

# Production scheduling

## 1 The project task

The project task is to define and implement a decomposition algorithm for the scheduling problem described below. For this, there are two alternatives, composed by the two models *the engineer's model* and *the time-discrete model*. The algorithm that should be used is the *Dantzig-Wolfe reformulation solved by column generation*.

### 1.1 Problem files and implementation

At your service, AMPL model files (.mod) for the two mathematical models are provided at the course homepage together with a script file (.run) and a set of data files (.dat) for each model. These files can be used to solve the problem cases by the Cplex solver, to get key solutions for comparisons. All files named \*MTC3\* correspond to the engineer's model and all files named \*MTC4\* correspond to the time-discrete model.

The decomposition algorithm chosen should be implemented in AMPL or Matlab (or C/C++ if you prefer). The sub and master problems should be solved by Cplex MILP solver.

Instructions on how to write scripts in the AMPL command language can be found, e.g., at the web page [www.ampl.com/NEW/index.html#LoopTest](http://www.ampl.com/NEW/index.html#LoopTest) under "Looping and Testing 1 & 2". Instructions for the AMPL and Matlab interfaces to Cplex is found, e.g., in [www.math.chalmers.se/Math/Grundutb/CTH/mve165/0910/Exercises/LP\\_Exc\\_1003.pdf](http://www.math.chalmers.se/Math/Grundutb/CTH/mve165/0910/Exercises/LP_Exc_1003.pdf)

### 1.2 Deadlines and examination

A preliminary report should be sent to the teachers and the opponent(s) on Wednesday, 6th of October and the final report should be sent to the teachers and the opponent(s) on Monday, 11th of October. The program codes should be sent to the teachers on Monday, 11th of October.

The examination of the project includes an oral presentation of the same and an opposition to another groups project, during the seminar on Wednesday, 13th of October. The competition will also be decided at this seminar.

### 1.3 Applying the algorithm to the models

The machining problem (see Section 4) should be solved using the column generation method, but the feasibility problem may be solved directly, without decomposition, using the Cplex MILP-solver.

In the column generation algorithm applied to the machining problem, a column is suitably defined by one schedule (composed by a sequence of operations) for each machine (and for the time-discrete model also the corresponding allocation of operations to machines). When a final set of columns has been generated (what “final” means will be your decision), the corresponding restricted master problem, with integer requirements on the appropriate variables, may be solved by Cplex MIP-solver.

If any of these instructions or the AMPL-files is unclear or seems unsuitable, please tell the teachers and your fellow students.

You are also welcome to discuss any difficulty or indistinct instruction in this assignment with any of the teachers, but don't forget to discuss also with your fellow students.

### 1.4 Presentation of results

The types of results that should be presented in the report and at the seminars include upper and lower bounds on the optimal value as functions of number of iterations and of CPU-time, the best solution (i.e., schedule) found, the value of the best solution found relative to the key solution (found using the AMPL-files supplied).

In order to receive solutions (schedules) with different properties, different objective functions should be formulated and tried out in the computations.

The specific competition task will be revealed around the 11th of October.

## 2 Definition of the problem

### 2.1 Indices and sets

The queue of jobs  $j$  to the Multi task cell go through three different phases:

- Planned orders not yet released, i.e. exist only in the planning system.
- Released jobs, or so called production orders, i.e. physical parts being processed elsewhere on their way to the MT cell.
- Jobs checked in into the MT cell, i.e. parts inside the MT cell waiting to be processed.

$\mathcal{J}$  denotes the whole set of jobs to be done during the planning period. Some jobs are to be processed on the same part, and the pairs of two such jobs adjacent in the routing, form the set  $\mathcal{Q} \subset \mathcal{J} * \mathcal{J}$ . For the part, which routing is described by Figure 1, the pairs  $(j, q)$  and  $(q, l)$  belongs to the set  $\mathcal{Q}$ .

Each job  $j$  consists of  $n_j$  operations  $i$  to be processed inside the MT cell,  $i \in \mathcal{N}_j = \{1, \dots, n_j\}$ . A typical routing for a job is listed in the table below.

