
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Extra simplex iterations after uncrush: 1

1	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s
---	---------------	---------------	---------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros  
Model fingerprint: 0xe4e69cae  
Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[0e+00, 0e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros  
Model fingerprint: 0x3391296a  
Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[2e-01, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 4 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s



Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros  
Model fingerprint: 0x02f74169  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 2 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.0000000e+00	0.000000e+00	2.800000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -2.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros  
Model fingerprint: 0x741f3ebf  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 3e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 1 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x741f3ebf

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 3e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0xdd6b92a1

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[8e-01, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.0000000e+00	0.000000e+00	8.000000e-01	0s

Extra simplex iterations after uncrush: 1

1	-1.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x02f74169

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.0000000e+00	0.000000e+00	2.800000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x741f3ebf

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x741f3ebf

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 3e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.01s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-3.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -3.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x02f74169

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.0000000e+00	0.000000e+00	2.800000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set

[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x02f74169

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.0000000e+00	0.000000e+00	2.800000e+00	0s

Extra simplex iterations after uncrush: 1

1	-2.0000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.000000000e+00

approx Lagrange obj= -15.000

=== Iteration 5 ===

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range [8e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0xbf95be45

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[1e-01, 6e-01]
Bounds range	[1e+00, 1e+00]
RHS range	[3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x3dd5b98d

Coefficient statistics:

Matrix range	[1e+00, 2e+02]
Objective range	[3e-01, 3e-01]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xbeda2654

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x21fe9abf

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 9e-01]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x8349a3cf

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

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    RHS range      [2e+02, 2e+02]
Presolve removed 15 rows and 16 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0      0.0000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros
Model fingerprint: 0x3391296a
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [2e-01, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 4 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0      0.0000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x70dd165b
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed

```



Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.8000000e+00	0.000000e+00	2.700000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.8000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.800000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x81c8f31d

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 3e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.8000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.800000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x81c8f31d

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 3e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.8000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
 Optimal objective -2.800000000e+00  
 Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
 Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros  
 Model fingerprint: 0x4d4a6d95

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
 Objective range [7e-01, 4e+00]  
 Bounds range [1e+00, 1e+00]  
 RHS range [1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-8.0000000e-01	0.000000e+00	7.000000e-01	0s

Extra simplex iterations after uncrush: 1

1	-8.0000000e-01	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)  
 Optimal objective -8.000000000e-01  
 Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
 Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros  
 Model fingerprint: 0x70dd165b

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
 Objective range [1e+00, 4e+00]  
 Bounds range [1e+00, 1e+00]  
 RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.8000000e+00	0.000000e+00	2.700000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.8000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -1.800000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros  
Model fingerprint: 0x81c8f31d  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 3e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 1 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.8000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -2.800000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros  
Model fingerprint: 0x81c8f31d  
Coefficient statistics:  
Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 3e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]  
Presolve removed 1 rows and 4 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.8000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective -2.800000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5

LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x70dd165b

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.8000000e+00	0.000000e+00	2.700000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.8000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.800000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5  
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x70dd165b

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.8000000e+00	0.000000e+00	2.700000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.8000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.800000000e+00

approx Lagrange obj= -16.500

=== Iteration 6 ===

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range [8e+00, 2e+02]

Objective range [0e+00, 0e+00]

Bounds range [1e+00, 1e+00]

RHS range [5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0xbf95be45

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 6e-01]

Bounds range [1e+00, 1e+00]

RHS range [3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x3dd5b98d

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0xbeda2654

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x21fe9abf

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 9e-01]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 4e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x8349a3cf

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros

Model fingerprint: 0x3391296a

Coefficient statistics:

```

Matrix range      [1e+00, 1e+00]
Objective range   [2e-01, 4e+00]
Bounds range      [1e+00, 1e+00]
RHS range         [1e+00, 1e+00]
Presolve removed 4 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0    0.0000000e+00  0.0000000e+00  0.0000000e+00    0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective  0.000000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x158a475a
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration    Objective      Primal Inf.    Dual Inf.      Time
           0   -1.7000000e+00  0.0000000e+00  2.500000e+00    0s
Extra simplex iterations after uncrush: 1
           1   -1.7000000e+00  0.0000000e+00  0.0000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -1.700000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x426f46a1
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 3e+00]

```



```

    Bounds range      [1e+00, 1e+00]
    RHS range         [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
           0    -2.7000000e+00  0.000000e+00  0.000000e+00    0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.700000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0x426f46a1
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 3e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration   Objective      Primal Inf.    Dual Inf.      Time
           0    -2.7000000e+00  0.000000e+00  0.000000e+00    0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.700000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros
Model fingerprint: 0x8b300611
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [5e-01, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 3 rows and 4 columns
Presolve time: 0.00s

```

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-7.0000000e-01	0.000000e+00	5.000000e-01	0s

Extra simplex iterations after uncrush: 1

1	-7.0000000e-01	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -7.000000000e-01

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x158a475a

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.7000000e+00	0.000000e+00	2.500000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.7000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x426f46a1

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 3e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0x426f46a1

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 3e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.7000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x158a475a

Coefficient statistics:

Matrix range	[1e+00, 1e+00]
Objective range	[1e+00, 4e+00]
Bounds range	[1e+00, 1e+00]
RHS range	[1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.7000000e+00	0.000000e+00	2.500000e+00	0s

Extra simplex iterations after uncrush: 1

1   -1.7000000e+00   0.000000e+00   0.000000e+00   0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.700000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x158a475a

Coefficient statistics:

Matrix range        [1e+00, 1e+00]

Objective range     [1e+00, 4e+00]

Bounds range        [1e+00, 1e+00]

RHS range           [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.7000000e+00	0.000000e+00	2.500000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.7000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.700000000e+00

approx Lagrange obj= -15.600

=== Iteration 7 ===

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 8 columns and 8 nonzeros

Model fingerprint: 0xf98ed5cb

Coefficient statistics:

Matrix range        [8e+00, 2e+02]

Objective range     [0e+00, 0e+00]

Bounds range        [1e+00, 1e+00]

RHS range           [5e+02, 5e+02]

Presolve removed 2 rows and 8 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 10 rows, 12 columns and 24 nonzeros

Model fingerprint: 0xbf95be45

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e-01, 6e-01]

Bounds range [1e+00, 1e+00]

RHS range [3e+02, 3e+02]

Presolve removed 10 rows and 12 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 20 columns and 48 nonzeros

Model fingerprint: 0x3dd5b98d

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [1e+02, 2e+02]

Presolve removed 20 rows and 20 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.0000000e+00	0.0000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros  
Model fingerprint: 0xbeda2654  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [1e-01, 3e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+02, 2e+02]  
Presolve removed 15 rows and 16 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros  
Model fingerprint: 0x21fe9abf  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [1e-01, 9e-01]  
Bounds range [1e+00, 1e+00]  
RHS range [2e+02, 4e+02]  
Presolve removed 15 rows and 16 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)  
Optimal objective 0.000000000e+00  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 15 rows, 16 columns and 36 nonzeros

Model fingerprint: 0x8349a3cf

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [3e-01, 3e-01]

Bounds range [1e+00, 1e+00]

RHS range [2e+02, 2e+02]

Presolve removed 15 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 4 rows, 4 columns and 7 nonzeros

Model fingerprint: 0x3391296a

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [2e-01, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 4 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	0.0000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective 0.000000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set  
[SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x1de24bc9

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 2 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.5000000e+00	0.000000e+00	2.400000e+00	0s

Extra simplex iterations after uncrush: 1

1	-1.5000000e+00	0.000000e+00	0.000000e+00	0s
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Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.500000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0xff00d541

Coefficient statistics:

Matrix range [1e+00, 1e+00]

Objective range [1e+00, 2e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.5000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.500000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros

Model fingerprint: 0xff00d541



Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [1e+00, 2e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 1 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-2.5000000e+00	0.000000e+00	0.000000e+00	0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)

Optimal objective -2.500000000e+00

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 3 rows, 4 columns and 6 nonzeros

Model fingerprint: 0x8083da61

Coefficient statistics:

Matrix range [1e+00, 1e+00]  
Objective range [4e-01, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 1e+00]

Presolve removed 3 rows and 4 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-5.0000000e-01	0.000000e+00	4.000000e-01	0s

Extra simplex iterations after uncrush: 1

1	-5.0000000e-01	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -5.000000000e-01

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros

Model fingerprint: 0x1de24bc9

Coefficient statistics:

Matrix range [1e+00, 1e+00]

```

Objective range [1e+00, 4e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -1.5000000e+00    0.000000e+00    2.400000e+00      0s
Extra simplex iterations after uncrush: 1
      1      -1.5000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -1.500000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0xff00d541
Coefficient statistics:
  Matrix range [1e+00, 1e+00]
  Objective range [1e+00, 2e+00]
  Bounds range [1e+00, 1e+00]
  RHS range [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -2.5000000e+00    0.000000e+00    0.000000e+00      0s

Solved in 0 iterations and 0.01 seconds (0.00 work units)
Optimal objective -2.500000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 1 rows, 4 columns and 4 nonzeros
Model fingerprint: 0xff00d541
Coefficient statistics:
  Matrix range [1e+00, 1e+00]
  Objective range [1e+00, 2e+00]
  Bounds range [1e+00, 1e+00]

```

```

RHS range      [1e+00, 1e+00]
Presolve removed 1 rows and 4 columns
Presolve time: 0.00s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -2.5000000e+00    0.000000e+00   0.000000e+00    0s

Solved in 0 iterations and 0.00 seconds (0.00 work units)
Optimal objective -2.500000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x1de24bc9
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns
Presolve time: 0.01s
Presolve: All rows and columns removed
Iteration      Objective      Primal Inf.    Dual Inf.      Time
      0      -1.5000000e+00    0.000000e+00   2.400000e+00    0s
Extra simplex iterations after uncrush: 1
      1      -1.5000000e+00    0.000000e+00   0.000000e+00    0s

Solved in 1 iterations and 0.01 seconds (0.00 work units)
Optimal objective -1.500000000e+00
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 2 rows, 4 columns and 5 nonzeros
Model fingerprint: 0x1de24bc9
Coefficient statistics:
  Matrix range      [1e+00, 1e+00]
  Objective range   [1e+00, 4e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Presolve removed 2 rows and 4 columns

```

Presolve time: 0.00s

Presolve: All rows and columns removed

Iteration	Objective	Primal Inf.	Dual Inf.	Time
0	-1.5000000e+00	0.000000e+00	2.400000e+00	0s

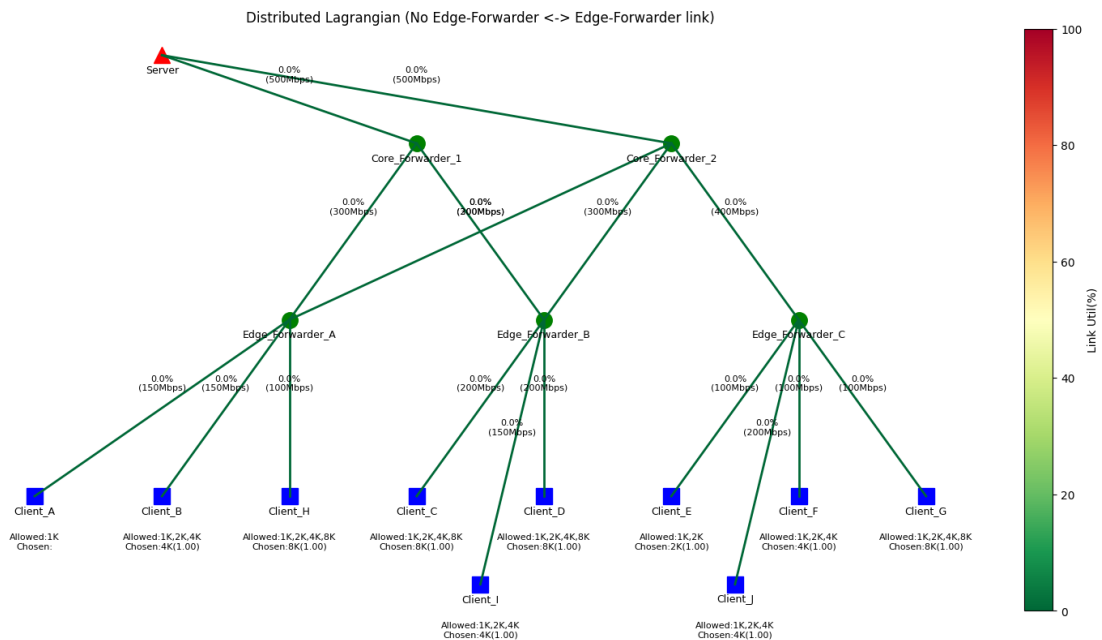
Extra simplex iterations after uncrush: 1

1	-1.5000000e+00	0.000000e+00	0.000000e+00	0s
---	----------------	--------------	--------------	----

Solved in 1 iterations and 0.01 seconds (0.00 work units)

Optimal objective -1.500000000e+00

approx Lagrange obj= -13.800



Link usage summary:

	From	To	Used	Cap	Util(%)
0	Server	Core_Forwarder_1	0.0	500	0.0
1	Server	Core_Forwarder_2	0.0	500	0.0
2	Core_Forwarder_1	Edge_Forwarder_A	0.0	300	0.0
3	Core_Forwarder_1	Edge_Forwarder_B	0.0	300	0.0
4	Core_Forwarder_2	Edge_Forwarder_A	0.0	200	0.0
5	Core_Forwarder_2	Edge_Forwarder_B	0.0	300	0.0
6	Core_Forwarder_2	Edge_Forwarder_C	0.0	400	0.0
7	Edge_Forwarder_A	Client_A	0.0	150	0.0
8	Edge_Forwarder_A	Client_B	0.0	150	0.0
9	Edge_Forwarder_B	Client_C	0.0	200	0.0
10	Edge_Forwarder_B	Client_D	0.0	200	0.0
11	Edge_Forwarder_C	Client_E	0.0	100	0.0
12	Edge_Forwarder_C	Client_F	0.0	100	0.0

13	Edge_Forwarder_C	Client_G	0.0	100	0.0
14	Edge_Forwarder_A	Client_H	0.0	100	0.0
15	Edge_Forwarder_B	Client_I	0.0	150	0.0
16	Edge_Forwarder_C	Client_J	0.0	200	0.0

Alpha:

