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In [1]: import duckdb
import plotly.io as pio
from IPython.display import display

con = duckdb.execute("""IMPORT DATABASE '../data/db';""")
df = con.execute("""
                SELECT *
                FROM tasks
                """).df()

con.close()
df
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Out[1]: _____

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In [2]: import pandas as pd

ships = set(df["ship_id"])
roles = set(df["role"])

def _sort_tasks(task_list):
    return sorted(task_list, key=lambda t: pd.to_datetime(t.get("start_ts")))

tasks = {}
for ship in ships:
    tasks[ship] = {}

for _, row in df.iterrows():
    ship = row["ship_id"]
    role = row["role"]
    task_list = row["tasks"]
    tasks[ship][role] = _sort_tasks(task_list)
```

```
In [3]: import plotly.graph_objects as go

# Build prepared data per (ship, role)
def prepare_tasks_for(ship_id, role_name):
    task_list = tasks.get(ship_id, {}).get(role_name, [])
    prepared = []
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for t in task_list:
    start = pd.to_datetime(t.get("start_ts"))
    end = pd.to_datetime(t.get("end_ts"))
    if pd.isna(start) or pd.isna(end) or end <= start:
        continue
    prepared.append({
        "start": start,
        "end": end,
        "task_id": t.get("task_id") or t.get("id") or None,
        "sailing_id": t.get("sailing_id")
    })
if not prepared:
    return [], [], [], [], [], ship_id, role_name

prepared.sort(key=lambda x: x["start"])

# Greedy lane assignment to avoid overlaps
lanes_end = []
lane_idx_for_task = []
for task in prepared:
    placed = False
    for i, lane_end in enumerate(lanes_end):
        if task["start"] >= lane_end:
            lane_idx_for_task.append(i)
            lanes_end[i] = task["end"]
            placed = True
            break
    if not placed:
        lane_idx_for_task.append(len(lanes_end))
        lanes_end.append(task["end"])

t0 = prepared[0]["start"]
start_hours = [(t["start"] - t0).total_seconds() / 3600 for t in prepared]
end_hours = [(t["end"] - t0).total_seconds() / 3600 for t in prepared]
dur_hours = [eh - sh for sh, eh in zip(start_hours, end_hours)]
y_vals = [i + 1 for i in lane_idx_for_task]
return prepared, start_hours, dur_hours, y_vals, t0, ship_id, role_name

# Collect all combinations
ship_role_pairs = []
for s in sorted(tasks.keys()):
    for r in sorted(tasks[s].keys()):
        ship_role_pairs.append((s, r))

# Precompute traces for each combination
combo_traces = []
combo_meta = [] # metadata for each combo: (ship, role, n_traces, title, header)
for (s, r) in ship_role_pairs:
    prepared, start_hours, dur_hours, y_vals, t0, ship_id, role_name = prepare_tasks(s, r)
    traces = []
    for idx, (sh, dh, y, t) in enumerate(zip(start_hours, dur_hours, y_vals, prepared)):
        label = t["task_id"] if t["task_id"] is not None else f"{idx + 1}"
        sailing_label = t.get("sailing_id") or "N/A"
        hovertext = (
            f"Ship: {ship_id}<br>Role: {role_name}"

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        f"<br>Task ID: {label}"
        f"<br>Sailing ID: {sailing_label}"
        f"<br>Start: {t['start']}"
        f"<br>End: {t['end']}"
        f"<br>Duration (h): {dh:.2f}"
    )
    traces.append(go.Bar(
        x=[dh],
        y=[y],
        base=[sh],
        orientation="h",
        name=f"Task {label} | Sailing {sailing_label}",
        hovertext=hovertext,
        hoverinfo="text",
        marker=dict(line=dict(width=0.5, color="#777")),
        showlegend=True
    ))
    max_lane = max(y_vals) if y_vals else 1
    title = f"Tasks for Ship {ship_id} - Role {role_name}"
    height = max(420, 140 + 40 * max_lane)
    combo_traces.append(traces)
    combo_meta.append((s, r, len(traces), title, height))

# Build a single figure with all traces, only first combo visible
fig = go.Figure()
visibility = []

for ci, traces in enumerate(combo_traces):
    for tr in traces:
        fig.add_trace(tr)
        visibility.append(ci == 0) # only first combo visible