In order to fix the order in the schedule between two jobs of the same type for the same type of parts, the set  $\mathcal{P} \subset \mathcal{J} \times \mathcal{J}$  is populated by pairs  $(j, q)$  of these jobs where the release date of job  $j$  is less than or equal to the release

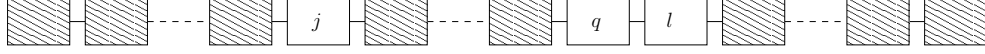


Figure 1: Routing of a part with jobs  $j$ ,  $q$  and  $l$ .  $(j, q), (q, l) \in \mathcal{Q}$ .

| $i$ | Description              |
|-----|--------------------------|
| 1   | Mounting into fixture    |
| 2   | Turning/Milling/Drilling |
| 3   | Manual deburring         |
| 4   | Automatic deburring      |
| 5   | Demounting               |

Table 1: The possible route operations  $i$  in the Multi task cell.

date of job  $q$ . The set  $\mathcal{K}$  denotes the 10 resources  $k$  that the Multi task cell consists of.

| $k$      | Description                 |
|----------|-----------------------------|
| MC 1-5   | Multi Task machines         |
| Man Gr   | Manual deburring station    |
| DBR      | Automatic deburring machine |
| M/DM 1-3 | Mount/demount stations      |

Table 2: The resources  $k$  of the Multi task cell

## 2.2 Parameters

$$\lambda_{ijk} = \begin{cases} 1, & \text{if operation } (i, j) \text{ can be processed on resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$a_k$  : the time when resource  $k$  will be available the first time.

$d_j$  : the due date of job  $j$ , i.e. the point in time when the last operation  $n_j$  of job  $j$  is planned to be completed.

$r_j$  : the release date for job  $j$ .

$\tau_{ij}$  : the processing time in hours for operation  $i$  of job  $j$

$w$  : the transportation time for a product inside the multi task cell

$v_{jq}$  : the interoperation time between the jobs  $j$  and  $q$ , where  $(j, q) \in \mathcal{Q}$

$M$  : a sufficiently large positive number, i.e. greater than the planning horizon.

All dates described above are given in hours relative to a time point,  $t_0$ , which is the starting time of the schedule to be calculated.

## 2.3 Realistic release dates and interoperation times

If a job  $j$  is checked in into the MT cell, i.e. the part is ready to be processed at time  $t_0$ ,  $r_j$  is set to 0. Release dates for the other two phases, i.e. released jobs and planned orders, see 2.1, are not that easy to get hold on. In the planning

system of the MT cell, there are a planned latest release date for each job, let us denote it  $\varrho_j$ . This means that the job  $j$  in the MT cell is planned to be started at the latest at this point in time. The desired release date,  $r_j$ , we want to get hold on in the optimization model is however the realistic point in time when the part arrives to the MT cell. The best guess we can make is that this is given by

$$r_j = \max(\varrho_j - t_0 - \delta_j; \nu_j^0),$$

where  $\delta_j$  is the planned transport time for transporting the physical part from the preceding operation of job  $j$  to the MT cell and  $\nu_j^0$  is the standard lead time from the operation where the part is about to be processed at time  $t_0$  till it arrives in the Multi task cell. Let  $\mu_{act}$  denote this actual operation. Then  $\nu_j^0$  is given by

$$\nu_j^0 = \sum_{\mu=\mu_{act}}^{\mu_{m_j-1}} (\rho_\mu + \varsigma_\mu + \vartheta_\mu) + \delta_j,$$

where  $\rho_\mu$ ,  $\varsigma_\mu$  and  $\vartheta_\mu$  denote the process, setup and queue times of operation  $\mu$  processed elsewhere, i.e. not in the MT cell, see Figure 2.

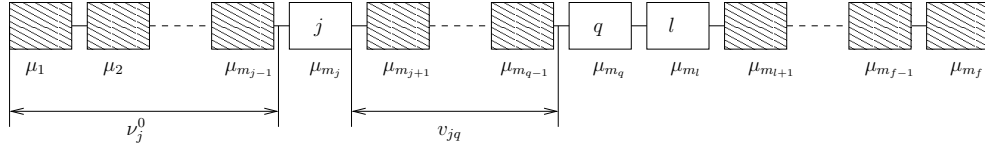


Figure 2: The interoperation time,  $v_{jq}$ , is marked out together with  $\nu_j^0$  for the case of a planned order, i.e.  $\mu_{act} = \mu_1$ .

In order to prevent the possibility that job  $j$  and  $q$  are scheduled too close in time, when both jobs are to be performed on the same physical part, the parameter  $v_{jq}$  is introduced. This is the planned interoperation time between the completion time of job  $j$  and the starting time of job  $q$  for jobs done outside the multi task cell. The definition of  $v_{jq}$  is

$$v_{jq} = \sum_{\mu=\mu_{m_j+1}}^{\mu_{m_q-1}} (\rho_\mu + \varsigma_\mu + \vartheta_\mu) + \delta_q,$$

where  $\rho_\mu$ ,  $\varsigma_\mu$  and  $\vartheta_\mu$  denote the process, setup and queue times of operation  $\mu$  processed elsewhere as in (2.3). As mentioned before,  $\delta_q$  denotes the planned transport time from the job preceding job  $q$  to the Multi task cell. Unfortunately this data is not available in the Volvo Aero planning system, so it is estimated to be 20% of the job queue time.

### 3 The engineer's mathematical model

#### 3.1 Variables

$$z_{ijk} = \begin{cases} 1, & \text{if operation } (i, j) \text{ is allocated to resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$$y_{ijpqk} = \begin{cases} 1, & \text{if op } (i, j) \text{ is being processed before op } (p, q) \text{ on resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$t_{ij}$  = the starting time of operation  $(i, j)$ .

$s_j = t_{n_j, j} + \tau_{n_j, j}$ , the completion time of job  $j$ .

$$h_j = \begin{cases} s_j - d_j, & \text{if } s_j > d_j, \text{ i.e. the tardiness of job } j, \\ 0, & \text{otherwise.} \end{cases}$$

#### 3.2 Objective functions

The objective function, is chosen so that it minimizes the job finish times together with the total tardiness, according to

$$\sum_{j \in \mathcal{J}} (s_j + h_j).$$

If also the time used in the fixure is considered, the objective function becomes

$$\sum_{j \in \mathcal{J}} (s_j - \varepsilon t_{1j} + h_j),$$

where  $\varepsilon \in [0, 1)$ .

#### 3.3 The optimization model

$$\begin{aligned} \text{Minimize} \quad & \sum_{j \in \mathcal{J}} s_j + h_j & (1a) \\ \text{subject to} \quad & \sum_{k \in \mathcal{K}} z_{ijk} = 1, \quad i \in \mathcal{N}_j, j \in \mathcal{J}, & (1b) \\ & z_{ijk} \leq \lambda_{ijk}, \quad i \in \mathcal{N}_j, j \in \mathcal{J}, k \in \mathcal{K}, & (1c) \\ & y_{ijpqk} + y_{pqijk} \leq z_{ijk}, \quad i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), & (1d) \\ & y_{ijpqk} + y_{pqijk} + 1 \geq z_{ijk} + z_{pqk}, \quad i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), & (1e) \\ & t_{ij} + \tau_{ij} - M(1 - y_{ijpqk}) \leq t_{pq}, \quad i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), & (1f) \\ & t_{ij} + \tau_{ij} + w \leq t_{i+1, j}, \quad i \in \mathcal{N}_j \setminus \{n_j\}, j \in \mathcal{J}, & (1g) \\ & t_{1j} \geq r_j, \quad j \in \mathcal{J}, & (1h) \\ & t_{ij} \geq a_k z_{ijk}, \quad j \in \mathcal{J}, k \in \mathcal{K}, & (1i) \\ & t_{1q} \geq s_j + v_{jq}, \quad (j, q) \in \mathcal{Q}, & (1j) \\ & s_j = t_{n_j, j} + \tau_{n_j, j}, \quad j \in \mathcal{J}, & (1k) \\ & h_j \geq s_j - d_j, \quad j \in \mathcal{J}, & (1l) \\ & h_j \geq 0, \quad j \in \mathcal{J}, & (1m) \\ & t_{ij} \geq 0, \quad i \in \mathcal{N}_j, j \in \mathcal{J}, & (1n) \\ & z_{ijk} \in \{0, 1\}, \quad i \in \mathcal{N}_j, j \in \mathcal{J}, k \in \mathcal{K}, & (1o) \\ & y_{ijpqk} \in \{0, 1\}, \quad i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), & (1p) \end{aligned}$$

where (1b) ensures that every operation is processed exactly once, and (1c) makes sure that each operation is scheduled on a resource allowed for that operation. The constraints (1d) and (1e) determine an ordering of the operations that are processed on the same resource. The constraints (1d) make sure that at most one of the variables  $y_{ijpqk}$  and  $y_{pqijk}$  may attain the value 1, and the constraints (1e) regulates that at least one of the variables  $y_{ijpqk}$  and  $y_{pqijk}$  must have the value 1 if operations  $(i, j)$  and  $(p, q)$  are to be performed on the same resource.

The constraint (1f) makes sure that the starting time of operation  $(p, q)$  is scheduled after the completion of the previous operation on the same resource. Generally, in scheduling problems, the symmetry preventing constraint  $t_{pq} + \tau_{pq} - My_{ijpqk} \leq t_{ij}$  is required, but this becomes redundant here since the variables  $y_{ijpqk}$  and  $y_{pqijk}$  are regulated by the inequalities (1d) and (1e). The constraints (1g) ensure that the operations within the same job,  $j$ , are scheduled in the right order and that each operation starts after the previous operation is completed and the goods is transported to the next resource.

Equation (1h) regulates the starting times of the first operation of every job, so that no job is scheduled before its release date. Equation (1i) makes sure that no operation is scheduled on resource  $k$  before this resource is available for the first time. The constraints (1k) through (1m) calculate the finish times and the tardiness for the objective function. The constraints (1n) are the nonnegativity constraints of the starting times. This is redundant due to equations (1g) through (1i) provided that  $r_j$  is nonnegative for all  $j$ . The constraints (1o) and (1p) are the binary constraints on the variables.

## 4 Decomposition into a machining and a feasibility problem

### 4.1 Variables

$$z_{jk} = \begin{cases} 1, & \text{if job } j \text{ is allocated to resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$$y_{jqk} = \begin{cases} 1, & \text{if job } j \text{ is being processed before job } q \text{ on resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$t_j$  = the starting time of the machining operation of job  $j$ .

$s_j = t_j + p_j^m + p_j^{\text{pm}}$ , the completion time of job  $j$ ,

where  $p_j^{\text{pm}}$  is the sum of the post-machining route operations.

$$h_j = \begin{cases} s_j - d_j, & \text{if } s_j > d_j, \text{ i.e. the tardiness of job } j, \\ 0, & \text{otherwise.} \end{cases}$$

### 4.2 The machining problem

$$\begin{aligned} & \text{Minimize} && \sum_{j \in \mathcal{J}} (s_j^m + h_j^m) \\ & \text{subject to} && \sum_{k \in \mathcal{K}} z_{jk}^m = 1, && j \in \mathcal{J}, \\ & && z_{jk}^m \leq \lambda_{jk}^m, && j \in \mathcal{J}, k \in \mathcal{K}, \end{aligned}$$

$$\begin{aligned}
y_{jqk}^m + y_{qjk}^m &\leq z_{jk}^m, & j, q \in \mathcal{J}, k \in \mathcal{K}, j \neq q, \\
y_{jqk}^m + y_{qjk}^m + 1 &\geq z_{jk}^m + z_{qk}^m, & j, q \in \mathcal{J}, k \in \mathcal{K}, j \neq q, \\
t_j^m + p_j^m - M(1 - y_{jqk}^m) &\leq t_q^m, & j, q \in \mathcal{J}, k \in \mathcal{K}, j \neq q, \\
t_j^m &\geq r_j^m, & j \in \mathcal{J}, \\
t_j^m &\geq a_k z_{jk}^m, & j \in \mathcal{J}, \\
t_q^m &\geq s_j^m + v_{jq}^m, & (j, q) \in \mathcal{Q}, \\
s_j^m &= t_j^m + p_j^m + p_j^{pm}, & j \in \mathcal{J}, \\
h_j^m &\geq s_j^m - d_j^m, & j \in \mathcal{J}, \\
h_j^m &\geq 0, & j \in \mathcal{J}, \\
t_j^m &\geq 0, & j \in \mathcal{J}, \\
z_{jk}^m &\in \{0, 1\}, & j \in \mathcal{J}, k \in \mathcal{K}, \\
y_{jqk}^m &\in \{0, 1\}, & j, q \in \mathcal{J}, k \in \mathcal{K}, j \neq q,
\end{aligned}$$

where  $v_{jq}^m = v_{jq} + t_{1q}$  and  $p_j^{pm} = \sum_{i=3}^{n_j} p_{ij}$ .

### 4.3 The feasibility problem

The aim of the feasibility problem is to make good feasible schedules of the remaining resources of the Multi Task cell, i.e. the three setup and the two deburring stations. The objective function consists of three terms. The first term is minimizing the total processing lead time, i.e. the time the fixture for each job is occupied,  $s_j - t_{1j}$ . The next term is minimizing the total tardiness and the last term is a weight  $\omega_k$  times the variable  $z_{ijk}$ , in order to avoid large computation times by symmetric solutions for the three setup stations.

$$\begin{aligned}
\text{Minimize} \quad & \sum_{j \in \mathcal{J}} \left( s_j - 0.001 t_{1j} + h_j + \sum_{i \in \mathcal{N}_j} \sum_{k \in \mathcal{K}} \omega_k z_{ijk} \right) \\
\text{subject to} \quad & \sum_{k \in \mathcal{K}} z_{ijk} = 1, & i \in \mathcal{N}_j, j \in \mathcal{J}, \\
& z_{ijk} \leq \lambda_{ijk}, & i \in \mathcal{N}_j, j \in \mathcal{J}, k \in \mathcal{K}, \\
& y_{ijpqk} + y_{pqijk} \leq z_{ijk}, & i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), \\
& y_{ijpqk} + y_{pqijk} + 1 \geq z_{ijk} + z_{pqk}, & i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), \\
& t_{ij} + p_{ij} - M(1 - y_{ijpqk}) \leq t_{pq}, & i \in \mathcal{N}_j, p \in \mathcal{N}_q, j, q \in \mathcal{J}, k \in \mathcal{K}, (i, j) \neq (p, q), \\
& t_{ij} + p_{ij} + w \leq t_{i+1, j}, & i \in \mathcal{N}_j \setminus \{n_j\}, j \in \mathcal{J}, \\
& t_{1j} \geq r_j, & j \in \mathcal{J}, \\
& t_{ij} \geq a_k z_{ijk}, & j \in \mathcal{J}, k \in \mathcal{K}, \\
& t_{1q} \geq s_j + v_{jq}, & (j, q) \in \mathcal{Q}, \\
& s_j = t_{n, j} + p_{n, j}, & j \in \mathcal{J}, \\
& h_j \geq s_j - d_j, & j \in \mathcal{J}, \\
& h_j \geq 0, & j \in \mathcal{J}, \\
& t_{ij} \geq 0, & i \in \mathcal{N}_j, j \in \mathcal{J}, \\
& y_{2j2qk} = y_{jqk}^m, & j, q \in \mathcal{J}, k \in \mathcal{K}, \\
& z_{2jk} = z_{jk}^m, & j \in \mathcal{J}, k \in \mathcal{K}, \\
& z_{ijk} \in \{0, 1\}, & i \in \mathcal{N}_j, j \in \mathcal{J}, k \in \mathcal{K},
\end{aligned}$$

$$y_{ijpqk} \in \{0, 1\}, \quad i \in \mathcal{N}_j, \quad p \in \mathcal{N}_q, \quad j, q \in \mathcal{J}, \quad k \in \mathcal{K}, \quad (i, j) \neq (p, q),$$

where  $y_{jqk}^m$  and  $z_{jk}^m$  are the solutions obtained from the machining problem.

## 5 The time-discrete machining model

### 5.1 Time intervals

The time horizon of the schedule is divided into  $T + 1$  time intervals, and the index  $u \in \mathcal{T} = \{0, 1, \dots, T\}$  denotes a time interval starting at  $u$  and ending at  $u + 1$  with length  $\ell$ .



### 5.2 Variables

$$x_{jku} = \begin{cases} 1, & \text{if job } j \text{ is to start at the beginning of time interval } u \text{ on} \\ & \text{resource } k, \\ 0, & \text{otherwise.} \end{cases}$$

$s_j$  = the completion time of job  $j$ .

$$h_j = \begin{cases} s_j + p_j^{\text{pm}} - d_j, & \text{if } s_j + p_j^{\text{pm}} > d_j, \text{ i.e. the tardiness of job } j, \\ 0, & \text{otherwise.} \end{cases}$$

### 5.3 The time-discrete optimization model

$$\begin{aligned} & \text{Minimize} && \sum_{j \in \mathcal{J}} (s_j + h_j), \\ & \text{subject to} && \sum_{k \in \mathcal{K}} \sum_{u \in \mathcal{T}} x_{jku} = 1, \quad j \in \mathcal{J}, \\ & && \sum_{u \in \mathcal{T}} x_{jku} \leq \lambda_{jk}, \quad j \in \mathcal{J}, \quad k \in \mathcal{K}, \\ & && x_{jku} + x_{q,k,u+\mu} \leq 1, \quad j, q \in \mathcal{J}, \quad j \neq q, \quad k \in \mathcal{K}, \quad u = 0, \dots, T - (p_j - 1), \quad \mu = 0, \dots, p_j - 1, \\ & && \sum_{k \in \mathcal{K}} (x_{jku} + \sum_{u_{\text{ext}}=0}^{u+v_{jq}^{\text{ext}}-1} x_{qku_{\text{ext}}}) \leq 1, \quad (j, q) \in \mathcal{Q}, \quad u = 0, \dots, T - (v_{jq}^{\text{ext}} - 1), \\ & && x_{jku} = 0, \quad (j, q) \in \mathcal{Q}, \quad k \in \mathcal{K}, \quad u = T - v_{jq}^{\text{ext}}, \dots, T, \\ & && x_{jku} = 0, \quad j \in \mathcal{J}, \quad k \in \mathcal{K}, \quad u = 0, 1, \dots, r_j, \\ & && x_{jku} = 0, \quad j \in \mathcal{J}, \quad k \in \mathcal{K}, \quad u = 0, 1, \dots, a_k, \\ & && \sum_{k \in \mathcal{K}} \sum_{u \in \mathcal{T}} u x_{jku} + p_j + p_j^{\text{pm}} = s_j, \quad j \in \mathcal{J}, \\ & && h_j \geq s_j - d_j, \quad j \in \mathcal{J}, \\ & && h_j \geq 0, \quad j \in \mathcal{J}, \\ & && x_{jku} \in \{0, 1\}, \quad j \in \mathcal{J}, \quad k \in \mathcal{K}, \quad u \in \mathcal{T}, \end{aligned}$$

where  $v_{jq}^{\text{ext}} = p_j + p_j^{\text{pm}} + v_{jq}$ .