```
{('Core_Forwarder_1', '8K'): 0.6, ('Core_Forwarder_1', '4K'): 0.6,
('Core_Forwarder_1', '2K'): 0.6, ('Core_Forwarder_1', '1K'): 0.6,
('Edge_Forwarder_C', '8K'): 0.3, ('Edge_Forwarder_C', '4K'): 0.3,
('Edge_Forwarder_C', '2K'): 0.3, ('Edge_Forwarder_C', '1K'): 0.3,
('Edge_Forwarder_A', '8K'): 0.10349999999999998, ('Edge_Forwarder_A', '4K'):
0.3, ('Edge_Forwarder_A', '2K'): 0.3, ('Edge_Forwarder_A', '1K'): 0.3,
('Core_Forwarder_2', '8K'): 0.8999999999999999, ('Core_Forwarder_2', '4K'):
0.8999999999999999, ('Core_Forwarder_2', '2K'): 0.8999999999999999,
('Core_Forwarder_2', '1K'): 0.8999999999999999, ('Edge_Forwarder_B', '8K'): 0.3,
('Edge_Forwarder_B', '4K'): 0.3, ('Edge_Forwarder_B', '2K'): 0.3,
('Edge_Forwarder_B', '1K'): 0.3, ('A', '8K'): 0.0, ('A', '4K'): 0.0, ('A',
'2K'): 0.0, ('A', '1K'): 1.2, ('B', '8K'): 0.0, ('B', '4K'): 1.8, ('B', '2K'):
0.6, ('B', '1K'): 0.0, ('C', '8K'): 1.8, ('C', '4K'): 0.6, ('C', '2K'): 0.0,
('C', '1K'): 0.0, ('D', '8K'): 1.8, ('D', '4K'): 0.6, ('D', '2K'): 0.0, ('D',
'1K'): 0.0, ('E', '8K'): 0.0, ('E', '4K'): 0.0, ('E', '2K'): 1.8, ('E', '1K'):
0.6, ('F', '8K'): 0.0, ('F', '4K'): 1.8, ('F', '2K'): 0.6, ('F', '1K'): 0.0,
('G', '8K'): 1.8, ('G', '4K'): 0.6, ('G', '2K'): 0.0, ('G', '1K'): 0.0, ('H',
'8K'): 1.8, ('H', '4K'): 0.6, ('H', '2K'): 0.0, ('H', '1K'): 0.0, ('I', '8K'):
0.0, ('I', '4K'): 1.8, ('I', '2K'): 0.6, ('I', '1K'): 0.0, ('J', '8K'): 0.0,
('J', '4K'): 1.8, ('J', '2K'): 0.6, ('J', '1K'): 0.0}
```

Obj history:

```
[-22.0, -20.200000000000003, -18.4, -15.467792500000005, -15.000000000000002,
-16.5, -15.600000000000001, -13.799999999999997]
```

[ ]:

[ ]:

[ ]:

```
[ ]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    # Resolution bandwidth requirements (Mbps)
    B = {
        '8K': 200,
        '4K': 45,
```

```

    '2K': 16,
    '1K': 8
}

# Quality scores for each resolution
q = {
    '8K': 4,
    '4K': 3,
    '2K': 2,
    '1K': 1
}

# Client maximum bitrates
max_bitrate = {
    'A': 10,
    'B': 50,
    'C': 200,
    'D': 200,
    'E': 20,
    'F': 50,
    'G': 200,
    'H': 200,
    'I': 100,
    'J': 150
}

# Link remaining estimated bandwidths (Mbps)
links = {
    ('Server', 'Core_Forwarder_1'): 500,
    ('Server', 'Core_Forwarder_2'): 500,
    ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
    ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
    ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
    ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
    ('Edge_Forwarder_A', 'Client_A'): 150,
    ('Edge_Forwarder_A', 'Client_B'): 150,
    ('Edge_Forwarder_B', 'Client_C'): 200,
    ('Edge_Forwarder_B', 'Client_D'): 200,
    ('Edge_Forwarder_C', 'Client_E'): 100,
    ('Edge_Forwarder_C', 'Client_F'): 100,
    ('Edge_Forwarder_C', 'Client_G'): 100,
    ('Edge_Forwarder_A', 'Client_H'): 100,
    ('Edge_Forwarder_B', 'Client_I'): 150,
    ('Edge_Forwarder_C', 'Client_J'): 200,
    # Parallel links
    ('Edge_Forwarder_A', 'Edge_Forwarder_B'): 100,

```

```

        ('Edge_Forwarder_B', 'Edge_Forwarder_C'): 150,
        ('Edge_Forwarder_A', 'Edge_Forwarder_C'): 120
    }

    # Define paths for each client (multiple paths are allowed)
    paths = {
        'A': [
            [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_A')],
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_A')]
        ],
        'B': [
            [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_B')],
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_B')]
        ],
        'C': [
            [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_C')],
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_C')]
        ],
        'D': [
            [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_D')],
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_D')]
        ],
        'E': [
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_E')],
        ],
        'F': [
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_F')]
        ],
        'G': [
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_G')]
        ],
        'H': [
            [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_H')]
        ],
    }

```

```

        'I': [
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2',
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_I')]
        ],
        'J': [
            [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2',
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_J')]
        ]
    }

    # Client importance weights (default to 1 for all clients)
    weights = {
        'A': 1,
        'B': 1,
        'C': 1,
        'D': 1,
        'E': 1,
        'F': 1,
        'G': 1,
        'H': 1,
        'I': 1,
        'J': 1
    }

    return B, q, max_bitrate, links, paths, weights

class VideoOptimizationModel:
    def __init__(self):
        self.B, self.q, self.max_bitrate, self.links, self.paths, self.weights,
↪ define_problem_data()
        self.clients = list(self.max_bitrate.keys())
        self.resolutions = list(self.B.keys())
        self.model = None
        self.x = None
        self.m = None
        self.b = None
        # Create the model right after initialization
        self.create_model()

    def create_model(self):
        """Create the optimization model and variables"""
        self.model = gp.Model("video_quality_optimization")

        # Create variables
        self.x = self.model.addVars(self.clients, self.resolutions, vtype=GRB.
↪ BINARY, name="x")

```



```

        self.m = self.model.addVars(self.links.keys(), self.resolutions,
↪vtype=GRB.BINARY, name="m")
        self.b = self.model.addVars(self.links.keys(), vtype=GRB.CONTINUOUS,
↪name="b")

    def set_objective(self):
        """Set the optimization objective"""
        obj = -gp.quicksum(
            self.weights[i] * self.q[j] * self.x[i, j]
            for i in self.clients for j in self.resolutions
        )
        self.model.setObjective(obj, GRB.MINIMIZE)

    def add_constraints(self):
        """Add all model constraints"""
        # User assignment constraint
        for i in self.clients:
            self.model.addConstr(
                gp.quicksum(self.x[i, j] for j in self.resolutions) <= 1,
                name=f"user_assignment_{i}"
            )

        # User capability constraint
        for i in self.clients:
            for j in self.resolutions:
                if self.B[j] > self.max_bitrate[i]:
                    self.model.addConstr(self.x[i, j] == 0,
↪name=f"user_capability_{i}_{j}")

        # Link bandwidth usage
        for l, k in self.links:
            self.model.addConstr(
                self.b[l, k] == gp.quicksum(self.m[l, k, j] * self.B[j] for j
↪in self.resolutions),
                name=f"link_bandwidth_usage_{l}_{k}"
            )

        # Link capacity constraint
        for l, k in self.links:
            self.model.addConstr(
                self.b[l, k] <= self.links[l, k],
                name=f"link_capacity_{l}_{k}"
            )

        # Multicast logic constraint
        for i in self.clients:
            for j in self.resolutions:

```

```

        for path in self.paths[i]:
            for l, k in path:
                self.model.addConstr(
                    self.m[l, k, j] >= self.x[i, j],
                    name=f"multicast_logic_{i}_{j}_{l}_{k}"
                )

def optimize(self):
    """Run the optimization"""
    self.model.optimize()

def print_results(self):
    """Print and visualize optimization results"""
    if self.model.status == GRB.OPTIMAL:
        print("\nOptimal solution found!")

        results = []
        for i in self.clients:
            for j in self.resolutions:
                if self.x[i, j].x > 0.5:
                    results.append({
                        'Client': i,
                        'Resolution': j,
                        'Bandwidth': self.B[j]
                    })

        df_results = pd.DataFrame(results)
        print("\nClient assignments:")
        print(df_results)

        link_usage = []
        for (l, k) in self.links:
            link_usage.append({
                'From': l,
                'To': k,
                'Usage (Mbps)': round(self.b[l, k].x, 2),
                'Capacity (Mbps)': self.links[l, k],
                'Utilization (%)': round(self.b[l, k].x / self.links[l, k],
↪* 100, 2)
            })

        df_links = pd.DataFrame(link_usage)
        print("\nLink bandwidth usage:")
        print(df_links)

        self.plot_network_with_utilization(df_links)

```

```

def plot_network_with_utilization(self, df_links):
    """Plot network topology with link utilization and client resolutions"""
    fig, ax = plt.subplots(figsize=(15, 10))

    pos = {
        'Server': (0.2, 0.8),
        'Core_Forwarder_1': (0.4, 0.7),
        'Core_Forwarder_2': (0.6, 0.7),
        'Edge_Forwarder_A': (0.3, 0.5),
        'Edge_Forwarder_B': (0.5, 0.5),
        'Edge_Forwarder_C': (0.7, 0.5),
        'Client_A': (0.1, 0.3),
        'Client_B': (0.2, 0.3),
        'Client_C': (0.4, 0.3),
        'Client_D': (0.5, 0.3),
        'Client_E': (0.6, 0.3),
        'Client_F': (0.7, 0.3),
        'Client_G': (0.8, 0.3),
        'Client_H': (0.3, 0.3),
        'Client_I': (0.45, 0.2),
        'Client_J': (0.65, 0.2),
    }

    client_resolutions = {}
    for i in self.clients:
        for j in self.resolutions:
            if self.x[i, j].x > 0.5:
                client_resolutions[i] = j

    utilization_dict = {(row['From'], row['To']): row['Utilization (%)']}
    for _, row in df_links.iterrows():

        def get_color(util_pct):
            return plt.cm.RdYlGn_r(util_pct / 100)

        for node, (x, y) in pos.items():
            if 'Client' in node:
                ax.plot(x, y, 'gs', markersize=20)
                ax.text(x, y - 0.02, node, ha='center')
                client_id = node.split('_')[1]
                resolution = client_resolutions.get(client_id, 'N/A')
                ax.text(x, y - 0.05, f'Max: {self.max_bitrate[client_id]}
↳Mbps\nAssigned: {resolution}', ha='center', fontsize=8)
            elif 'Forwarder' in node:
                ax.plot(x, y, 'bo', markersize=20)
                ax.text(x, y - 0.02, node, ha='center')
            else:

```

```

        ax.plot(x, y, 'rd', markersize=20)
        ax.text(x, y - 0.02, node, ha='center')

    for (l, k), util_pct in utilization_dict.items():
        x1, y1 = pos[l]
        x2, y2 = pos[k]
        ax.plot([x1, x2], [y1, y2], '-', color=get_color(util_pct),
↪linewidth=2)
        label = f'{round(util_pct, 1)}%\n({self.links[l, k]} Mbps)'
        ax.text((x1 + x2) / 2, (y1 + y2) / 2 + 0.02, label)

    norm = plt.Normalize(0, 100)
    sm = plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
    sm.set_array([])
    plt.colorbar(sm, ax=ax, label='Link Utilization (%)')

    ax.text(0.2, 0.85, 'Resolutions:', ha='left')
    ax.text(0.2, 0.83, '8K -- 200 Mbps', ha='left')
    ax.text(0.2, 0.81, '4K -- 45 Mbps', ha='left')
    ax.text(0.2, 0.79, '2K -- 16 Mbps', ha='left')
    ax.text(0.2, 0.77, '1K -- 8 Mbps', ha='left')

    ax.set_title('Network Topology with Link Utilization and Client_
↪Resolutions')
    ax.axis('off')
    plt.tight_layout()
    plt.show()

def run_optimization():
    model = VideoOptimizationModel()
    model.set_objective()
    model.add_constraints()
    model.optimize()
    model.print_results()