# Update layout for initial combo with dark theme
_, _, _, init_title, init_height = combo_meta[0]
fig.update_layout(
    template="plotly_dark",
    title=init_title,
    xaxis_title="Time (hours since first task start)",
    yaxis_title="Overlap lane",
    bargap=0.2,
    barmode="overlay",
    height=init_height,
    legend_title_text="Tasks (ID | sailing_id)",
    paper_bgcolor="#111111",
    plot_bgcolor="#111111",
    font=dict(color="#e6e6e6")
)
fig.update_yaxes(dtick=1, gridcolor="#333333")
fig.update_xaxes(gridcolor="#333333")

# Build dropdown for filtering (combined Ship & Role)
buttons = []
trace_offset = 0
combo_offsets = []
for (_, _, ntr, _, _) in combo_meta:
    combo_offsets.append((trace_offset, trace_offset + ntr))

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        trace_offset += ntr

    for ci, (s, r, ntr, title, height) in enumerate(combo_meta):
        vis = [False] * len(visibility)
        start_idx, end_idx = combo_offsets[ci]
        for i in range(start_idx, end_idx):
            vis[i] = True
        buttons.append(dict(
            label=f"{s} | {r}",
            method="update",
            args=[
                {"visible": vis},
                {"title": title, "height": height}
            ],
        ))

    fig.update_layout(
        updatemenus=[
            dict(
                type="dropdown",
                direction="down",
                buttons=buttons,
                x=0.0,
                y=1.15,
                xanchor="left",
                yanchor="top",
                showactive=True,
                bgcolor="#222222",
                bordercolor="#444444",
                font=dict(color="#f0f0f0", size=12)
            )
        ],
        margin=dict(t=110, r=20, l=60, b=40)
    )

    display(fig)

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In [4]: # Report count of tasks by sailing_id, role, and start_hour using prepare_tasks
def get_report(ship_id, role_name):
    target_ship = ship_id
    target_role = role_name

    prepared, start_hours, _, y_vals, t0, s_id, r_name = prepare_tasks_for(target_ship, target_role)

    report_rows = [
        {"sailing_id": task["sailing_id"], "role": r_name, "start_hour": start_hours, "count": count}
        for task, count in zip(prepared, y_vals)
    ]

    report_df = pd.DataFrame(report_rows)
    return report_df

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data = []
for sh in ships:
    for rl in roles:
        data.append((sh, roles, max(get_report(sh, rl)["count"])))
pd.DataFrame(data, columns=["ship_id", "role", "max_count"])

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Out[4]: _____

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In [5]: def get_tasks_by_sailing(ship_id, sailing_id):
        return {role: [t for t in tks if t["sailing_id"] == sailing_id] for role in roles}

selected_tasks = get_tasks_by_sailing('S1', 'S1-1')

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In [6]: def get_count(list_of_tasks):
        counts = {}
        for tup in list_of_tasks:
            counts.setdefault(tup, 0)
            counts[tup] += 1
        return counts

selected_3days_with_duplicates = {role: [(x['start_ts'], x['end_ts']) for x in selected_tasks
                                         if x['end_ts'] < pd.to_datetime('2025-01-01')] for role in roles}
selected_3days = {role: sorted(set(selected_3days_with_duplicates[role]), key=lambda x: x[0]) for role in roles}
counts = {role: get_count(selected_3days_with_duplicates[role]) for role in roles}
required_personal = {role: [counts[role][x] for x in selected_3days[role]] for role in roles}

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In [7]: import numpy as np

def gen_columns(rl, max_columns=None):
    tasks_sel = selected_3days[rl]
    number_of_tasks = len(tasks_sel)
    if number_of_tasks == 0:
        return np.zeros((0, 0), dtype=int)

    cols = []
    col = np.zeros(number_of_tasks, dtype=int)

    def dfs(i, next_one_allowed_at, has_one):
        if max_columns is not None and len(cols) >= max_columns:
            return
        if i == number_of_tasks:
            cols.append(col)
            return
        if i == next_one_allowed_at:
            col[i] = 1
            dfs(i+1, i+1, True)
            col[i] = 0
        else:
            dfs(i+1, next_one_allowed_at, has_one)
    dfs(0, 0, False)
    return cols

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        return
    if i == number_of_tasks:
        if has_one:
            cols.append(col.copy())
        return
    # place 0
    col[i] = 0
    dfs(i + 1, next_one_allowed_at, has_one)
    # place 1 if allowed (keep at least two zeros between ones)
    if i >= next_one_allowed_at:
        col[i] = 1
        dfs(i + 1, i + 3, True)
        col[i] = 0 # reset

dfs(0, 0, False)
if not cols:
    return np.zeros((number_of_tasks, 0), dtype=int)
return np.stack(cols, axis=1)

```

In [8]: `import gurobipy as gp`

