

# Experiments

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Constraint Based Production Scheduling

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## Key Points

- This section describes the scheduling tool
  - This is a *preview* of the current state, not released yet!
- How to load/create data
  - From files
  - By instance generator
  - From benchmark problems
- How to run the solvers
  - Which solvers are supported
  - What to expect in terms of performance
- Experiments to try
  - Limited time
  - Possible "test before invest" continuation

## 1 The Scheduling Tool

### The Scheduling Tool

- We create the tool as basis for experiments
- To test ideas and solvers
- As a teaching tool
- Slightly higher standard than usual academic prototypes
  - This is a *preview*, not released yet
- Not a commercial tool
  - But can use commercial solvers
  - Also open-source solvers
- Written in Java, JavaFX
- Can also be used as a back-end scheduling server
- Uses our Java application framework generator
- Will become available in early 2025

## 2 Under the hood

### Back-end solvers

- Provide both open-source and commercial solver interfaces
- Allow experimentation without having to buy commercial tools straightaway
- Gives a level playing field to compare solvers and models
- Provides out-of-the-box, generic performance

## 2.1 Google CPSSat

### Google OR-Tools CPSSat Solver

- Open-Source tool provided by Google
- Available at [https://developers.google.com/optimization/cp/cp\\_solver](https://developers.google.com/optimization/cp/cp_solver)
- Probably best open-source CP solver for scheduling
- This solver is packaged with scheduler

### Example Problem

Below is a simple example of a job shop problem, in which each task is labeled by a pair of numbers  $(m, p)$  where  $m$  is the number of the machine the task must be processed on and  $p$  is the *processing time* of the task – the amount of time it requires. (The numbering of jobs and machines starts at 0.)

- job 0 =  $[(0, 3), (1, 2), (2, 2)]$
- job 1 =  $[(0, 2), (2, 1), (1, 4)]$
- job 2 =  $[(1, 4), (2, 3)]$

In the example, job 0 has three tasks. The first,  $(0, 3)$ , must be processed on machine 0 in 3 units of time. The second,  $(1, 2)$ , must be processed on machine 1 in 2 units of time, and so on. Altogether, there are eight tasks.

### A solution for the problem

A solution to the job shop problem is an assignment of a start time for each task, which meets the constraints given above. The diagram below shows one possible solution for the problem:



You can check that the tasks for each job are scheduled at non-overlapping time intervals, in the order given by the problem.

The length of this solution is 12, which is the first time when all three jobs are complete. However, as you will see [below](#), this is not the optimal solution to the problem.

(from OR-Tools website)

## 2.2 IBM CPOptimizer

### CP Optimizer from IBM

- Commercial tool of IBM
- <https://www.ibm.com/products/ilog-cplex-optimization-studio/cplex-cp-optimizer>
- Part of optimization suite with Cplex, OPL
- We do **not** provide this solver, we allow to interface with it

- Academic licenses available
- Well-known for capabilities for scheduling

# Resources



## Applications of constraint programming

Explore applications of constraint programming including production problem and scheduling use cases.

[Read the documentation →](#)

(from CPOptimizer website)

### 2.3 MiniZinc

#### MiniZinc from Monash University

- Modelling language and backend tools from Monash University in Melbourne, Australia
- Available from <https://www.minizinc.org/>

- Widely used for teaching
- Allows different backend solver to run from same model
- Generic CP tool, not optimized for scheduling
- Requires separate installation, open-source

The screenshot shows the MiniZinc website. On the left, there's a 'Get started' button and a note about Windows 10 compatibility. Below that are links for the latest release (2.8.7), packages, source code, and license information. A footer note says MiniZinc is developed at Monash University with support from OPTIMA. On the right, there's a large grid titled 'Rostering' showing a weekly schedule for ten people (Aimee to Paula) across five days (Mon to Fri). The grid uses colors to represent shifts: dark blue for Night, light blue for Day, and yellow for Evening. Some cells are blacked out, likely indicating unavailable times. The grid is labeled 'Rostering' at the bottom right.

(from MiniZinc Website)

## 2.4 Which solver is better?

### Which Solver is Better?

- We present results on a few benchmark types
- Fair comparison between solvers
  - Same hardware, Windows 11 laptop
  - CPU i7-10875H @ 2.3GHz, 64GB, four cores
  - Same timeout (600 s)
- Not a fair comparison to state-of-the-art
  - Uses out-of-the-box model
  - Significant improvements possible
  - More specific models
  - Parameter tuning
  - Unlimited runtime

### Taillard Job-Shop Benchmarks

Group	Nr	All Instances				Optimal Only		Non Optimal Only					
		Optimal (%)	All Instances (%)	Both	CPO	CPSat	None	Time (%) of VB)	CPO	CPSat	Cost (%) of VB)	CPO	CPSat
15/15	10	90.00	0.00	0.00	10.00	105.19	141.18	100.00	100.00	97.17	100.00	100.00	
20/15	10	20.00	0.00	0.00	80.00	267.27	263.20	100.99	100.05	98.50	99.93		
20/20	10	0.00	0.00	0.00	100.00	n/a	n/a	100.74	100.06	97.96	100.00		
30/15	10	10.00	0.00	10.00	80.00	174.32	100.00	100.18	100.49	99.87	100.00		
30/20	10	0.00	0.00	0.00	100.00	n/a	n/a	100.30	101.30	99.40	100.00		
50/15	10	100.00	0.00	0.00	0.00	100.00	685.09	n/a	n/a	n/a	n/a		
50/20	10	10.00	60.00	0.00	30.00	100.00	381.38	100.00	101.60	100.00	100.00		
100/20	10	10.00	90.00	0.00	0.00	100.00	416.13	100.00	101.73	100.00	66.81		

- Significant number of problems solved to optimality in 600s
- In terms of quality, solvers are quite similar
- CPO wins in terms of solution times for larger instances

## Results for Hybrid Flexible Flow-Shop

Group	Nr	All Instances			Optimal Only		Non Optimal Only			
		Optimal (% of All Instances)			Time (% of VB)		Cost (% of VB)		Bound (% of VB)	
		Both	CPO	CPSat	CPO	CPSat	CPO	CPSat	CPO	CPSat
20	25	76.00	0.00	20.00	4.00	100.00	580.71	100.00	96.52	100.00
25	25	80.00	0.00	8.00	12.00	101.65	238.02	100.00	100.37	97.67
30	25	60.00	0.00	4.00	36.00	100.35	264.69	100.18	101.05	100.00
40	25	4.00	16.00	0.00	80.00	100.00	2554.03	100.00	104.68	100.00
50	25	0.00	4.00	0.00	96.00	n/a	n/a	100.00	107.87	100.00
100	25	0.00	0.00	0.00	100.00	n/a	n/a	100.00	120.43	100.00
200	25	0.00	0.00	0.00	100.00	n/a	n/a	100.00	188.60	100.00
300	24	0.00	0.00	0.00	100.00	n/a	n/a	100.00	263.22	100.00
400	25	0.00	0.00	0.00	100.00	n/a	n/a	100.00	246.34	100.00

- Only smaller/medium instances solved to optimality
- For those problems, both solvers perform well
- CPO significantly better on large instances

## General Recommendations

- If you already have access to CPO, use it!
- For new problem types, do an evaluation with CPSat first
- Out of the box, CPO performs more consistently
- May be easier to extend CPSat with your own research
- Use multiple cores and memory to your advantage

## 3 Input Data

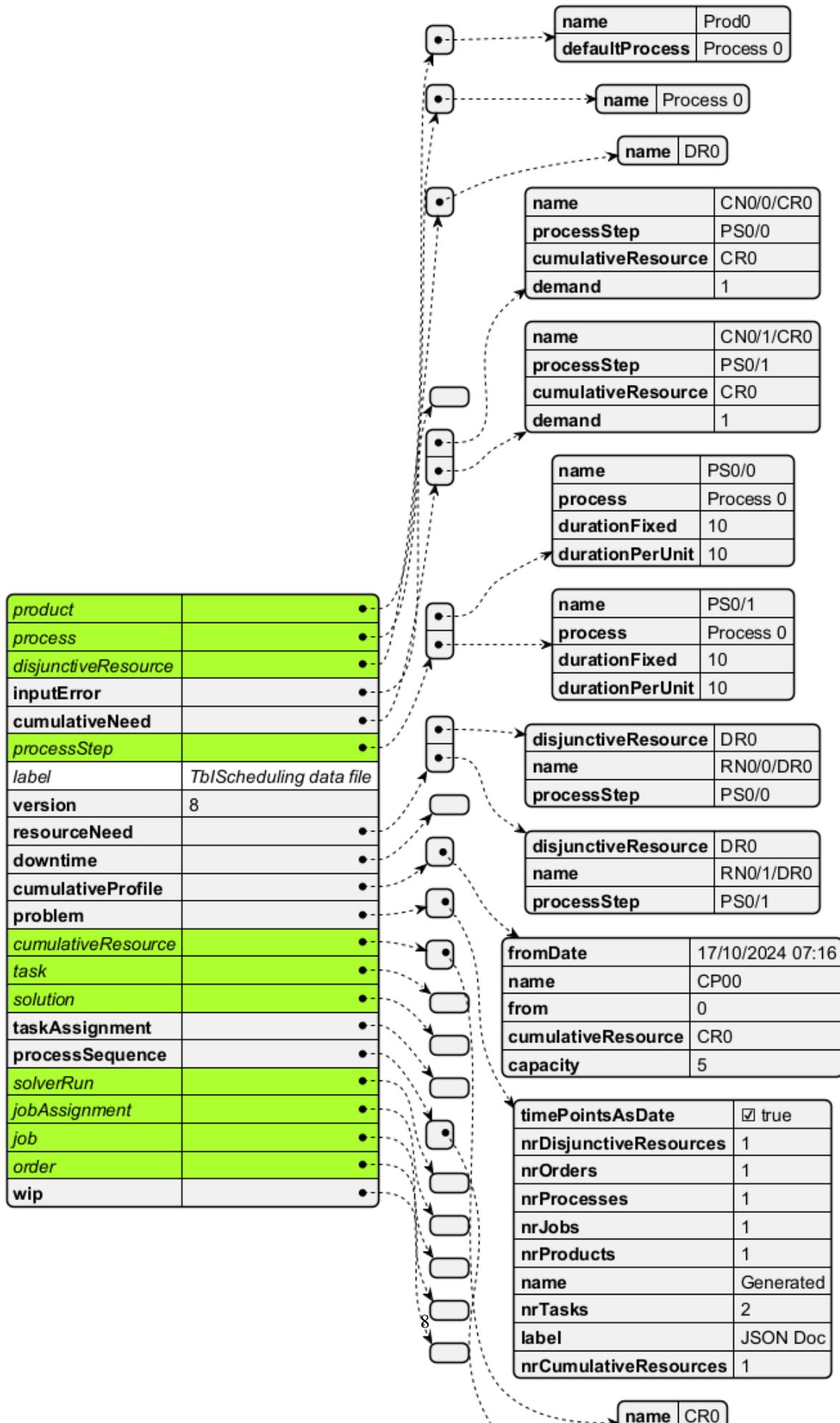
### Input Data

- We have defined a specific JSON data format to describe scheduling problems
- This is different from the native/XML data format of the application (do not use)
- Load with menu File – Load DataFile...
- Save with menu File – Save DataFile...
- The format is described in a document