if __name__ == "__main__":
    run_optimization()

```

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 227 rows, 140 columns and 505 nonzeros

Model fingerprint: 0xff9c2e42

Variable types: 20 continuous, 120 integer (120 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 5e+02]

Found heuristic solution: objective 0.0000000

Presolve removed 227 rows and 140 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)

Thread count was 1 (of 64 available processors)

Solution count 2: -29 0

No other solutions better than -29

Optimal solution found (tolerance 1.00e-04)

Best objective -2.9000000000000e+01, best bound -2.9000000000000e+01, gap 0.0000%

Optimal solution found!

Client assignments:

	Client	Resolution	Bandwidth
0	A	1K	8
1	B	4K	45
2	C	8K	200
3	D	8K	200
4	E	2K	16
5	F	4K	45
6	G	4K	45
7	H	4K	45
8	I	4K	45
9	J	4K	45

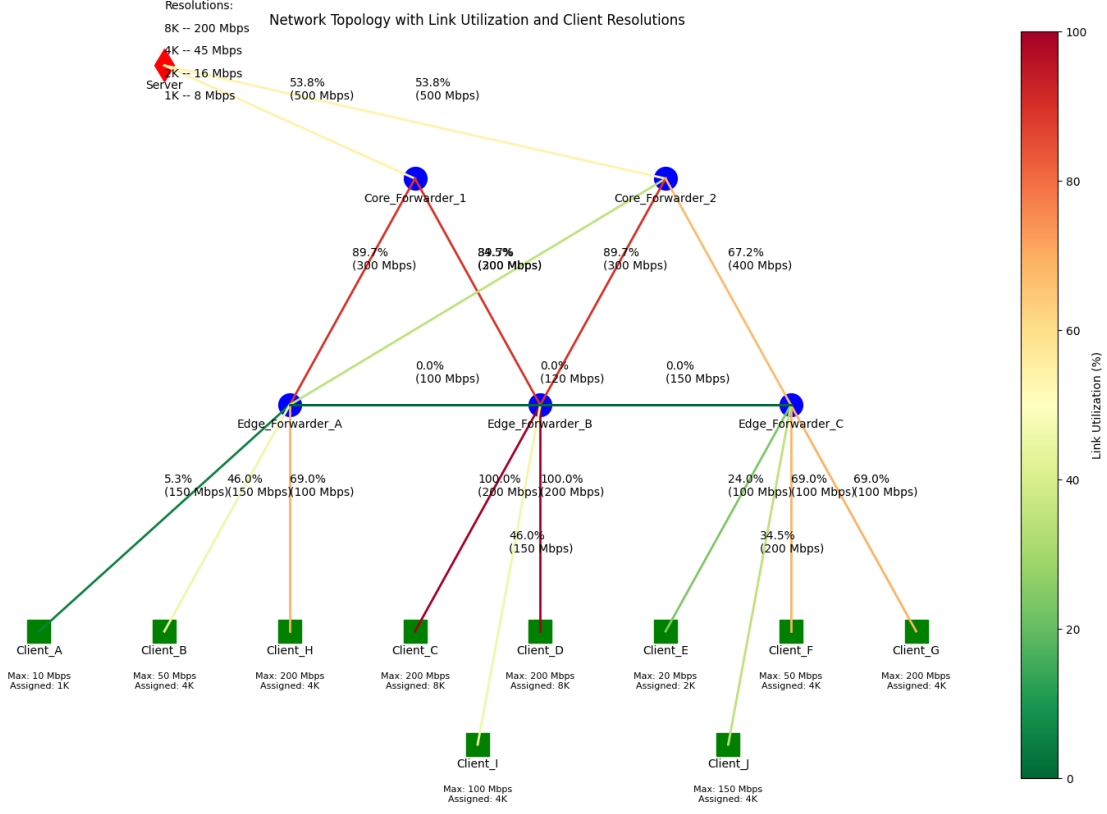
Link bandwidth usage:

	From	To	Usage (Mbps)	Capacity (Mbps)	\
0	Server	Core_Forwarder_1	269.0	500	
1	Server	Core_Forwarder_2	269.0	500	
2	Core_Forwarder_1	Edge_Forwarder_A	269.0	300	
3	Core_Forwarder_1	Edge_Forwarder_B	269.0	300	
4	Core_Forwarder_2	Edge_Forwarder_A	69.0	200	
5	Core_Forwarder_2	Edge_Forwarder_B	269.0	300	
6	Core_Forwarder_2	Edge_Forwarder_C	269.0	400	
7	Edge_Forwarder_A	Client_A	8.0	150	
8	Edge_Forwarder_A	Client_B	69.0	150	
9	Edge_Forwarder_B	Client_C	200.0	200	

10	Edge_Forwarder_B	Client_D	200.0	200
11	Edge_Forwarder_C	Client_E	24.0	100
12	Edge_Forwarder_C	Client_F	69.0	100
13	Edge_Forwarder_C	Client_G	69.0	100
14	Edge_Forwarder_A	Client_H	69.0	100
15	Edge_Forwarder_B	Client_I	69.0	150
16	Edge_Forwarder_C	Client_J	69.0	200
17	Edge_Forwarder_A	Edge_Forwarder_B	0.0	100
18	Edge_Forwarder_B	Edge_Forwarder_C	0.0	150
19	Edge_Forwarder_A	Edge_Forwarder_C	0.0	120

#### Utilization (%)

0	53.80
1	53.80
2	89.67
3	89.67
4	34.50
5	89.67
6	67.25
7	5.33
8	46.00
9	100.00
10	100.00
11	24.00
12	69.00
13	69.00
14	69.00
15	46.00
16	34.50
17	0.00
18	0.00
19	0.00



[ ]:

### 3 Distributed Optimization Model and System Architecture

#### 3.1 Distributed Down-to-Top Optimization Model

The distributed model adopts a hierarchical approach where forwarders independently optimize their local configurations based on downstream clients. Results are aggregated and propagated upward, ensuring global consistency. This approach reduces computational complexity while maintaining near-optimal performance.

##### 3.1.1 Objective Function

$$\text{Maximize: } \sum_{i \in I_r} w_i \cdot \sum_j q_j \cdot x_{ij}^r$$

where: -  $I_r$  is the set of users managed by forwarder  $r$  -  $x_{ij}^r \in \{0,1\}$  indicates whether forwarder  $r$  assigns resolution  $j$  to user  $i$

### 3.1.2 Constraints

1. Each user can be assigned at most one resolution by the forwarder:

$$\sum_j x_{ij}^r \leq 1, \quad \forall i \in I_r$$

2. Resolutions cannot exceed the user's bandwidth capability:

$$x_{ij}^r = 0, \quad \forall j \text{ such that } B(j) > \text{max\_bitrate}(i)$$

3. Bandwidth usage on each link must not exceed capacity:

$$b_{lk}^r = \sum_{i \in I_r} \sum_j x_{ij}^r B(j), \quad b_{lk}^r \leq C_{lk}, \quad \forall (l, k) \in N_r$$

4. Higher-layer forwarders must respect constraints aggregated from lower layers:

$$x_{ij}^r \geq x_{ij}^{\text{lower}}, \quad \forall i \in I_r, \forall j$$

### 3.1.3 Complexity Analysis

For a forwarder managing  $n_r$  users and  $L_r$  links:

$$\text{Variables: } n_r \cdot k + L_r + L_r \cdot k$$

$$\text{Constraints: } n_r + n_r \cdot k + 2L_r$$

Since  $n_r \ll n$  and  $L_r \ll L$ , the complexity per forwarder is orders of magnitude lower than in the centralized model. The total complexity is approximately:

$$O\left(\sum_r 2^{V_r} \cdot \text{poly}(C_r)\right)$$

where  $V_r$  and  $C_r$  are the variables and constraints for each forwarder  $r$ . This localized optimization enables near real-time computation even in large networks.

```
[8]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt

def define_problem_data():
    # Resolution bandwidth requirements (Mbps)
    B = {
        '8K': 200,
        '4K': 45,
```



```

        '2K': 16,
        '1K': 8
    }

    # Quality scores for each resolution
    q = {
        '8K': 4,
        '4K': 3,
        '2K': 2,
        '1K': 1
    }

    # Client maximum bitrates
    max_bitrate = {
        'A': 10,
        'B': 50,
        'C': 200,
        'D': 200,
        'E': 20,
        'F': 50,
        'G': 200,
        'H': 200,
        'I': 100,
        'J': 150
    }

    # Link remaining estimated bandwidths (Mbps)
    links = {
        ('Server', 'Core_Forwarder_1'): 500,
        ('Server', 'Core_Forwarder_2'): 500,
        ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
        ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
        ('Core_Forwarder_2', 'Edge_Forwarder_A'): 200,
        ('Core_Forwarder_2', 'Edge_Forwarder_B'): 300,
        ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
        ('Edge_Forwarder_A', 'Client_A'): 150,
        ('Edge_Forwarder_A', 'Client_B'): 150,
        ('Edge_Forwarder_B', 'Client_C'): 200,
        ('Edge_Forwarder_B', 'Client_D'): 200,
        ('Edge_Forwarder_C', 'Client_E'): 100,
        ('Edge_Forwarder_C', 'Client_F'): 100,
        ('Edge_Forwarder_C', 'Client_G'): 100,
        ('Edge_Forwarder_A', 'Client_H'): 100,
        ('Edge_Forwarder_B', 'Client_I'): 150,
        ('Edge_Forwarder_C', 'Client_J'): 200
    }

```

```

# Client importance weights (default to 1 for all clients)
weights = {client: 1 for client in max_bitrate.keys()}

return B, q, max_bitrate, links, weights

class VideoOptimizationModel:
    def __init__(self, forwarder_name, clients, links, edge_results=None):
        self.forwarder_name = forwarder_name
        self.B, self.q, self.max_bitrate, self.all_links, self.weights = _
    ↪ define_problem_data()
        self.clients = clients
        self.resolutions = list(self.B.keys())
        self.links = links
        self.edge_results = edge_results or {} # Edge
        self.model = None
        self.x = None
        self.create_model()

    def create_model(self):
        self.model = gp.Model(f"optimization_{self.forwarder_name}")
        #      i      j
        self.x = self.model.addVars(self.clients, self.resolutions, vtype=GRB.
    ↪ BINARY, name="x")

    def set_objective(self):
        #
        obj = gp.quicksum(self.weights[i] * self.q[j] * self.x[i, j]
                           for i in self.clients for j in self.resolutions)
        self.model.setObjective(obj, GRB.MAXIMIZE)

    def add_constraints(self):
        """Add constraints for the model"""
        # 1.
        for i in self.clients:
            self.model.addConstr(
                gp.quicksum(self.x[i, j] for j in self.resolutions) <= 1,
                name=f"user_assignment_{i}"
            )

        # 2.
        for i in self.clients:
            for j in self.resolutions:
                if self.B[j] > self.max_bitrate[i]:
                    self.model.addConstr(
                        self.x[i, j] == 0,
                        name=f"user_capability_{i}_{j}"
                    )

```

```

# 3.
for (l, k), capacity in self.links.items():
    if 'Client_' in k:
        #
        client = k.split('_')[1]
        self.model.addConstr(
            gp.quicksum(self.x[client, j] * self.B[j]
                        for j in self.resolutions) <= capacity,
            name=f"client_link_capacity_{client}"
        )
    else:
        #
        use_resolution = self.model.addVars(
            self.resolutions, vtype=GRB.BINARY,
            name=f"use_resolution_{l}_{k}"
        )

        #          use_resolution 1
        for j in self.resolutions:
            for i in self.clients:
                self.model.addConstr(
                    use_resolution[j] >= self.x[i, j],
                    name=f"resolution_usage_{l}_{k}_{j}_{i}"
                )

        #
        self.model.addConstr(
            gp.quicksum(use_resolution[j] * self.B[j]
                        for j in self.resolutions) <= capacity,
            name=f"link_capacity_{l}_{k}"
        )

# 4. Core Edge
if 'Core_Forwarder' in self.forwarder_name and self.edge_results:
    for i in self.clients:
        for j in self.resolutions:
            if (i, j) in self.edge_results:
                self.model.addConstr(
                    self.x[i, j] >= self.edge_results[(i, j)],
                    name=f"edge_demand_{i}_{j}"
                )

def optimize(self):
    self.model.optimize()

def get_results(self):

```

```

        """Return the results in terms of decision variables"""
        results = {(i, j): self.x[i, j].x
                    for i in self.clients
                    for j in self.resolutions
                    if self.x[i, j].x > 0.5}
        return results

def print_results(self):
    """Print results"""
    if self.model.status == GRB.OPTIMAL:
        print(f"\nResults for {self.forwarder_name}:")
        for (i, j), value in self.get_results().items():
            print(f"Client {i} assigned resolution {j} (x={value})")
    else:
        print(f"No optimal solution found for {self.forwarder_name}.")

def plot_final_results(all_results):
    """Plot final network topology with all results"""
    fig, ax = plt.subplots(figsize=(15, 10))

    # Define node positions
    pos = {
        'Server': (0.5, 1.0),
        'Core_Forwarder_1': (0.3, 0.7),
        'Core_Forwarder_2': (0.7, 0.7),
        'Edge_Forwarder_A': (0.2, 0.4),
        'Edge_Forwarder_B': (0.5, 0.4),
        'Edge_Forwarder_C': (0.8, 0.4)
    }

    # Add client positions
    client_positions = {
        'A': (0.1, 0.1), 'B': (0.2, 0.1), 'C': (0.4, 0.1),
        'D': (0.5, 0.1), 'E': (0.6, 0.1), 'F': (0.7, 0.1),
        'G': (0.8, 0.1), 'H': (0.3, 0.1), 'I': (0.45, 0.0),
        'J': (0.65, 0.0)
    }

    pos.update({f'Client_{client}': position for client, position in
    ↪client_positions.items()})

    # Get problem data
    B, q, max_bitrate, links, weights = define_problem_data()

    # Initialize link usage dictionary
    link_usage = {}
    for link in links.keys():
        link_usage[link] = 0

```

```

# Define client paths through the network
client_paths = {
    'A': [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_A')],
    'B': [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_B')],
    'C': [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_C')],
    'D': [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_D')],
    'E': [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_E')],
    'F': [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_F')],
    'G': [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_G')],
    'H': [('Server', 'Core_Forwarder_1'), ('Core_Forwarder_1', □
↪ 'Edge_Forwarder_A'), ('Edge_Forwarder_A', 'Client_H')],
    'I': [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_B'), ('Edge_Forwarder_B', 'Client_I')],
    'J': [('Server', 'Core_Forwarder_2'), ('Core_Forwarder_2', □
↪ 'Edge_Forwarder_C'), ('Edge_Forwarder_C', 'Client_J')]
}

# Create dictionary to store final client assignments
final_assignments = {}
# First get edge forwarder results (they take precedence)
for forwarder_name, results in all_results.items():
    if 'Edge_Forwarder' in forwarder_name:
        for (client, res), value in results.items():
            if value > 0.5:
                final_assignments[client] = res

# Update link usage based on final assignments and paths
unique_resolutions_per_link = {}
for link in links:
    unique_resolutions_per_link[link] = set()

for client_id, path in client_paths.items():
    resolution = final_assignments.get(client_id)
    if resolution:
        bandwidth = B[resolution]
        for link in path:
            if link in links:
                #

```

```

        unique_resolutions_per_link[link].add(resolution)

#
for link in links:
    #
    link_usage[link] = sum(B[res] for res in
↪unique_resolutions_per_link[link])

# Plot nodes and links
for node, (x, y) in pos.items():
    if 'Client' in node:
        color = 'lightgreen'
        size = 100
        client_id = node.split('_')[1]
        resolution = final_assignments.get(client_id)

        ax.scatter(x, y, c=color, s=size, zorder=2)
        ax.text(x, y-0.03, node, ha='center', va='top')
        if resolution:
            bandwidth = B[resolution]
            ax.text(x, y-0.07, f'{resolution}\n({bandwidth} Mbps)\nMax:
↪{max_bitrate[client_id]} Mbps',
                    ha='center', va='top', fontsize=8)

    elif 'Forwarder' in node:
        color = 'lightblue'
        size = 200
        ax.scatter(x, y, c=color, s=size, zorder=2)
        ax.text(x, y-0.03, node, ha='center', va='top')

    else: # Server
        color = 'salmon'
        size = 200
        ax.scatter(x, y, c=color, s=size, zorder=2)
        ax.text(x, y-0.03, node, ha='center', va='top')

# Plot links with utilization colors
for (l, k), capacity in links.items():
    if l in pos and k in pos:
        start_pos = pos[l]
        end_pos = pos[k]

        usage = link_usage.get((l, k), 0)
        utilization = (usage / capacity) * 100
        color = plt.cm.RdYlGn_r(utilization / 100)

        ax.plot([start_pos[0], end_pos[0]], [start_pos[1], end_pos[1]],

```

```

        color=color, linewidth=2, zorder=1)

    mid_x = (start_pos[0] + end_pos[0]) / 2
    mid_y = (start_pos[1] + end_pos[1]) / 2
    ax.text(mid_x, mid_y+0.02, f'{round(utilization)}%\n({round(usage)}/{
↪{capacity} Mbps)',
            ha='center', va='bottom', fontsize=8)