```

class Solution:
    def __init__(self, role):
        self.role = role
        self.a_matrix = gen_columns(self.role)
        self.required_personal = np.array(required_personal[self.role])
        self.ordered_unique_tasks = selected_3days[self.role]
        self.model = gp.Model(self.role)
        self.variables = self.model.addMVar(self.a_matrix.shape[1], vtype=gp.GRB.BINARY)
        self.solution = None

    def solve(self):
        self.model.setObjective(self.variables.sum(), gp.GRB.MINIMIZE)
        self.model.addConstr(self.a_matrix @ self.variables == self.required_personal)
        self.model.optimize()
        self.model.write(f'{self.role}-sp.mps')
        self.model.write(f'{self.role}-sp.lp')
        return self

    def get_solution(self):
        selected_patterns = []
        for var in self.model.getVars():
            var_name = var.varName
            var_value = var.x
            if var_value > 0.5:
                selected_patterns.append(int(var_name.split("[")[1].split("]")[0]))
                print(selected_patterns)
        self.solution = {"patterns": self.a_matrix[:, selected_patterns]}
        schedules = {}
        for i in selected_patterns:
            schedules[f"schedule_{selected_patterns.index(i)}"] = [self.ordered_unique_tasks[selected_patterns.index(i)]]
        self.solution["schedules"] = schedules
        return self

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def run(self):
    self.solve()
    self.get_solution()
    return self
```

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In [9]: solutions = {role: Solution(role).run().solution for role in roles}
```

Restricted license - for non-production use only - expires 2026-11-23
Gurobi Optimizer version 12.0.3 build v12.0.3rc0 (linux64 - "Ubuntu 24.04.3 LTS")
CPU model: 12th Gen Intel(R) Core(TM) i5-12500H, instruction set [SSE2|AVX|AVX2]
Thread count: 16 physical cores, 16 logical processors, using up to 16 threads
Optimize a model with 11 rows, 87 columns and 208 nonzeros
Model fingerprint: 0x4d9dd83c
Variable types: 0 continuous, 87 integer (87 binary)
Coefficient statistics:
Matrix range [1e+00, 1e+00]
Objective range [1e+00, 1e+00]
Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Found heuristic solution: objective 11.0000000
Presolve time: 0.00s
Presolved: 11 rows, 87 columns, 208 nonzeros
Variable types: 0 continuous, 87 integer (87 binary)
Root relaxation: objective 3.000000e+00, 23 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node		Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node Time
*	0	0		0	3.0000000	3.000000	0.00%	- 0s

Explored 1 nodes (23 simplex iterations) in 0.03 seconds (0.00 work units)
Thread count was 16 (of 16 available processors)
Solution count 2: 3 11
Optimal solution found (tolerance 1.00e-04)
Best objective 3.000000000000e+00, best bound 3.000000000000e+00, gap 0.0000%
[39]
[39, 58]
[39, 58, 86]
Gurobi Optimizer version 12.0.3 build v12.0.3rc0 (linux64 - "Ubuntu 24.04.3 LTS")
CPU model: 12th Gen Intel(R) Core(TM) i5-12500H, instruction set [SSE2|AVX|AVX2]
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Bounds range [1e+00, 1e+00]
RHS range [1e+00, 1e+00]
Found heuristic solution: objective 11.0000000
Presolve time: 0.00s

```

Presolved: 11 rows, 87 columns, 208 nonzeros
Variable types: 0 continuous, 87 integer (87 binary)
Root relaxation: objective 3.000000e+00, 23 iterations, 0.00 seconds (0.00 work units)
      Nodes |      Current Node |      Objective Bounds |      Work
  Expl Unexpl |  Obj  Depth IntInf | Incumbent    BestBd   Gap | It/Node Time
*    0      0 |          0        | 3.00000000    3.00000 0.00% | -      0s
Explored 1 nodes (23 simplex iterations) in 0.02 seconds (0.00 work units)
Thread count was 16 (of 16 available processors)
Solution count 2: 3 11
Optimal solution found (tolerance 1.00e-04)
Best objective 3.000000000000e+00, best bound 3.000000000000e+00, gap 0.0000%
[39]
[39, 58]
[39, 58, 86]
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  Objective range   [1e+00, 1e+00]
  Bounds range      [1e+00, 1e+00]
  RHS range         [1e+00, 1e+00]
Found heuristic solution: objective 11.00000000
Presolve time: 0.00s
Presolved: 11 rows, 87 columns, 208 nonzeros
Variable types: 0 continuous, 87 integer (87 binary)
Root relaxation: objective 3.000000e+00, 23 iterations, 0.00 seconds (0.00 work units)
      Nodes |      Current Node |      Objective Bounds |      Work
  Expl Unexpl |  Obj  Depth IntInf | Incumbent    BestBd   Gap | It/Node Time
*    0      0 |          0        | 3.00000000    3.00000 0.00% | -      0s
Explored 1 nodes (23 simplex iterations) in 0.03 seconds (0.00 work units)
Thread count was 16 (of 16 available processors)
Solution count 2: 3 11
Optimal solution found (tolerance 1.00e-04)
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Optimize a model with 11 rows, 87 columns and 208 nonzeros
Model fingerprint: 0x4d9dd83c