### Base Data

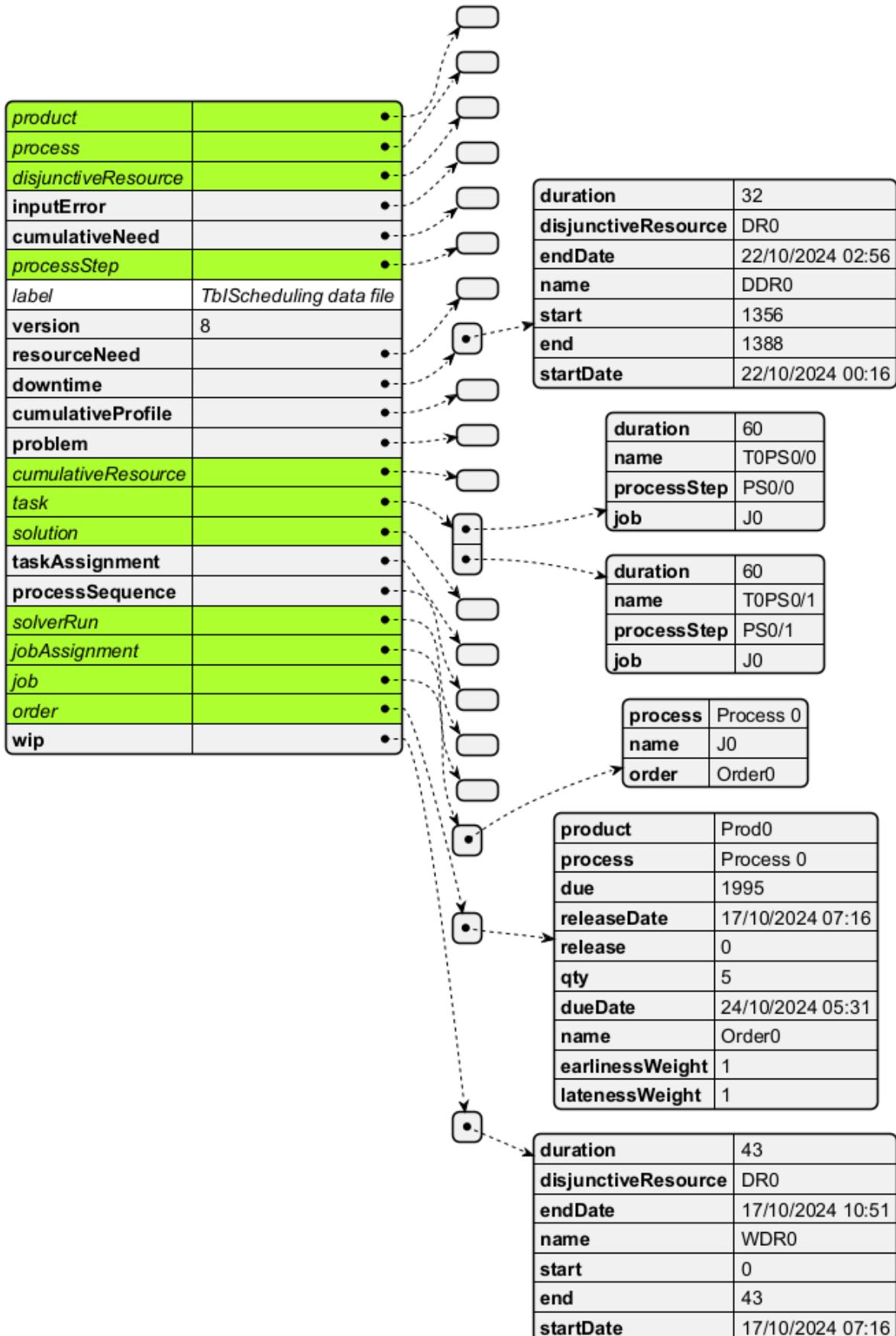
- Description of
  - Product
  - Process
  - DisjunctiveResource
  - CumulativeNeed
  - ProcessStep
  - ResourceNeed
  - CumulativeProfile
  - Problem
  - CumulativeResource

- ProcessSequence



## **Schedule Input Data**

- Description of
  - Downtime
  - Task (x2)
  - Job
  - Order
  - WiP



## 4 Result Output

### Result Data

- We use the same JSON format to describe the results of the schedule
- Added field types for SolverRun, Solution, assigned Jobs and Tasks

### Sample Results

- Description of
  - Solution
  - SolverRun
  - Job Assignment
  - Task Assignment

product	•
process	•
disjunctiveResource	•
inputError	•
cumulativeNeed	•
processStep	•
label	TblScheduling data file
version	8
resourceNeed	•
downtime	•
cumulativeProfile	•
problem	•
cumulativeResource	•
task	•
solution	•
taskAssignment	•
processSequence	•
solverRun	•
jobAssignment	•
job	•
order	•
wip	•

Relationships:

- From processStep to processStep (self-loop)
- From processStep to process
- From process to processStep

makeSpan	163
totalEarliness	1832
endDate	17/10/2024 20:51
totalWaitBefore	0
nrEarly	1
percentEarly	100
duration	120
maxLateness	0
maxWaitAfter	0
nrLate	0
gap	0
solverRun	Run1
end	163
weightedEarliness	1832
totalLateness	0
maxWaitBefore	0
bound	163
solverStatus	Optimal
start	43
totalWaitAfter	0
percentLate	0
weightedLateness	0
maxEarliness	1832
name	Sol1
objectiveValue	163
flowtime	163
startDate	17/10/2024 10:51

duration	60
waitAfter	0
task	T0PS0/0
disjunctiveResource	DR0
endDate	17/10/2024 15:51
name	TA0
start	43
end	103
jobAssignment	JA0
waitBefore	0
startDate	17/10/2024 10:51

duration	60
waitAfter	0
task	T0PS0/1
disjunctiveResource	DR0
endDate	17/10/2024 20:51
name	TA1
start	103
end	163
jobAssignment	JA0
waitBefore	0
startDate	17/10/2024 15:51

nrThreads	2
weightEarliness	1
seed	42
solverBackend	None
weightMakespan	1
producePDF	<input type="checkbox"/> false
solverStatus	Optimal
description	
label	
modelType	CPO
enforceWip	<input checked="" type="checkbox"/> true
objectiveType	Makespan
timeout	30
weightFlowtime	1
enforceCumulative	<input checked="" type="checkbox"/> true
name	Run1
weightLateness	1
time	0.142
produceReport	<input type="checkbox"/> false
enforceDowntime	<input checked="" type="checkbox"/> true
removeSolution	<input type="checkbox"/> false
enforceReleaseDate	<input checked="" type="checkbox"/> true
enforceDueDate	<input type="checkbox"/> false

## 5 Instance Generator

### Instance Generator

- Application allows to generate different types of test problems
- Different types of resource models
- Different numbers of orders, resources, WiP, downtime
- Useful to generate more life-like examples combining different constraint types

### Instance Generator Dialog

- Resource Model
  - Select a resource model defining the overall structure of problem
- Nr Disjunctive Resources
  - Describe how many disjunctive resources are generated
- Resource Probability
  - The probability that a resource is compatible with a task
  - Only for some resource models

 Data Generator Parameters X

---

Label:

StartDate:  Start Time: 0 : 00 Clock

---

Resource Model: HybridFlowShop ▾

Nr Disjunctive Resources:

Resource Probability:

---

▶ Products

▶ Duration

▼ Cumulative Resource

Nr Cumulative Resources:

Cumul Demand Range:

Profile Pieces:

Cumul Capacity Range:

---

▶ Orders

▶ WiP

▶ Downtime

▶ Other Parameters

---

Run Cancel

## Resource Models

- Flow-Shop
  - Multiple stages, all jobs use machines in same order
- Job-Shop
  - Multiple stages, jobs use machines in different order
- Open-Shop
  - Multiple stages, no predefined order of machines
- Hybrid Flow-Shop (default)

- Hybrid Job-Shop
- Hybrid Open-Shop
  - Like x-shop, but with multiple machines per stage
- Random
  - Multiple stages, each stage using a random subset of machines
- All
  - Multiple stages, each stage allowing all machines

### **Instance Generator - Products**

- Nr Products
  - Number of products to be generated
  - Products may be reused by multiple orders
- Stages Range
  - Range slider, sets lower and upper bound on number of stages

**Data Generator Parameters**

Label:

StartDate:   Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

**Products**

Nr Products:

Stages Range:

► Duration

► Cumulative Resource

► Orders

► WiP

► Downtime

► Other Parameters

### Instance Generator - Duration

- Duration Model
  - Different ways to link duration of processSteps
- Duration Range
  - Range slider to set lower and upper bounds on perUnit duration
- Duration Fixed Factor
  - How fixed and perUnit duration values are linked

**Data Generator Parameters**

Label:

StartDate:   Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

**Products**

**Duration**

Duration Model:

Duration Range:

Duration Fixed Factor:

**Cumulative Resource**

**Orders**

**WiP**

**Downtime**

**Other Parameters**

### Instance Generator - Cumulative

- Nr Cumulative Resources
  - Number of cumulative resources generated
- Cumul Demand Range
  - Range slider to select lower and upper bound on cumulativeResourceNeed demands
- Profile Pieces
  - Number of segments of CumulativeProfile generated for each resource
- Cumul Capacity Range

- Range slider to select lower and upper bounds on cumulative profile capacity values

**Data Generator Parameters**

Label:

StartDate:  Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

**Cumulative Resource**

Nr Cumulative Resources:

Cumul Demand Range:

Profile Pieces:

Cumul Capacity Range:

Run

### Instance Generator - Orders

- Nr Orders
  - Number of orders generated, each order is assigned a random product/process
- Qty Range
  - Range slider to select lower and upper bounds on quantity for each order

**Data Generator Parameters**

Label:

StartDate:  Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

**Products**

**Duration**

**Cumulative Resource**

**Orders**

Nr Orders:

Qty Range:

**WiP**

**Downtime**

**Other Parameters**

### Instance Generator - WiP (Work in Progress)

- WiP Probability
  - Probability of generating a WiP for a disjunctive resource
- WiP Range
  - Range slider to set lower and upper bound on WiP duration

**Data Generator Parameters**

Label:

StartDate:   Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

► Products

► Duration

► Cumulative Resource

► Orders

▼ WiP

WiP Probability:

WiP Range:

► Downtime

► Other Parameters

### Instance Generator - Downtime

- Downtime Probability
  - Probability of generating a downtime for a disjunctive resource
- Downtime Range
  - Range slider to select lower and upper bounds on downtime duration

**Data Generator Parameters**

Label:

StartDate:   Start Time:

Resource Model:

Nr Disjunctive Resources:

Resource Probability:

► Products

► Duration

► Cumulative Resource

► Orders

► WiP

▼ Downtime

Downtime Probability:

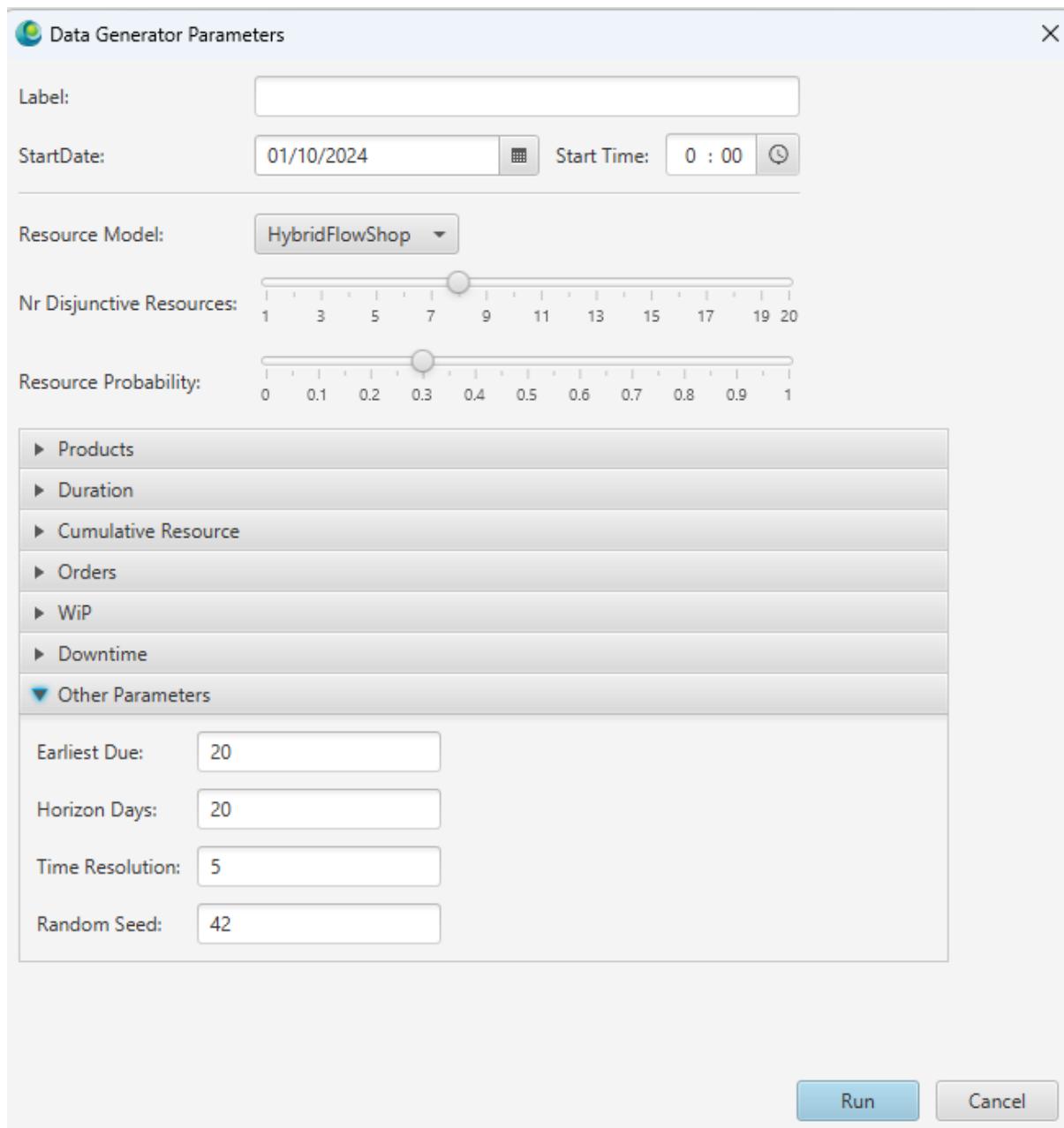
Downtime Range:

► Other Parameters

### Instance Generator - Other Parameters

- Earliest Due
  - Smallest allowed value for a due date
- Horizon Days
  - What planning horizon to consider (in days)
- Time Resolution
  - In minutes, links internal and external time presentation
- Random Seed

- Random seed to make reproducible random choices



## 6 Predefined Problem Sets

### 6.1 Taillard

#### Taillard Scheduling Benchmarks

- Three datasets of different sizes
  - Job-shop
  - Flow-shop
  - Open-shop

- Load with menu `File -> Load DataFile...` – Taillard –
- Larger instances need more solver time to reach good solutions (600 s)

## 6.2 SALBP

### Simple Assembly Line Balancing Problem (SALBP)

- Will be discussed in more details as case study
- Design an assembly line setup by solving a scheduling problem
- Balance a set of operations across a number of stations of an assembly line
- Precedence graph is not a chain, can be very complex
- Specialized problem normally solved with specialized tools
- Load with menu `File -> Load SALBP Problem...`

## 6.3 Test Scheduling

### Test Scheduling Benchmark set from ABB

- Will be discussed in more details as case study
- Schedule a set of tests on a number of machines, minimizing total duration
- Single stage tests, possibly large number of resources
- Closely related to bin-packing
- Load with menu `File -> Load Test Scheduling Problem...`

## 7 Some Suggested Experiments

### Experiment 1

- Start the application
  - Our running example will be automatically generated
- Look at the process diagram `Window–Product–Process Diagram`
- Run the solver `Scenario -> Run ScheduleJobs Solver`
- Observe the results in Gantt Chart
- Customize display
- Look at Cumulative Resource Chart `Window–Solution–Cumulative Resource Chart`

### Experiment 2

- Re-run solver disabling cumulative constraint
- Observe result in Gantt chart
- See impact on Cumulative Resource chart
- Switch between solutions in charts

### **Experiment 3**

- Check Gantt chart display for delayed tasks, enabling lateness display
- Re-run solver, enforcing due-date constraints
- What impact does this have on objective

### **Experiment 4**

- Change objective to on-time delivery
- Results are very different, why?
- More explanations on this tomorrow

### **Experiment 5**

- Load one of the other example types
- For example, Taillard Job-shop 15x15
- Understand process diagram
- Run solver
- Look at intermediate solutions found

## **8 Summary**

### **Summary**

- We presented an overview of our generic scheduling tool
- Discussed available solvers, both commercial and open-source
- Described the JSON data format for input and output
- Gave an overview of the instance generator provided
- Shows example problems included with tool
- Suggested some experiments to run