    # Add color bar for utilization
    norm = plt.Normalize(0, 100)
    sm = plt.cm.ScalarMappable(cmap=plt.cm.RdYlGn_r, norm=norm)
    plt.colorbar(sm, ax=ax, label='Link Utilization (%)')

    # Add resolution legend
    resolutions = [('8K', 200), ('4K', 45), ('2K', 16), ('1K', 8)]
    legend_text = 'Available Resolutions:\n' + '\n'.join(f'{res} -- {bw} Mbps'
↪for res, bw in resolutions)
    ax.text(0.02, 0.98, legend_text, transform=ax.transAxes,
            ha='left', va='top', fontsize=8, bbox=dict(facecolor='white',
↪alpha=0.8))

    # Set title and layout
    ax.set_title('Final Network Topology with Link Utilization and Client
↪Resolutions')
    ax.set_xlim(-0.1, 1.1)
    ax.set_ylim(-0.1, 1.1)
    ax.axis('off')
    plt.tight_layout()
    plt.show()

    # Print tabulated results
    print("\nFinal Client Assignments:")
    results_table = []
    for client_id, resolution in final_assignments.items():
        results_table.append({
            'Client': client_id,
            'Resolution': resolution,
            'Bandwidth': B[resolution],
            'Max Bitrate': max_bitrate[client_id],
            'Utilization (%)': round(B[resolution] / max_bitrate[client_id] *
↪100, 1)
        })

    df_results = pd.DataFrame(results_table)
    # Sort results by client ID
    df_results = df_results.sort_values('Client')
    print(df_results.to_string(index=False))

```

```

def run_distributed_optimization():
    # Define the clients for each forwarder
    forwarder_clients = {
        "Core_Forwarder_1": ['A', 'B', 'C', 'D', 'H'],
        "Core_Forwarder_2": ['E', 'F', 'G', 'I', 'J'],
        "Edge_Forwarder_A": ['A', 'B', 'H'],
        "Edge_Forwarder_B": ['C', 'D', 'I'],
        "Edge_Forwarder_C": ['E', 'F', 'G', 'J']
    }

    # Define links for each forwarder
    forwarder_links = {
        "Core_Forwarder_1": {('Server', 'Core_Forwarder_1'): 500},
        "Core_Forwarder_2": {('Server', 'Core_Forwarder_2'): 500},
        "Edge_Forwarder_A": {
            ('Core_Forwarder_1', 'Edge_Forwarder_A'): 300,
            ('Edge_Forwarder_A', 'Client_A'): 150,
            ('Edge_Forwarder_A', 'Client_B'): 150,
            ('Edge_Forwarder_A', 'Client_H'): 100
        },
        "Edge_Forwarder_B": {
            ('Core_Forwarder_1', 'Edge_Forwarder_B'): 300,
            ('Edge_Forwarder_B', 'Client_C'): 200,
            ('Edge_Forwarder_B', 'Client_D'): 200,
            ('Edge_Forwarder_B', 'Client_I'): 150
        },
        "Edge_Forwarder_C": {
            ('Core_Forwarder_2', 'Edge_Forwarder_C'): 400,
            ('Edge_Forwarder_C', 'Client_E'): 100,
            ('Edge_Forwarder_C', 'Client_F'): 100,
            ('Edge_Forwarder_C', 'Client_G'): 100,
            ('Edge_Forwarder_C', 'Client_J'): 200
        }
    }

    # Step 1: Edge Forwarders Optimization
    edge_forwarder_a = VideoOptimizationModel(
        "Edge_Forwarder_A",
        forwarder_clients["Edge_Forwarder_A"],
        forwarder_links["Edge_Forwarder_A"]
    )
    edge_forwarder_a.set_objective()
    edge_forwarder_a.add_constraints()
    edge_forwarder_a.optimize()
    edge_results_a = edge_forwarder_a.get_results()
    edge_forwarder_a.print_results()

```



```

edge_forwarder_b = VideoOptimizationModel(
    "Edge_Forwarder_B",
    forwarder_clients["Edge_Forwarder_B"],
    forwarder_links["Edge_Forwarder_B"]
)
edge_forwarder_b.set_objective()
edge_forwarder_b.add_constraints()
edge_forwarder_b.optimize()
edge_results_b = edge_forwarder_b.get_results()
edge_forwarder_b.print_results()

edge_forwarder_c = VideoOptimizationModel(
    "Edge_Forwarder_C",
    forwarder_clients["Edge_Forwarder_C"],
    forwarder_links["Edge_Forwarder_C"]
)
edge_forwarder_c.set_objective()
edge_forwarder_c.add_constraints()
edge_forwarder_c.optimize()
edge_results_c = edge_forwarder_c.get_results()
edge_forwarder_c.print_results()

# Step 2: Core Forwarders Optimization
# Edge A B Core 1
edge_results_for_core1 = {**edge_results_a, **edge_results_b}
core_forwarder_1 = VideoOptimizationModel(
    "Core_Forwarder_1",
    forwarder_clients["Core_Forwarder_1"],
    forwarder_links["Core_Forwarder_1"],
    edge_results_for_core1
)
core_forwarder_1.set_objective()
core_forwarder_1.add_constraints()
core_forwarder_1.optimize()
core_forwarder_1.print_results()

# Edge C Core 2
edge_results_for_core2 = {**edge_results_c}
core_forwarder_2 = VideoOptimizationModel(
    "Core_Forwarder_2",
    forwarder_clients["Core_Forwarder_2"],
    forwarder_links["Core_Forwarder_2"],
    edge_results_for_core2
)
core_forwarder_2.set_objective()
core_forwarder_2.add_constraints()

```

```

core_forwarder_2.optimize()
core_forwarder_2.print_results()

# Collect all results
all_results = {
    'Edge_Forwarder_A': edge_results_a,
    'Edge_Forwarder_B': edge_results_b,
    'Edge_Forwarder_C': edge_results_c,
    'Core_Forwarder_1': core_forwarder_1.get_results(),
    'Core_Forwarder_2': core_forwarder_2.get_results()
}

# Plot final results
plot_final_results(all_results)

if __name__ == "__main__":
    run_distributed_optimization()

```

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 23 rows, 16 columns and 56 nonzeros

Model fingerprint: 0x3ef3ccad

Variable types: 0 continuous, 16 integer (16 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 3e+02]

Found heuristic solution: objective -0.0000000

Presolve removed 23 rows and 16 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)

Thread count was 1 (of 64 available processors)

Solution count 2: 7 -0

Optimal solution found (tolerance 1.00e-04)

Best objective 7.000000000000e+00, best bound 7.000000000000e+00, gap 0.0000%

Results for Edge\_Forwarder\_A:

Client A assigned resolution 1K (x=1.0)  
Client B assigned resolution 4K (x=1.0)  
Client H assigned resolution 4K (x=1.0)  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 20 rows, 16 columns and 53 nonzeros  
Model fingerprint: 0x70d50b1b  
Variable types: 0 continuous, 16 integer (16 binary)  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [1e+00, 4e+00]  
Bounds range [1e+00, 1e+00]  
RHS range [1e+00, 3e+02]  
Found heuristic solution: objective -0.0000000  
Presolve removed 20 rows and 16 columns  
Presolve time: 0.00s  
Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)  
Thread count was 1 (of 64 available processors)

Solution count 2: 11 -0

Optimal solution found (tolerance 1.00e-04)  
Best objective 1.100000000000e+01, best bound 1.100000000000e+01, gap 0.0000%

Results for Edge\_Forwarder\_B:  
Client C assigned resolution 8K (x=1.0)  
Client D assigned resolution 8K (x=1.0)  
Client I assigned resolution 4K (x=1.0)  
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]  
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 29 rows, 20 columns and 72 nonzeros  
Model fingerprint: 0xa24bc74d  
Variable types: 0 continuous, 20 integer (20 binary)  
Coefficient statistics:  
Matrix range [1e+00, 2e+02]  
Objective range [1e+00, 4e+00]

```

    Bounds range      [1e+00, 1e+00]
    RHS range         [1e+00, 4e+02]
Found heuristic solution: objective -0.0000000
Presolve removed 29 rows and 20 columns
Presolve time: 0.00s
Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)
Thread count was 1 (of 64 available processors)

Solution count 2: 11 -0

Optimal solution found (tolerance 1.00e-04)
Best objective 1.1000000000000e+01, best bound 1.1000000000000e+01, gap 0.0000%

Results for Edge_Forwarder_C:
Client E assigned resolution 2K (x=1.0)
Client F assigned resolution 4K (x=1.0)
Client G assigned resolution 4K (x=1.0)
Client J assigned resolution 4K (x=1.0)
Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5
LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set
[SSE2|AVX|AVX2|AVX512]
Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 35 rows, 24 columns and 73 nonzeros
Model fingerprint: 0x9c8419ab
Variable types: 0 continuous, 24 integer (24 binary)
Coefficient statistics:
    Matrix range      [1e+00, 2e+02]
    Objective range   [1e+00, 4e+00]
    Bounds range      [1e+00, 1e+00]
    RHS range         [1e+00, 5e+02]
Found heuristic solution: objective 15.0000000
Presolve removed 35 rows and 24 columns
Presolve time: 0.00s
Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.00 seconds (0.00 work units)
Thread count was 1 (of 64 available processors)

Solution count 1: 15

Optimal solution found (tolerance 1.00e-04)
Best objective 1.5000000000000e+01, best bound 1.5000000000000e+01, gap 0.0000%

```

Results for Core\_Forwarder\_1:

Client A assigned resolution 1K (x=1.0)

Client B assigned resolution 4K (x=1.0)

Client C assigned resolution 8K (x=1.0)

Client D assigned resolution 8K (x=1.0)

Client H assigned resolution 4K (x=1.0)

Gurobi Optimizer version 12.0.0 build v12.0.0rc1 (linux64 - "Ubuntu 22.04.5 LTS")

CPU model: Intel(R) Xeon(R) Platinum 8358P CPU @ 2.60GHz, instruction set [SSE2|AVX|AVX2|AVX512]

Thread count: 64 physical cores, 64 logical processors, using up to 32 threads

Optimize a model with 35 rows, 24 columns and 73 nonzeros

Model fingerprint: 0x1aa6d0ff

Variable types: 0 continuous, 24 integer (24 binary)

Coefficient statistics:

Matrix range [1e+00, 2e+02]

Objective range [1e+00, 4e+00]

Bounds range [1e+00, 1e+00]

RHS range [1e+00, 5e+02]

Found heuristic solution: objective 14.0000000

Presolve removed 35 rows and 24 columns

Presolve time: 0.00s

Presolve: All rows and columns removed

Explored 0 nodes (0 simplex iterations) in 0.01 seconds (0.00 work units)

Thread count was 1 (of 64 available processors)

Solution count 1: 14

Optimal solution found (tolerance 1.00e-04)

Best objective 1.400000000000e+01, best bound 1.400000000000e+01, gap 0.0000%

Results for Core\_Forwarder\_2:

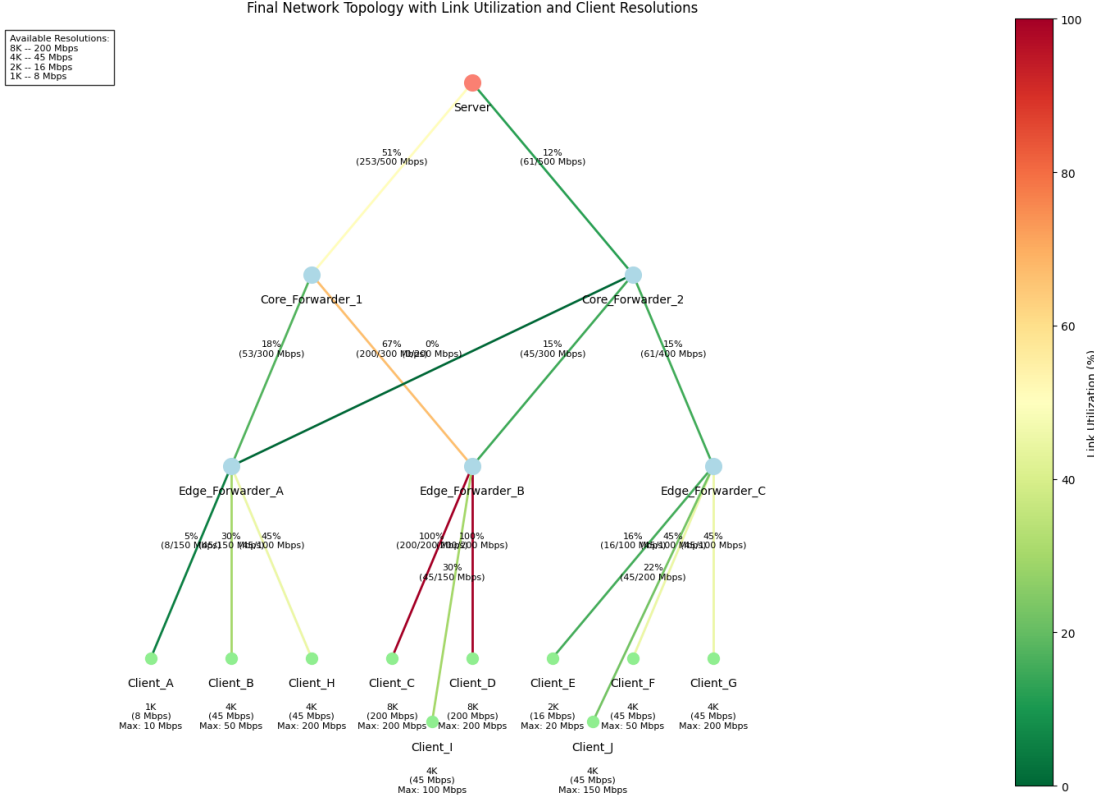
Client E assigned resolution 2K (x=1.0)

Client F assigned resolution 4K (x=1.0)

Client G assigned resolution 4K (x=1.0)

Client I assigned resolution 4K (x=1.0)

Client J assigned resolution 4K (x=1.0)



Final Client Assignments:

Client	Resolution	Bandwidth	Max Bitrate	Utilization (%)
A	1K	8	10	80.0
B	4K	45	50	90.0
C	8K	200	200	100.0
D	8K	200	200	100.0
E	2K	16	20	80.0
F	4K	45	50	90.0
G	4K	45	200	22.5
H	4K	45	200	22.5
I	4K	45	100	45.0
J	4K	45	150	30.0

## 4 Code Implementation Analysis

### 4.1 Overview

This implementation provides a scalability analysis framework for our distributed video optimization system. The code simulates and evaluates how the system performs with increasing numbers of users, demonstrating the effectiveness of our hierarchical optimization approach.

## 4.2 Implementation

### 4.2.1 1. Network Topology Generation

The system dynamically generates a hierarchical network topology based on the number of users:

- **Edge Layer Scaling:**
  - Each Edge forwarder manages ~50 users
  - Number of Edge forwarders =  $\text{ceil}(\text{num\_users} / 50)$
- **Core Layer Scaling:**
  - Each Core forwarder manages ~10 Edge forwarders
  - Number of Core forwarders =  $\text{ceil}(\text{num\_edge} / 10)$
- **Link Capacities:**
  - Server  $\rightarrow$  Core: 800 Mbps
  - Core  $\rightarrow$  Edge: 300 Mbps
  - Edge  $\rightarrow$  Client: Randomly assigned {100, 150, 200} Mbps

### 4.2.2 2. Optimization Process

The optimization follows a bottom-up approach:

1. **Edge Layer Optimization**
  - Processes batches of 50 users independently
  - Considers individual user constraints and preferences
  - Optimizes video resolution assignments for local user groups
2. **Core Layer Optimization**
  - Aggregates results from Edge forwarders
  - Ensures bandwidth constraints across the network
  - Implements multicast optimization
  - Maintains consistency with Edge layer decisions

### 4.2.3 3. Performance Metrics

The implementation tracks several key metrics: - Edge layer optimization time - Core layer optimization time - Total optimization time

### 4.2.4 Performance Results

- Shows linear scaling in optimization time with user count

### 4.2.5 Current Limitation

The implementation has a practical limitation due to the Gurobi license restrictions: - Maximum testable user count: ~300 users - Error message when exceeded: “Model too large for size-limited license”

However, this limitation doesn’t affect the validity of our analysis, as we can already observe the scaling patterns within this range.

```
[2]: import gurobipy as gp
      from gurobipy import GRB
      import numpy as np
```

```

import time
import matplotlib.pyplot as plt
import pandas as pd
from tqdm import tqdm

# Define problem data
def define_problem_data(num_users):
    B = {'8K': 200, '4K': 45, '2K': 16, '1K': 8} # Bandwidth requirements
    q = {'8K': 4, '4K': 3, '2K': 2, '1K': 1} # Quality scores

    max_bitrate = {f'Client_User_{i}': np.random.choice([10, 50, 200]) for i in
    ↪range(num_users)}
    weights = {f'Client_User_{i}': 1 for i in range(num_users)} # Equal
    ↪weights for all users
    links = generate_network_topology(num_users)
    return B, q, max_bitrate, links, weights

# Generate network topology dynamically
def generate_network_topology(num_users):
    links = {}
    num_edge = max(num_users // 50, 1) # Each Edge forwarder handles ~50 users
    num_core = max(num_edge // 10, 1) # Each Core forwarder handles ~10 Edge
    ↪forwarders

    # Server to Core
    for i in range(num_core):
        links[(f'Server', f'Core_Forwarder_{i}')] = 800

    # Core to Edge
    for i in range(num_edge):
        core_id = i % num_core
        links[(f'Core_Forwarder_{core_id}', f'Edge_Forwarder_{i}')] = 300

    # Edge to Client
    for i in range(num_users):
        client_id = f'User_{i}'
        edge_id = i // 50
        links[(f'Edge_Forwarder_{edge_id}', f'Client_User_{client_id}')] = np.
    ↪random.choice([100, 150, 200])

    return links

class VideoOptimizationModel:
    def __init__(self, forwarder_name, clients, problem_data,
    ↪lower_layer_results=None):
        self.forwarder_name = forwarder_name
        self.clients = clients

```



```

        self.B, self.q, self.max_bitrate, self.links, self.weights = _
↪problem_data
        self.resolutions = list(self.B.keys())
        self.lower_layer_results = lower_layer_results or {}
        self.model = None
        self.x = None
        self.use_resolution = {} # For multicast constraints
        self.create_model()

    def create_model(self):
        self.model = gp.Model(f"optimization_{self.forwarder_name}")
        self.model.setParam('OutputFlag', 0) # Disable solver output
        self.x = self.model.addVars(self.clients, self.resolutions, vtype=GRB.
↪BINARY, name="x")

    def set_objective(self):
        obj = gp.quicksum(self.weights[i] * self.q[j] * self.x[i, j]
                           for i in self.clients for j in self.resolutions)
        self.model.setObjective(obj, GRB.MAXIMIZE)

    def add_constraints(self):
        # Each client can only get one resolution
        for i in self.clients:
            self.model.addConstr(
                gp.quicksum(self.x[i, j] for j in self.resolutions) <= 1,
                name=f"single_resolution_{i}"
            )

        # Client bandwidth limits
        for i in self.clients:
            for j in self.resolutions:
                if self.B[j] > self.max_bitrate[i]:
                    self.model.addConstr(self.x[i, j] == 0, _
↪name=f"bandwidth_limit_{i}_{j}")

        # Link capacity constraints
        for (l, k), capacity in self.links.items():
            if k in self.clients:
                # Client link: simple sum of bandwidths
                client = k # 'Client_User_0' or 'Edge_Forwarder_0'
                self.model.addConstr(
                    gp.quicksum(self.x[client, j] * self.B[j] for j in self.
↪resolutions) <= capacity,
                    name=f"link_capacity_{l}_{k}"
                )
            else:
                # Forwarder link: multicast constraints

```

```

        # Introduce binary variables to indicate if a resolution is
        ↪used on this link
        for j in self.resolutions:
            var = self.model.addVar(vtype=GRB.BINARY,
            ↪name=f"use_resolution_{l}_{k}_{j}")
            self.use_resolution[(l, k, j)] = var
            # Link usage for resolution j must be >= any client's usage
            self.model.addConstr(
                var >= gp.quicksum(self.x[i, j] for i in self.clients
            ↪if i in self.clients),
                name=f"multicast_usage_{l}_{k}_{j}"
            )
            # Total bandwidth used on the link
            self.model.addConstr(
                gp.quicksum(self.use_resolution[(l, k, j)] * self.B[j] for
            ↪j in self.resolutions) <= capacity,
                name=f"multicast_capacity_{l}_{k}"
            )

        # Lower layer results
        for (client, resolution), value in self.lower_layer_results.items():
            if client in self.clients and resolution in self.resolutions:
                self.model.addConstr(
                    self.x[client, resolution] >= value,
                    name=f"lower_layer_demand_{client}_{resolution}"
                )

    def optimize(self):
        self.set_objective()
        self.add_constraints()
        self.model.optimize()

    def get_results(self):
        if self.model.status == GRB.OPTIMAL:
            return {(i, j): self.x[i, j].x
                    for i in self.clients
                    for j in self.resolutions
                    if self.x[i, j].x > 0.5}
        return {}

class DistributedPerformanceAnalyzer:
    def __init__(self):
        self.results = []

    def analyze_scale(self, num_users):
        problem_data = define_problem_data(num_users)
        B, q, max_bitrate, links, weights = problem_data

```

```

    clients = [f'Client_User_{i}' for i in range(num_users)] # Ensure
↳consistent naming

    # Split clients into regions (Edge forwarders)
    regions = [clients[i:i + 50] for i in range(0, len(clients), 50)]
    edge_results = {}

    start_time = time.time()

    # Optimize Edge layer
    edge_start_time = time.time()
    for region_id, region_clients in enumerate(regions):
        model = VideoOptimizationModel(f"Edge_Forwarder_{region_id}",
↳region_clients, problem_data)
        model.optimize()
        edge_results[f"Edge_Forwarder_{region_id}"] = model.get_results()
        edge_time = time.time() - edge_start_time

    # Optimize Core layer
    core_results = {}
    core_start_time = time.time()
    num_core = max(len(regions) // 10, 1)
    for core_id in range(num_core):
        print(f"Optimizing Core Forwarder {core_id + 1}...")
        # Determine which edges are managed by this core
        managed_edges = [f"Edge_Forwarder_{i}" for i in range(len(regions))
↳if i % num_core == core_id]
        # Extract traffic demands from edge_results
        lower_layer_results = {}
        for edge in managed_edges:
            results = edge_results.get(edge, {})
            for (client, res), value in results.items():
                if value > 0.5:
                    lower_layer_results[(edge, res)] = value

        # Calculate weights_core and max_bitrate_core based on
↳lower_layer_results
        weights_core = {}
        max_bitrate_core = {}
        for (edge, res), value in lower_layer_results.items():
            weights_core[edge] = weights_core.get(edge, 0) + q[res]
            max_bitrate_core[edge] = max_bitrate_core.get(edge, 0) + B[res]

        # Create problem data for core layer
        problem_data_core = (B, q, max_bitrate_core, links, weights_core)

        # Populate core_lower_layer with (edge, resolution): value

```

```

        core_lower_layer = lower_layer_results.copy()

        # Optimize core forwarder
        model = VideoOptimizationModel(
            forwarder_name=f"Core_Forwarder_{core_id}",
            clients=managed_edges,
            problem_data=problem_data_core,
            lower_layer_results=core_lower_layer
        )
        model.optimize()
        core_results[f"Core_Forwarder_{core_id}"] = model.get_results()
        core_time = time.time() - core_start_time

    total_time = time.time() - start_time

    print(f"\nEdge layer optimization took {edge_time:.3f} seconds.")
    print(f"Core layer optimization took {core_time:.3f} seconds.")
    print(f"Total optimization time for {num_users} users: {total_time:.3f}↵
↵seconds.\n")

    return {
        'num_users': num_users,
        'total_time': total_time,
        'edge_time': edge_time,
        'core_time': core_time,
        'status': 'Success',
    }

def run_analysis(self, user_scales):
    summary = []
    for num_users in user_scales:
        print(f"\nAnalyzing scale: {num_users} users")
        result = self.analyze_scale(num_users)
        self.results.append(result)
        summary.append(result)

    self.plot_results(summary)

def plot_results(self, summary):
    df = pd.DataFrame(summary)

    # Plot edge time, core time, and total time
    plt.figure(figsize=(10, 6))
    plt.plot(df['num_users'], df['edge_time'], label='Edge Layer Time',↵
↵marker='o', color='blue')
    plt.plot(df['num_users'], df['core_time'], label='Core Layer Time',↵
↵marker='^', color='green')

```

```

plt.plot(df['num_users'], df['total_time'], label='Total Time',
↪marker='s', color='orange')
plt.title('Optimization Time by Number of Users')
plt.xlabel('Number of Users')
plt.ylabel('Time (seconds)')
plt.grid(True)
plt.legend()
plt.show()

print("\nPerformance Summary:")
print(df)

def main():
    analyzer = DistributedPerformanceAnalyzer()
    user_scales = [100, 200, 300, 400]
    analyzer.run_analysis(user_scales)

if __name__ == "__main__":
    main()

```

Analyzing scale: 100 users  
Optimizing Core Forwarder 1...

Edge layer optimization took 0.136 seconds.  
Core layer optimization took 0.019 seconds.  
Total optimization time for 100 users: 0.155 seconds.

Analyzing scale: 200 users  
Optimizing Core Forwarder 1...

Edge layer optimization took 0.559 seconds.  
Core layer optimization took 0.042 seconds.  
Total optimization time for 200 users: 0.601 seconds.

Analyzing scale: 300 users  
Optimizing Core Forwarder 1...

Edge layer optimization took 1.424 seconds.  
Core layer optimization took 0.070 seconds.  
Total optimization time for 300 users: 1.495 seconds.

Analyzing scale: 400 users

```

-----
GurobiError                                Traceback (most recent call last)
Cell In[2], line 236
    233 analyzer.run_analysis(user_scales)
    235 if __name__ == "__main__":
--> 236     main()

Cell In[2], line 233, in main()
    231 analyzer = DistributedPerformanceAnalyzer()
    232 user_scales = [100, 200, 300, 400]
--> 233 analyzer.run_analysis(user_scales)

Cell In[2], line 206, in DistributedPerformanceAnalyzer.run_analysis(self,
↳ user_scales)
    204 for num_users in user_scales:
    205     print(f"\nAnalyzing scale: {num_users} users")
--> 206     result = self.analyze_scale(num_users)
    207     self.results.append(result)
    208     summary.append(result)

Cell In[2], line 144, in DistributedPerformanceAnalyzer.analyze_scale(self,
↳ num_users)
    142 for region_id, region_clients in enumerate(regions):
    143     model = VideoOptimizationModel(f"Edge_Forwarder_{region_id}",
↳ region_clients, problem_data)
--> 144     model.optimize()
    145     edge_results[f"Edge_Forwarder_{region_id}"] = model.get_results()
    146 edge_time = time.time() - edge_start_time

Cell In[2], line 115, in VideoOptimizationModel.optimize(self)
    113 self.set_objective()
    114 self.add_constraints()
--> 115 self.model.optimize()

File src/gurobipy/_model.pyx:901, in gurobipy._model.Model.optimize()

GurobiError: Model too large for size-limited license; visit https://gurobi.com
↳ unrestricted for more information

```

```
[24]: !jupyter nbconvert --to pdf model.ipynb
```

```
[NbConvertApp] Converting notebook model.ipynb to pdf
```

```
[NbConvertApp] Writing 1991535 bytes to model.pdf
```

## 5 Scalability Analysis (for centralized model)

```
[22]: import gurobipy as gp
from gurobipy import GRB
import pandas as pd
import matplotlib.pyplot as plt
import time
import numpy as np
from typing import Dict, List, Tuple

class ScalabilityAnalysis:
    def __init__(self):
        # Base video resolution and quality parameters
        self.B = {
            '8K': 200,
            '4K': 45,
            '2K': 16,
            '1K': 8
        }
        self.q = {
            '8K': 4,
            '4K': 3,
            '2K': 2,
            '1K': 1
        }

    def generate_network(self, num_users: int) -> Tuple[Dict, Dict, Dict]:
        """
        Generate network topology based on number of users
        Returns: max_bitrate, links, paths
        """
        # Calculate required number of routers (each router supports 8 users)
        num_routers = max(1, (num_users + 7) // 8)

        # Generate user maximum bitrates (randomly assign different user
        ↪capabilities)
        max_bitrate = {}
        bitrate_options = [10, 50, 200]
        for i in range(num_users):
            client_id = chr(65 + i) if i < 26 else f'User_{i}'
            max_bitrate[client_id] = np.random.choice(bitrate_options)

        # Generate link capacities
        links = {}
        # Server to core router link
        core_capacity = num_users * 50 # Estimate required capacity
        links[('Server', 'Router_Core')] = core_capacity
```

```

# Core to access router links
for i in range(num_routers):
    router_id = f'Router_{i}'
    links[('Router_Core', router_id)] = core_capacity // num_routers

# Access router to client links
start_user = i * 8
end_user = min((i + 1) * 8, num_users)
for j in range(start_user, end_user):
    client_id = chr(65 + j) if j < 26 else f'User_{j}'
    links[(router_id, f'Client_{client_id}')] = 200

# Generate paths
paths = {}
for i in range(num_users):
    client_id = chr(65 + i) if i < 26 else f'User_{i}'
    router_id = f'Router_{i // 8}'
    paths[client_id] = [
        ('Server', 'Router_Core'),
        ('Router_Core', router_id),
        (router_id, f'Client_{client_id}')]
]

return max_bitrate, links, paths

def plot_network_topology(self, links: Dict, client_resolutions: Dict = None):
    """Plot network topology with optional client resolution information"""
    fig, ax = plt.subplots(figsize=(15, 10))

    # Create directed graph for visualization
    unique_nodes = set()
    for (src, dst) in links.keys():
        unique_nodes.add(src)
        unique_nodes.add(dst)

    # Calculate layout
    pos = {}
    servers = [n for n in unique_nodes if 'Server' in n]
    core_routers = [n for n in unique_nodes if 'Router_Core' in n]
    access_routers = [n for n in unique_nodes if 'Router_' in n and 'Core'
    not in n]
    clients = [n for n in unique_nodes if 'Client_' in n]

    # Position calculations
    y_levels = {'Server': 0.9, 'Core': 0.7, 'Access': 0.5, 'Client': 0.2}

```



```

# Position servers
for i, node in enumerate(servers):
    pos[node] = (0.5, y_levels['Server'])

# Position core routers
for i, node in enumerate(core_routers):
    pos[node] = (0.5, y_levels['Core'])

# Position access routers
router_spacing = 1.0 / (len(access_routers) + 1)
for i, node in enumerate(access_routers):
    pos[node] = ((i + 1) * router_spacing, y_levels['Access'])

# Position clients
clients_per_router = len(clients) / len(access_routers)
current_client = 0
for i, router in enumerate(access_routers):
    router_clients = clients[int(i * clients_per_router):int((i + 1) *
↪clients_per_router)]
    client_spacing = router_spacing / (len(router_clients) + 1)
    router_x = pos[router][0]
    start_x = router_x - router_spacing/2
    for j, client in enumerate(router_clients):
        pos[client] = (start_x + (j + 1) * client_spacing,
↪y_levels['Client'])

# Draw nodes
node_colors = {'Server': 'red', 'Router': 'blue', 'Client': 'green'}
for node in unique_nodes:
    x, y = pos[node]
    if 'Server' in node:
        color = node_colors['Server']
        marker = 's'
    elif 'Router' in node:
        color = node_colors['Router']
        marker = 'o'
    else:
        color = node_colors['Client']
        marker = '^'
    ax.scatter(x, y, c=color, marker=marker, s=200)

# Add node labels
if client_resolutions and 'Client_' in node:
    client_id = node.split('_')[1]
    resolution = client_resolutions.get(client_id, 'N/A')

```

```

        ax.text(x, y-0.05, f'{node}\n{resolution}', ha='center',
↪va='top')
    else:
        ax.text(x, y-0.05, node, ha='center', va='top')

    # Draw edges
    for (src, dst), capacity in links.items():
        x1, y1 = pos[src]
        x2, y2 = pos[dst]
        ax.plot([x1, x2], [y1, y2], 'k-', alpha=0.5)
        # Add capacity labels
        mid_x = (x1 + x2) / 2
        mid_y = (y1 + y2) / 2
        ax.text(mid_x, mid_y, f'{capacity}Mbps', ha='center', va='center',
↪bbox=dict(facecolor='white', alpha=0.7))

    # Add legend
    legend_elements = [
        plt.Line2D([0], [0], marker='s', color='w', label='Server',
↪markerfacecolor='red', markersize=10),
        plt.Line2D([0], [0], marker='o', color='w', label='Router',
↪markerfacecolor='blue', markersize=10),
        plt.Line2D([0], [0], marker='^', color='w', label='Client',
↪markerfacecolor='green', markersize=10)
    ]
    ax.legend(handles=legend_elements, loc='upper right')

    plt.title('Network Topology')
    plt.grid(True, linestyle='--', alpha=0.3)
    plt.tight_layout()
    plt.show()

def create_and_solve_model(self, num_users: int) -> Dict:
    """
    Create and solve optimization model for specific scale
    Returns solution statistics
    """
    try:
        # Generate network topology
        max_bitrate, links, paths = self.generate_network(num_users)
        clients = list(max_bitrate.keys())
        resolutions = list(self.B.keys())

        # Create model
        start_time = time.time()
        model = gp.Model("video_quality_optimization")
        model.setParam('OutputFlag', 0)

```

```

        # Create variables
        x = model.addVars(clients, resolutions, vtype=GRB.BINARY, name="x")
        m = model.addVars(links.keys(), resolutions, vtype=GRB.BINARY,
↪name="m")
        b = model.addVars(links.keys(), vtype=GRB.CONTINUOUS, name="b")
        x_slack = model.addVars(clients, vtype=GRB.CONTINUOUS,
↪name="x_slack")

        # Set objective function
        alpha, gamma, lambda_val = 1.0, 0.5, 1000
        obj = -alpha * gp.quicksum(self.q[j] * x[i,j] for i in clients for
↪j in resolutions)
        obj += -gamma * gp.quicksum(x[i,j] for i in clients for j in
↪resolutions)
        obj += lambda_val * gp.quicksum(x_slack[i] for i in clients)
        model.setObjective(obj, GRB.MINIMIZE)

        # Add constraints
        self._add_constraints(model, x, m, b, x_slack, clients,
↪resolutions, links, paths, max_bitrate)

        # Solve and collect statistics
        setup_time = time.time() - start_time
        model.optimize()
        solve_time = model.Runtime

        # Plot network topology
        if model.status == GRB.OPTIMAL:
            client_resolutions = {}
            for i in clients:
                for j in resolutions:
                    if x[i,j].x > 0.5:
                        client_resolutions[i] = j
            self.plot_network_topology(links, client_resolutions)

        return {
            'num_users': num_users,
            'num_variables': model.NumVars,
            'num_constraints': model.NumConstrs,
            'setup_time': setup_time,
            'solve_time': solve_time,
            'total_time': setup_time + solve_time,
            'objective_value': model.ObjVal if model.status == GRB.OPTIMAL
↪else None,
            'model_status': model.status,

```

```

        'status': 'Success'
    }

except Exception as e:
    return {
        'num_users': num_users,
        'num_variables': 0,
        'num_constraints': 0,
        'setup_time': 0,
        'solve_time': 0,
        'total_time': 0,
        'objective_value': None,
        'model_status': None,
        'status': 'Failed',
        'error': str(e)
    }

def _add_constraints(self, model, x, m, b, x_slack, clients, resolutions,
    ↪links, paths, max_bitrate):
    """Add all model constraints"""
    # User assignment constraint
    for i in clients:
        model.addConstr(gp.quicksum(x[i,j] for j in resolutions) +
    ↪x_slack[i] == 1)

    # User capability constraint
    for i in clients:
        for j in resolutions:
            if self.B[j] > max_bitrate[i]:
                model.addConstr(x[i,j] == 0)

    # Link bandwidth constraints
    for l, k in links:
        model.addConstr(b[l,k] == gp.quicksum(m[l,k,j] * self.B[j] for j in
    ↪resolutions))
        model.addConstr(b[l,k] <= links[l,k])

    # Multicast logic constraint
    for i in clients:
        for j in resolutions:
            for l,k in paths[i]:
                model.addConstr(m[l,k,j] >= x[i,j])

def run_scalability_analysis(self, user_scales: List[int]) -> pd.DataFrame:
    """
    Run performance analysis for different scales
    """

```

```

results = []
for num_users in user_scales:
    print(f"\nAnalyzing user count: {num_users}")
    stats = self.create_and_solve_model(num_users)
    results.append(stats)
    print(f"Solution status: {stats['status']}")
    if stats['status'] == 'Success':
        print(f"Completed - Total time: {stats['total_time']:.2f}s")
    else:
        print(f"Failed - Error: {stats.get('error', 'Unknown error')}")

return pd.DataFrame(results)

def plot_scalability_results(self, results: pd.DataFrame):
    """
    Plot performance analysis results
    """
    success_results = results[results['status'] == 'Success'].copy()

    if len(success_results) == 0:
        print("No successful solutions to plot")
        return

    fig, ((ax1, ax2), (ax3, ax4)) = plt.subplots(2, 2, figsize=(15, 12))

    # Variables growth
    ax1.plot(success_results['num_users'],
    ↪ success_results['num_variables'], 'bo-')
    ax1.set_title('Variables Growth')
    ax1.set_xlabel('Number of Users')
    ax1.set_ylabel('Number of Variables')
    ax1.grid(True)

    # Constraints growth
    ax2.plot(success_results['num_users'],
    ↪ success_results['num_constraints'], 'ro-')
    ax2.set_title('Constraints Growth')
    ax2.set_xlabel('Number of Users')
    ax2.set_ylabel('Number of Constraints')
    ax2.grid(True)

    # Solution time growth
    ax3.plot(success_results['num_users'], success_results['solve_time'],
    ↪ 'go-')
    ax3.set_title('Solution Time Growth')
    ax3.set_xlabel('Number of Users')
    ax3.set_ylabel('Solution Time (seconds)')

```

```

        ax3.grid(True)

        # Total time growth
        ax4.plot(success_results['num_users'], success_results['total_time'],
↪ 'mo-')
        ax4.set_title('Total Time Growth')
        ax4.set_xlabel('Number of Users')
        ax4.set_ylabel('Total Time (seconds)')
        ax4.grid(True)

        plt.tight_layout()
        plt.show()

def main():
    # Create analysis instance
    analyzer = ScalabilityAnalysis()

    # Define test scales
    user_scales = [8, 16, 32, 64, 128]

    # Run analysis
    results = analyzer.run_scalability_analysis(user_scales)

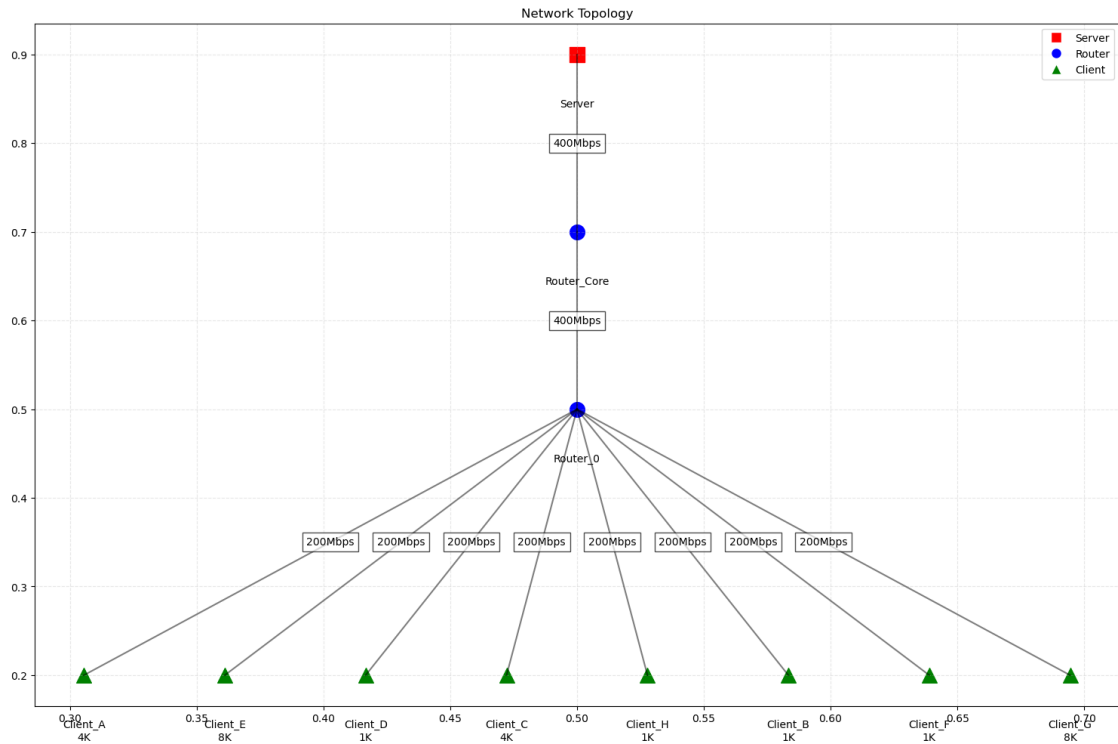
    # Display results table
    print("\nPerformance Analysis Results:")
    print(results.to_string())

    # Plot results
    analyzer.plot_scalability_results(results)

if __name__ == "__main__":
    main()

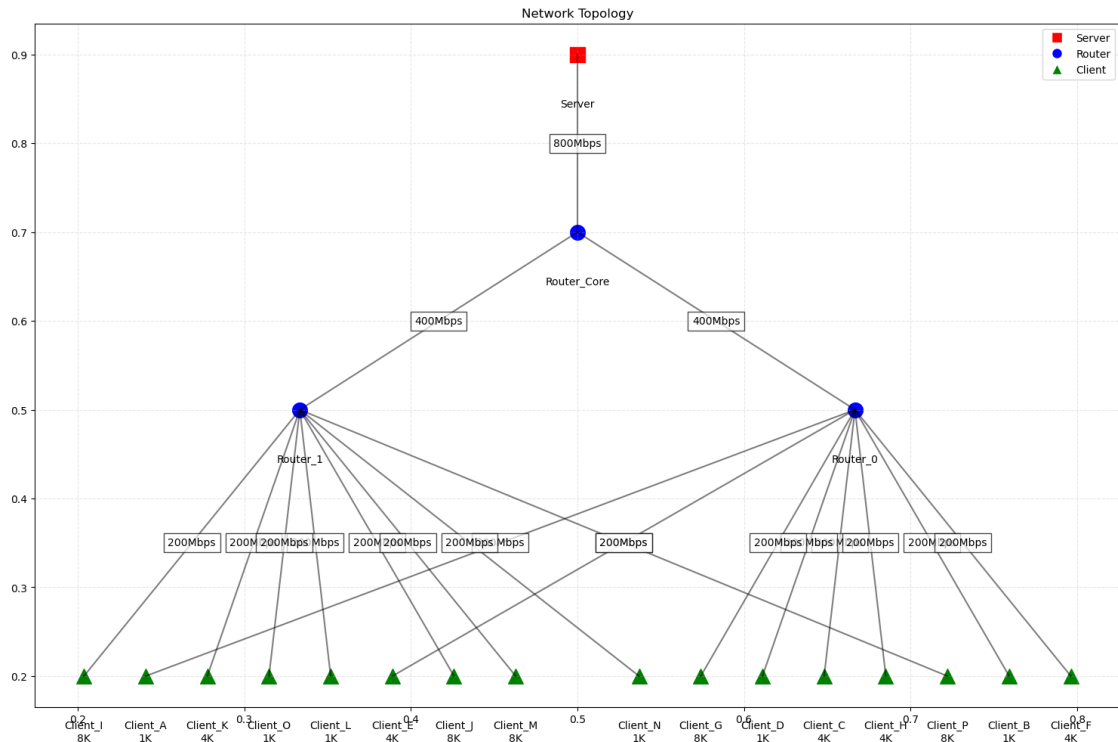
```

Analyzing user count: 8



Solution status: Success  
Completed - Total time: 0.00s

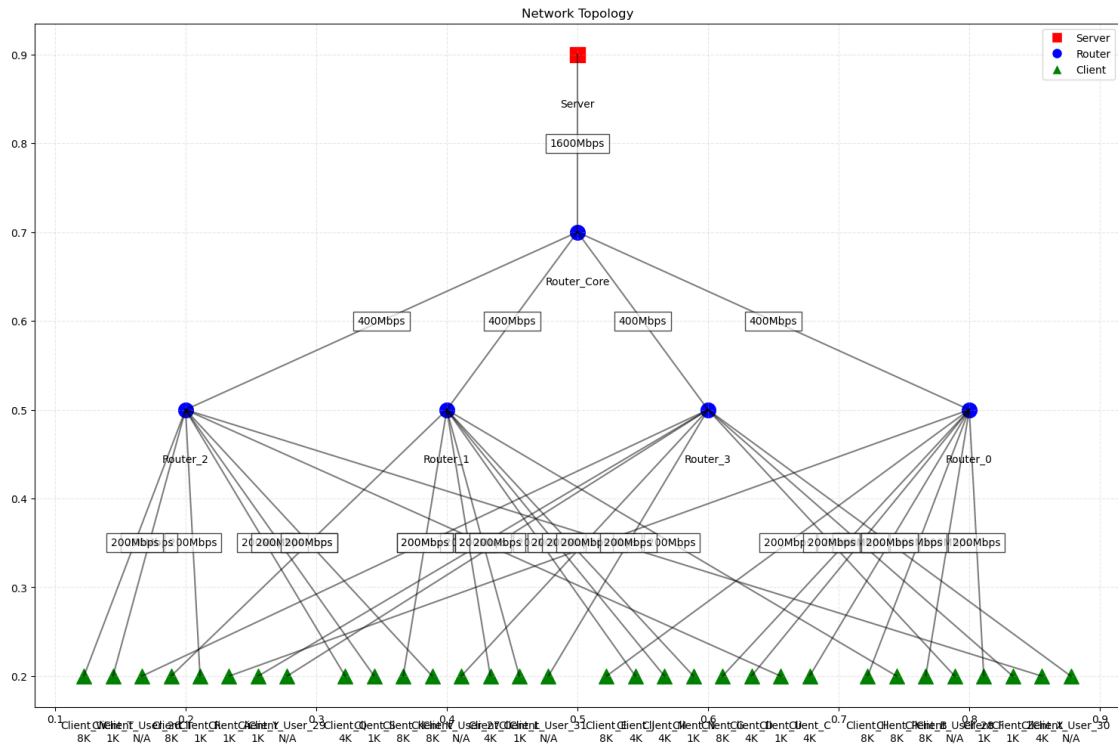
Analyzing user count: 16



Solution status: Success  
 Completed - Total time: 0.00s

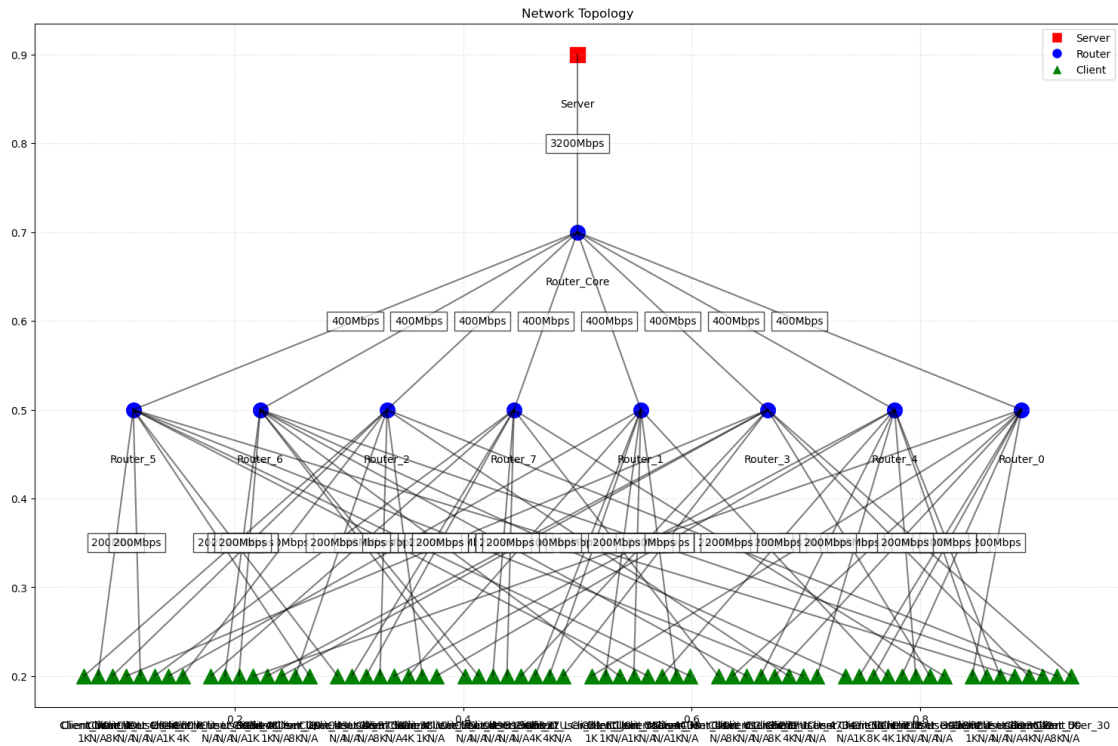
Analyzing user count: 32





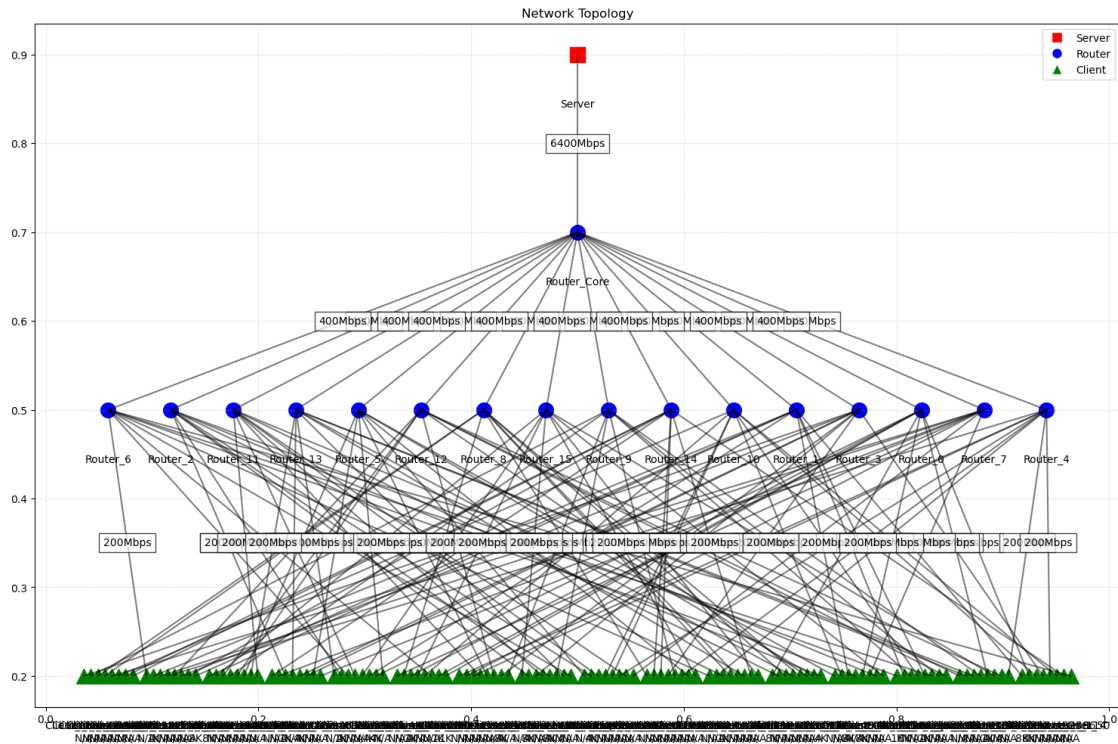
Solution status: Success  
 Completed - Total time: 0.01s

Analyzing user count: 64



Solution status: Success  
 Completed - Total time: 0.01s

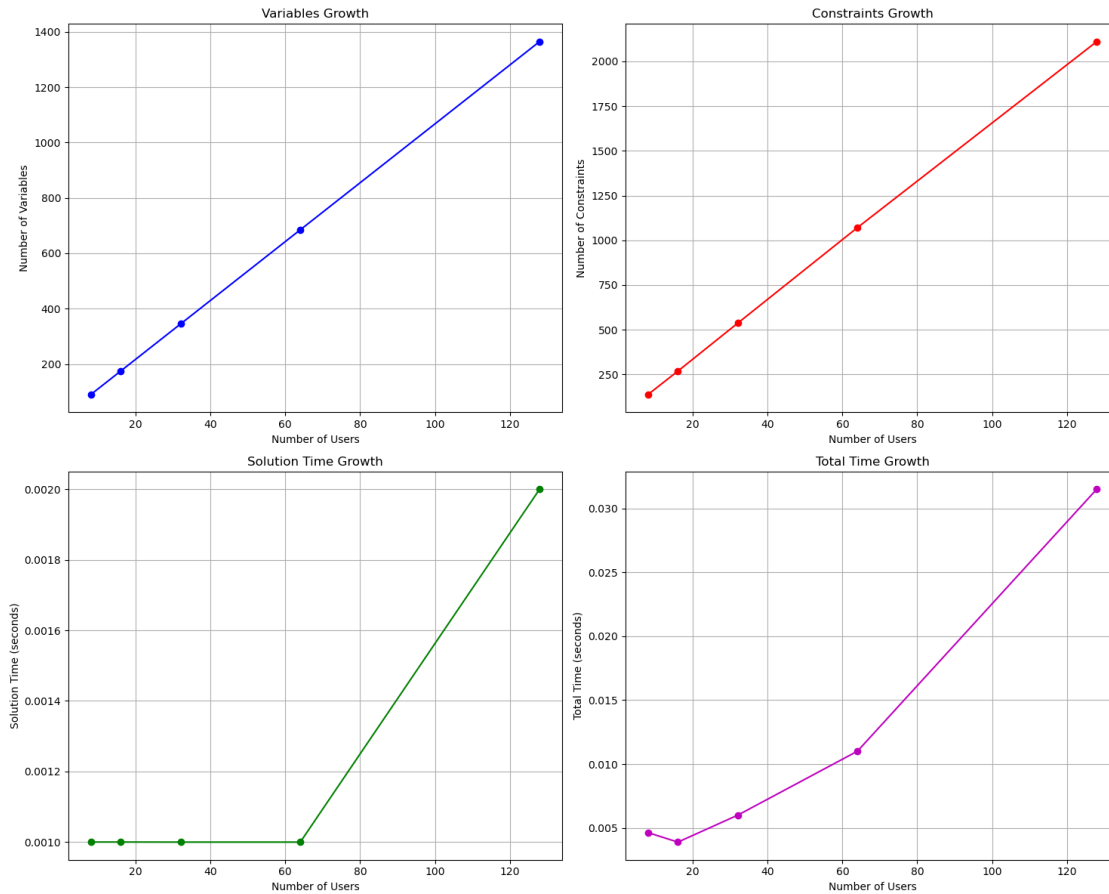
Analyzing user count: 128



Solution status: Success  
Completed - Total time: 0.03s

#### Performance Analysis Results:

	num_users	num_variables	num_constraints	setup_time	solve_time	total_time
0	8	90	138	0.003630	0.001	0.004630
-22.0		2	Success			
1	16	175	269	0.002888	0.001	0.003888
-49.0		2	Success			
2	32	345	537	0.004997	0.001	0.005997
-97.0		2	Success			
3	64	685	1071	0.010005	0.001	0.011005
-195.0		2	Success			
4	128	1365	2111	0.029518	0.002	0.031518
-419.0		2	Success			



## 6 Directly Using Optimization Algorithms (no gorubi)

```
[6]: import random
import math
import copy
import time
from typing import Dict, List, Tuple
import numpy as np

def measure_execution_time(func):
    """Decorator to measure execution time of methods"""
    def wrapper(*args, **kwargs):
        start_time = time.time()
        result = func(*args, **kwargs)
        end_time = time.time()
        execution_time = end_time - start_time
        print(f"Execution time of {func.__name__}: {execution_time:.4f}␣
↪seconds")
```

```

        return result
    return wrapper

class BaseOptimization:
    def __init__(self):
        self.B, self.q, self.max_bitrate, self.links, self.paths = self.
↪define_problem_data()
        self.clients = list(self.max_bitrate.keys())
        self.resolutions = list(self.B.keys())

    def define_problem_data(self):
        # Resolution bandwidth requirements (Mbps)
        B = {
            '8K': 200,
            '4K': 45,
            '2K': 16,
            '1K': 8
        }

        # Quality scores
        q = {
            '8K': 4,
            '4K': 3,
            '2K': 2,
            '1K': 1
        }

        # Client maximum bitrates
        max_bitrate = {
            'A': 10,
            'B': 50,
            'C': 200,
            'D': 200,
            'E': 20,
            'F': 50,
            'G': 200,
            'H': 200
        }

        # Network link capacities
        links = {
            ('Server', 'Router_C'): 300,
            ('Router_C', 'Router_A'): 250,
            ('Router_C', 'Router_B'): 250,
            ('Router_A', 'Client_A'): 200,
            ('Router_A', 'Client_B'): 200,
            ('Router_A', 'Client_C'): 200,

```

```

        ('Router_A', 'Client_D'): 200,
        ('Router_B', 'Client_E'): 200,
        ('Router_B', 'Client_F'): 200,
        ('Router_B', 'Client_G'): 200,
        ('Router_B', 'Client_H'): 200
    }

    # Network paths for each client
    paths = {
        'A': [('Server', 'Router_C'), ('Router_C', 'Router_A'),
        ↪ ('Router_A', 'Client_A')],
        'B': [('Server', 'Router_C'), ('Router_C', 'Router_A'),
        ↪ ('Router_A', 'Client_B')],
        'C': [('Server', 'Router_C'), ('Router_C', 'Router_A'),
        ↪ ('Router_A', 'Client_C')],
        'D': [('Server', 'Router_C'), ('Router_C', 'Router_A'),
        ↪ ('Router_A', 'Client_D')],
        'E': [('Server', 'Router_C'), ('Router_C', 'Router_B'),
        ↪ ('Router_B', 'Client_E')],
        'F': [('Server', 'Router_C'), ('Router_C', 'Router_B'),
        ↪ ('Router_B', 'Client_F')],
        'G': [('Server', 'Router_C'), ('Router_C', 'Router_B'),
        ↪ ('Router_B', 'Client_G')],
        'H': [('Server', 'Router_C'), ('Router_C', 'Router_B'),
        ↪ ('Router_B', 'Client_H')]
    }

    return B, q, max_bitrate, links, paths

def check_capacity_constraints(self, solution: Dict[str, str]) -> bool:
    """Check link capacity constraints"""
    link_usage = {link: 0 for link in self.links.keys()}

    for client, resolution in solution.items():
        bandwidth = self.B[resolution]
        for link in self.paths[client]:
            link_usage[link] += bandwidth

        if link_usage[link] > self.links[link]:
            return False
    return True

def calculate_objective(self, solution: Dict[str, str], alpha=1.0, gamma=0.
↪5) -> float:
    """Calculate objective function value"""
    quality_sum = sum(self.q[res] for res in solution.values())

```

```

        resolution_count = len(solution)

        return -alpha * quality_sum - gamma * resolution_count

class GreedyOptimization(BaseOptimization):
    @measure_execution_time
    def optimize(self) -> Dict[str, str]:
        """
        Greedy algorithm implementation
        Time Complexity Analysis:
        - Sorting clients:  $O(C \log C)$ , where  $C$  is number of clients
        - For each client, we try each resolution:  $O(C * R)$ , where  $R$  is number of
        ↪ of resolutions
        - Checking constraints for each attempt:  $O(L)$ , where  $L$  is number of
        ↪ links

        Total Time Complexity:  $O(C * R * L)$ 
        """
        print("Running Greedy Algorithm:")
        print(f"Number of clients: {len(self.clients)}")
        print(f"Number of resolutions: {len(self.resolutions)}")
        print(f"Number of links: {len(self.links)}")

        solution = {}
        operations_count = 0 # Counter for actual operations

        sorted_clients = sorted(self.clients,
                                key=lambda x: self.max_bitrate[x],
                                reverse=True)

        for client in sorted_clients:
            for resolution in sorted(self.resolutions,
                                    key=lambda x: self.q[x],
                                    reverse=True):
                operations_count += 1
                if self.B[resolution] <= self.max_bitrate[client]:
                    solution[client] = resolution
                    if self.check_capacity_constraints(solution):
                        continue
                    else:
                        solution.pop(client)

            if client not in solution:
                solution[client] = '1K'

        print(f"Total operations performed: {operations_count}")
        return solution

```

```

class SimulatedAnnealing(BaseOptimization):
    def get_neighbor(self, solution: Dict[str, str]) -> Dict[str, str]:
        """
        Generate a neighbor solution
        Time Complexity:  $O(1)$  - Constant time operation
        """
        new_solution = solution.copy()
        client = random.choice(list(solution.keys()))
        current_res = solution[client]

        # Get available resolution options
        available_res = [res for res in self.resolutions
                        if self.B[res] <= self.max_bitrate[client]
                        and res != current_res]

        if available_res:
            new_solution[client] = random.choice(available_res)

        return new_solution

    @measure_execution_time
    def optimize(self, initial_temp=100.0, cooling_rate=0.95,
                iterations=1000) -> Dict[str, str]:
        """
        Simulated Annealing implementation
        Time Complexity Analysis:
        - Initial solution (Greedy):  $O(C * R * L)$ 
        - For each iteration:
            - Generate neighbor:  $O(1)$ 
            - Check constraints:  $O(L)$ 
        Total Time Complexity:  $O(C * R * L + I * L)$ , where  $I$  is number of
        iterations
        """
        print("Running Simulated Annealing:")
        print(f"Number of iterations: {iterations}")
        print(f"Initial temperature: {initial_temp}")
        print(f"Cooling rate: {cooling_rate}")

        operations_count = 0
        current_solution = GreedyOptimization().optimize()
        best_solution = current_solution.copy()
        current_cost = self.calculate_objective(current_solution)
        best_cost = current_cost

        temperature = initial_temp

        for i in range(iterations):

```



```

        operations_count += 1
        neighbor = self.get_neighbor(current_solution)

        if self.check_capacity_constraints(neighbor):
            neighbor_cost = self.calculate_objective(neighbor)
            cost_diff = neighbor_cost - current_cost

            if (cost_diff < 0 or
                random.random() < math.exp(-cost_diff / temperature)):
                current_solution = neighbor
                current_cost = neighbor_cost

            if current_cost < best_cost:
                best_solution = current_solution.copy()
                best_cost = current_cost

        temperature *= cooling_rate

        if i % 100 == 0: # Progress tracking
            print(f"Iteration {i}, Temperature: {temperature:.2f}, Best_
↪cost: {best_cost:.2f}")

        print(f"Total operations performed: {operations_count}")
        return best_solution

class GeneticAlgorithm(BaseOptimization):
    def create_individual(self) -> Dict[str, str]:
        """Create an individual (solution)"""
        individual = {}
        for client in self.clients:
            available_res = [res for res in self.resolutions
                             if self.B[res] <= self.max_bitrate[client]]
            individual[client] = random.choice(available_res)
        return individual

    def crossover(self, parent1: Dict[str, str],
                  parent2: Dict[str, str]) -> Dict[str, str]:
        """Perform crossover operation"""
        child = {}
        for client in self.clients:
            if random.random() < 0.5:
                child[client] = parent1[client]
            else:
                child[client] = parent2[client]
        return child

    def mutate(self, individual: Dict[str, str],

```

```

        mutation_rate: float = 0.1) -> Dict[str, str]:
"""Perform mutation operation"""
    mutated = individual.copy()
    for client in self.clients:
        if random.random() < mutation_rate:
            available_res = [res for res in self.resolutions
                             if self.B[res] <= self.max_bitrate[client]]
            mutated[client] = random.choice(available_res)
    return mutated

@measure_execution_time
def optimize(self, population_size=50, generations=100) -> Dict[str, str]:
    """
Genetic Algorithm implementation
Time Complexity Analysis:
- Population initialization:  $O(P * C)$ , where  $P$  is population size
- For each generation:
    - Fitness calculation:  $O(P * L)$ 
    - Sorting:  $O(P \log P)$ 
    - Creating new population:  $O(P * C)$ 
Total Time Complexity:  $O(G * P * (L + \log P + C))$ , where  $G$  is number of
    ↪ generations
    """

    print("Running Genetic Algorithm:")
    print(f"Population size: {population_size}")
    print(f"Number of generations: {generations}")

    operations_count = 0
    population = []
    for _ in range(population_size):
        individual = self.create_individual()
        if self.check_capacity_constraints(individual):
            population.append(individual)

    best_solution = None
    best_fitness = float('inf')

    for gen in range(generations):
        operations_count += population_size

        fitness_scores = [(self.calculate_objective(ind), ind)
                          for ind in population]
        fitness_scores.sort(key=lambda x: x[0])

        if fitness_scores[0][0] < best_fitness:
            best_fitness = fitness_scores[0][0]
            best_solution = fitness_scores[0][1].copy()

```

```

        new_population = [ind for _, ind in fitness_scores[:2]]

        while len(new_population) < population_size:
            parent_candidates = fitness_scores[:10]
            parent1 = random.choice(parent_candidates)[1]
            parent2 = random.choice(parent_candidates)[1]

            child = self.crossover(parent1, parent2)
            child = self.mutate(child)

            if self.check_capacity_constraints(child):
                new_population.append(child)

        population = new_population

        if gen % 10 == 0: # Progress tracking
            print(f"Generation {gen}, Best fitness: {best_fitness:.2f}")

        print(f"Total operations performed: {operations_count}")
        return best_solution

def compare_methods():
    """Compare the results and performance of different methods"""
    methods = [
        ("Greedy", GreedyOptimization()),
        ("Simulated Annealing", SimulatedAnnealing()),
        ("Genetic Algorithm", GeneticAlgorithm())
    ]

    results = {}
    print("\nPerformance Comparison:")
    print("-" * 50)

    for name, method in methods:
        print(f"\nExecuting {name} Algorithm...")
        start_time = time.time()
        solution = method.optimize()
        end_time = time.time()
        execution_time = end_time - start_time

        objective_value = method.calculate_objective(solution)
        results[name] = {
            'solution': solution,
            'objective_value': objective_value,
            'execution_time': execution_time
        }

```

```

        print(f"\n{name} Results:")
        print(f"Objective Value: {objective_value}")
        print(f"Execution Time: {execution_time:.4f} seconds")
        print("Client Assignments:")
        for client, resolution in solution.items():
            print(f"Client {client}: {resolution}")

    return results

if __name__ == "__main__":
    results = compare_methods()

    # Print comparative summary
    print("\nAlgorithm Comparison Summary:")
    print("-" * 50)
    for name, data in results.items():
        print(f"\n{name}:")
        print(f"Objective Value: {data['objective_value']:.2f}")
        print(f"Execution Time: {data['execution_time']:.4f} seconds")

```

Performance Comparison:

-----

Executing Greedy Algorithm...

Running Greedy Algorithm:

Number of clients: 8

Number of resolutions: 4

Number of links: 11

Total operations performed: 32

Execution time of optimize: 0.0012 seconds

Greedy Results:

Objective Value: -12.0

Execution Time: 0.0012 seconds

Client Assignments:

Client C: 1K

Client D: 1K

Client G: 1K

Client H: 1K

Client B: 1K

Client F: 1K

Client E: 1K

Client A: 1K

Executing Simulated Annealing Algorithm...

Running Simulated Annealing:

Number of iterations: 1000

Initial temperature: 100.0

Cooling rate: 0.95

Running Greedy Algorithm:

Number of clients: 8

Number of resolutions: 4

Number of links: 11

Total operations performed: 32

Execution time of optimize: 0.0000 seconds

Iteration 0, Temperature: 95.00, Best cost: -13.00

Iteration 100, Temperature: 0.56, Best cost: -23.00

Iteration 200, Temperature: 0.00, Best cost: -25.00

Iteration 300, Temperature: 0.00, Best cost: -25.00

Iteration 400, Temperature: 0.00, Best cost: -25.00

Iteration 500, Temperature: 0.00, Best cost: -25.00

Iteration 600, Temperature: 0.00, Best cost: -25.00

Iteration 700, Temperature: 0.00, Best cost: -25.00

Iteration 800, Temperature: 0.00, Best cost: -25.00

Iteration 900, Temperature: 0.00, Best cost: -25.00

Total operations performed: 1000

Execution time of optimize: 0.0040 seconds

Simulated Annealing Results:

Objective Value: -25.0

Execution Time: 0.0040 seconds

Client Assignments:

Client C: 4K

Client D: 4K

Client G: 4K

Client H: 4K

Client B: 4K

Client F: 4K

Client E: 2K

Client A: 1K

Executing Genetic Algorithm Algorithm...

Running Genetic Algorithm:

Population size: 50

Number of generations: 100

Generation 0, Best fitness: -23.00

Generation 10, Best fitness: -25.00

Generation 20, Best fitness: -25.00

Generation 30, Best fitness: -25.00

Generation 40, Best fitness: -25.00

Generation 50, Best fitness: -25.00

Generation 60, Best fitness: -25.00

Generation 70, Best fitness: -25.00

Generation 80, Best fitness: -25.00  
Generation 90, Best fitness: -25.00  
Total operations performed: 5000  
Execution time of optimize: 0.0302 seconds

Genetic Algorithm Results:  
Objective Value: -25.0  
Execution Time: 0.0302 seconds  
Client Assignments:  
Client A: 1K  
Client B: 4K  
Client C: 4K  
Client D: 4K  
Client E: 2K  
Client F: 4K  
Client G: 4K  
Client H: 4K

Algorithm Comparison Summary:

-----

Greedy:  
Objective Value: -12.00  
Execution Time: 0.0012 seconds

Simulated Annealing:  
Objective Value: -25.00  
Execution Time: 0.0040 seconds

Genetic Algorithm:  
Objective Value: -25.00  
Execution Time: 0.0302 seconds

## 6.1 Analysis of Optimization Algorithms

### 6.1.1 1. Greedy Algorithm

- **Approach:** Makes locally optimal choices at each step
- **Implementation Logic:**
  - Sorts clients by maximum bandwidth (highest to lowest)
  - For each client, tries to assign highest quality resolution that satisfies constraints
  - Falls back to lowest resolution (1K) if no feasible solution found
- **Advantages:** Fast, simple, deterministic
- **Disadvantages:** May get stuck in local optima

### 6.1.2 2. Simulated Annealing

- **Approach:** Probabilistic technique that simulates physical annealing process
- **Implementation Logic:**

- Starts with greedy solution
- Iteratively generates neighbor solutions by randomly changing one client's resolution
- Accepts improvements always, accepts worse solutions with decreasing probability
- Uses temperature parameter to control acceptance of worse solutions
- **Advantages:** Can escape local optima, good balance of exploration and exploitation
- **Disadvantages:** Results may vary between runs, requires parameter tuning

### 6.1.3 3. Genetic Algorithm

- **Implementation Logic:**
  - Population-based approach with selection, crossover, and mutation
  - Uses elitism to preserve best solutions
  - Crossover randomly selects resolution assignments from parents
  - Mutation randomly changes resolutions with low probability
  - Maintains feasibility through constraint checking
- **Advantages:** Can explore multiple solution paths simultaneously, good for complex search spaces
- **Disadvantages:** Computationally intensive, requires careful parameter tuning

## 6.2 Comparison of Approaches

### 6.2.1 Performance Characteristics

- **Greedy:** Generally provides good solutions quickly, but may miss global optimum
- **Simulated Annealing:** Often finds better solutions than greedy, but takes longer
- **Genetic Algorithm:** Can find high-quality solutions, but requires most computational resources

### 6.2.2 Use Case Recommendations

- **Greedy:** Best for quick solutions or as initial solutions for other methods
- **Simulated Annealing:** Good for medium-sized problems where solution quality is important
- **Genetic Algorithm:** Best for complex problems where computational time is not a major constraint