```


Variable types: 0 continuous, 87 integer (87 binary)

Coefficient statistics:

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

Found heuristic solution: objective 11.0000000

Presolve time: 0.00s

Presolved: 11 rows, 87 columns, 208 nonzeros

Variable types: 0 continuous, 87 integer (87 binary)

Root relaxation: objective 3.0000000e+00, 23 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node			Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
* 0	0			0	3.0000000	3.00000	0.00%	-	0s

Explored 1 nodes (23 simplex iterations) in 0.02 seconds (0.00 work units)

Thread count was 16 (of 16 available processors)

Solution count 2: 3 11

Optimal solution found (tolerance 1.00e-04)

Best objective 3.0000000000000e+00, best bound 3.0000000000000e+00, gap 0.0000%

[39]

[39, 58]

[39, 58, 86]

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Thread count: 16 physical cores, 16 logical processors, using up to 16 threads

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Model fingerprint: 0x4d9dd83c

Variable types: 0 continuous, 87 integer (87 binary)

Coefficient statistics:

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

Found heuristic solution: objective 11.0000000

Presolve time: 0.00s

Presolved: 11 rows, 87 columns, 208 nonzeros

Variable types: 0 continuous, 87 integer (87 binary)

Root relaxation: objective 3.0000000e+00, 23 iterations, 0.00 seconds (0.00 work units)

Nodes		Current Node			Objective Bounds			Work	
Expl	Unexpl	Obj	Depth	IntInf	Incumbent	BestBd	Gap	It/Node	Time
* 0	0			0	3.0000000	3.00000	0.00%	-	0s

Explored 1 nodes (23 simplex iterations) in 0.02 seconds (0.00 work units)

Thread count was 16 (of 16 available processors)

Solution count 2: 3 11

Optimal solution found (tolerance 1.00e-04)

Best objective 3.0000000000000e+00, best bound 3.0000000000000e+00, gap 0.0000%

[39]

[39, 58]

[39, 58, 86]

```

In [10]: import pandas as pd
import plotly.express as px

# --- Flatten your solutions dict into a DataFrame ---
records = []
for role, data in solutions.items():
    for schedule_name, intervals in data["schedules"].items():
        for start, end in intervals:
            records.append({
                "role": role,
                "schedule": schedule_name,
                "start": start,
                "end": end
            })

df = pd.DataFrame(records)
df = df.sort_values(by=["role", "schedule", "start"])

# --- Convert datetimes to hours since first start (relative time axis) ---
t0 = df["start"].min()
df["start_hours"] = (df["start"] - t0).dt.total_seconds() / 3600
df["duration_hours"] = (df["end"] - df["start"]).dt.total_seconds() / 3600

# --- Generate one plot per role ---
figs = {}
for role in df["role"].unique():
    role_df = df[df["role"] == role]
    fig = px.bar(
        role_df,
        x="duration_hours",
        y="schedule",
        base="start_hours", # start of bar
        color="schedule",
        orientation="h",
        title=f"Schedules for Role: {role} (Hours since first start)"
    )
    fig.update_layout(
        xaxis_title="Time (hours)",
        yaxis_title="Schedules",
        bargap=0.3,
        height=500
    )
    fig.update_yaxes(autorange="reversed") # keep first schedule on top
    figs[role] = fig

# --- Show one of them (for example, first role) ---
for rl in list(figs.keys()):
    display(figs[rl])

```

In []: