

Dynamic Knapsack Problem

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1 Questions/Comments

1.1 Questions

- Should *ul* definition include units violated?? it seems a constraint might be violated otherwise

1.2 Comments

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2 Assumptions

- Assume all new patients are arriving at the beginning of the period
 - Assume arrivals follow poisson distribution
- It is assumed that patients do not change complexities within some limit (TL).
 - Let's assume our limit is 10 periods, patients will never change complexities within the 10 periods.
 - However, if a patient waits for 11 periods there is a possibility they increase complexity
- We assume patient complexity transitions follow binary distribution - each period a patient can become more complex with a certain probability
 - While patient is waiting less than TL , the transition probability is 0.
 - After TL , transition probability is some arbitrary value
- It is assumed if a patient is scheduled in period one, it means they are served "immediately", regardless if in practice the appointment is at the beginning or end of the period.
- We distinguish between two types of reschedules: good and bad reschedules
 - Good reschedules are reschedules where a patient is rescheduled to an earlier period
 - Bad reschedules are reschedules where a patient is rescheduled to a later period
- We assume that only specific reschedules are allowed (to simplify the model, and remove redundancies)
 - Good reschedules are allowed from any period after 2 into period 1
 - Bad reschedules are only allowed from period 1 to any period after 2
- It is assumed that there is a certain default expected number of PPE units available for all periods
 - However, in the period 1, there is some random deviation from the expected number of units
 - This random deviation follows some uniform distribution
- We allow some violation of PPE units, in rare cases (with high cost), in order to accommodate changes due to variability.
 - If we have violation of PPE units - the extra capacity comes from external source

3 MDP Model

3.1 Decision Epochs

Decisions are made at the beginning of each time period. There are 3 relevant time intervals to look at:

- Pre-decision state (S)
 - Pre-decision state defines the initial state on which a decision is required
 - This information drives decision making
- Post-decision state (\hat{S})

- Post-decision state is the time when the decision has been executed, but no new info came in
- This state defines the immediate cost of an action
- Post-transition state (S')
 - This is the state when new information has come in (transition randomness). Defines next pre decision state
 - It is primarily used to generate expectation for the ADP

3.2 State Space

3.2.1 Description

State is defined by current available and used resources, patient waitlist/demand, and patients already scheduled

$$\vec{S} = (\vec{ul}, p\vec{w}, p\vec{s})$$

- $\vec{ul} = ul_p$ - Units left over from previous period for resource p plus deviation for that unit for period 1. This is only for resources that can be carried over
- $p\vec{w} = pw_{mdc}$ - Patients of complexity d , CPU c , on a wait list for m periods (m of 0 - just arrived)
- $p\vec{s} = ps_{tmdc}$ - Patients of complexity d , CPU c , scheduled to period t , waiting for m periods (m of 0 - just arrived)

3.3 Action Sets

3.3.1 Description

Decision consists of rescheduling currently scheduled patients, and scheduling patients on waitlist. There are also some goal and auxiliary variables

$$\vec{A} = (\vec{sc}, r\vec{sc}, \vec{uv}, \text{auxiliary variables})$$

- $\vec{sc} = sc_{tmdc}$ - Patients of complexity d , CPU c , waiting for m periods, to schedule in period t
- $r\vec{sc} = rsc_{tt'mdc}$ - Patients of complexity d , CPU c , waiting for m periods, to reschedule from period t to period t'
- $\vec{uv} = uv_{tp}$ - goal variable, violation on number of resources used for period t , of resource p

3.3.2 Auxiliary Variables

- uvb_{tp} - binary variable to enforce uv variable without objective function
- \hat{ul}_p - post-decision unit leftover at period 1
- ulb_p - binary variable to enforce ul variable without objective function
- $\hat{u}u_{tp}$ - post-decision units used
- $\hat{p}w_{mdc}$ - post-decision patients waiting
- $\hat{p}s_{tmdc}$ - post-decision patients scheduled

3.3.3 Auxiliary Variable Definition

$$\hat{u}u_{tp} = \sum_{mdc} U_{pdc} \hat{p}s_{tmdc} \quad \forall tp \quad (1)$$

$$\hat{p}w_{mdc} = pw_{mdc} - \sum_t sc_{tmdc} \quad \forall mdc \quad (2)$$

$$\hat{p}s_{tmdc} = ps_{tmdc} + sc_{tmdc} + \sum_t rsc_{tt'mdc} - \sum_{t'} rsc_{tt'mdc} \quad \forall tmdc \quad (3)$$

- Define Resource Violation Variable

$$\hat{u}v_{tp} \leq M(uvb_{tp}) \quad \forall tp \quad (4)$$

$$\hat{u}v_{1p} \leq (\hat{u}u_{1p} - uen_{1p} - ul_p) + M(1 - uvb_{tp}) \quad \forall p \in \{\text{Carry Over}\} \quad (5)$$

$$\hat{u}v_{1p} \leq (\hat{u}u_{1p} - uen_{1p}) + M(1 - uvb_{tp}) \quad \forall p \in \{\text{Non Carry Over}\} \quad (6)$$

$$\hat{u}v_{tp} \leq (\hat{u}u_{tp} - uen_{tp}) + M(1 - uvb_{tp}) \quad \forall t \in \{2..T\}p \quad (7)$$

$$(8)$$

- Define Units Left Over Variable

$$\hat{u}l_p \geq 0 \quad \forall p \in \{\text{Non Carry Over}\} \quad (9)$$

$$\hat{u}l_p \geq uen_{1p} + ul_p - \hat{u}u_{1p} \quad \forall p \in \{\text{Non Carry Over}\} \quad (10)$$

$$\hat{u}l_p \leq M(ulb_p) \quad \forall p \in \{\text{Non Carry Over}\} \quad (11)$$

$$\hat{u}l_p \leq (uen_{1p} + ul_p - \hat{u}u_{1p}) + M(1 - ulb_p) \quad \forall p \in \{\text{Non Carry Over}\} \quad (12)$$

3.3.4 State-Action Constraints

- Resource Usage Constraint

$$\hat{u}u_{1p} \leq uen_{1p} + ul_p + uv_{1p} \quad \forall p \quad (13)$$

$$\hat{u}u_{tp} \leq uen_{tp} + uv_{tp} \quad \forall t \in \{2..T\}p \quad (14)$$

- Custom bounds on when reschedules are allowed

$$rsc_{tt'mdc} = 0 \quad \forall t \in \{2..T\}, t' \in \{2..T\}mdc \quad (15)$$

$$rsc_{tt'mdc} = 0 \quad \forall tt'mdc, \text{ where } t = t' = 1 \quad (16)$$

- Bounds on Schedules/Reschedules

$$\sum_{t'} rsc_{tt'mdc} \leq pst_{mdc} \quad \forall tmdc \quad (17)$$

$$\sum_t sc_{tmdc} \leq pw_{mdc} \quad \forall mdc \quad (18)$$

- Bounds on states

$$ul_p \leq uen_p * 3 \quad \forall p \quad (19)$$

$$pw_{mdc} \leq pea_{dc} * 20 \quad \forall mdc \quad (20)$$

$$pst_{mdc} \leq pea_{dc} * 4 \quad \forall tmdc \quad (21)$$

3.4 Transition Probabilities

3.4.1 Uncertainty Sources

1. Number of patients arriving this period - pw_{0dc}

- pea_{dc} - represents the random number of patients arriving. Follows a poisson distribution.

2. Transition between patient complexities within the wait list - $pw_{m>=1,dc}$

- pwt_{mdc} represents the random number of patients of complexity d , CPU c , waiting for m period, who became more complex. Follows binomial distribution.
- ptp_{mdc} represents transition probability to a higher complexity

3. Transition between patient complexities within the scheduled list - pst_{mdc}

- pst_{mdc} represents the random number of patients of complexity d , CPU c , waiting for m period, scheduled into t period, who became more complex. Follows binomial distribution.

4. Amount of expected units of PPE resource for the next time period - ue_{1p}

- ued_p represents the random deviation of resource p from the expectation. Follows uniform distribution.
- uen_p expected units of resource p per period

3.4.2 Transition Constraints

1. Transition from $u\vec{e}$ to $u\vec{e}'$ - Resource Carry Over

$$ul'_p = \hat{ul}_p + ued_p \quad \forall p \in \{\text{Carry Over}\} \quad (22)$$

$$(23)$$

2. Transition from $p\vec{w}$ to $p\vec{w}'$ - Flow of patients on waitlist

$$pw'_{0dc} = pea_{dc} \quad \forall dc \quad (24)$$

$$pw'_{mdc} = \hat{pw}_{m-1,dc} + \overbrace{pwt_{m-1,d-1,c} - pwt_{m-1,dc}}^{\text{change in complexities}} \quad \forall m \in \{1 \dots M-1\}dc \quad (25)$$

$$pw'_{Mdc} = \sum_{M-1}^M (\hat{pw}_{mdc} + \overbrace{pwt_{md-1,c} - pwt_{mdc}}^{\text{change in complexities}}) \quad \forall dc \quad (26)$$

3. Transition from $p\vec{s}$ to $p\vec{s}'$ - Flow of patients scheduled

$$ps'_{t0dc} = 0 \quad \forall tdc \quad (27)$$

$$ps'_{Tmdc} = 0 \quad \forall mdc \quad (28)$$

$$ps'_{tmdc} = \hat{ps}_{t+1,m-1,dc} + \overbrace{pst_{t+1,m-1,d-1,c} - pst_{t+1,m-1,dc}}^{\text{change in complexities}} \quad \forall t \in \{1 \dots T-1\}m \in \{1 \dots M-1\}dc \quad (29)$$

$$ps'_{tMdc} = \sum_{M-1}^M (\hat{ps}_{t+1mdc} + \overbrace{pst_{t+1,md-1,c} - pst_{t+1,mdc}}^{\text{change in complexities}}) \quad \forall t \in \{1 \dots T-1\}dc \quad (30)$$

3.5 Costs

Cost will come from 4 sources:

- waiting (cw)
- rescheduling (cc)
- scheduling (cs)
- resource violation (M)

$$\begin{aligned} C = c(p\vec{w}, p\vec{s}, r\vec{sc}, u\vec{v}) = & \overbrace{\sum_{mdc} cw^{m+1}(\hat{pw}_{mdc})}^{\text{cost of waiting}} + \overbrace{\sum_{tmdc} cw^{M+1}(\hat{ps}_{tMdc})}^{\text{prevent infinite reschedules}} + \\ & \overbrace{\sum_{tmdc} cs^t(sc_{tmdc})}^{\text{Prefer earlier appointments}} \\ & + \overbrace{(1.5cc) \sum_{\substack{tt'mdc \\ t' > t}} rsc_{tt'mdc}}^{\text{Bad Reschedule}} - \overbrace{(0.5cc) \sum_{\substack{tt'mdc \\ t' < t}} rsc_{tt'mdc}}^{\text{Good Reschedule}} + M \sum_{tp} uv_{tp} \end{aligned} \quad (31)$$

4 LP Model

4.1 Full LP

Given a full MDP model, the equivalent LP would look as follows:

$$\max_{\vec{v}} \sum \alpha(\vec{S})v(\vec{S}) \quad (32)$$

subject to

$$c(p\vec{w}, p\vec{s}, r\vec{sc}, u\vec{v}) + \gamma \sum_{\vec{p}} p(p\vec{ea}, p\vec{wt}, p\vec{st}, u\vec{ed})v(\vec{S}'|\vec{S}, \vec{A}, p\vec{ea}, p\vec{wt}, p\vec{st}, u\vec{ed}) \geq v(\vec{S}) \quad \forall \vec{S}, \vec{A} \quad (33)$$

5 Aproximate Dynamic Programming Model

5.1 Full ADP Model

Let's convert it into ADP. We do that by changing $v(\vec{S})$ to an approximation as follows:

$$v(\vec{ul}, \vec{pw}, \vec{ps}) = \beta^0 + \sum_p \beta_p^{ul} ul_p + \sum_{mdc} \beta_{mdc}^{pw} pw_{mdc} + \sum_{tmdc} \beta_{tmdc}^{ps} ps_{tmdc} \quad (34)$$

This gives the following LP

$$\max_{\vec{\beta}} \sum \alpha(\vec{ul}, \vec{pw}, \vec{ps}) v(\vec{ul}, \vec{pw}, \vec{ps}) \quad (35)$$

Subject to:

$$\begin{aligned} & c(\vec{pw}, \vec{ps}, r\vec{sc}, \vec{uv}) + \gamma \sum_{\vec{p}} p \left(\beta^0 + \left(\beta_p^{ul} (\hat{ul}_p + uen_d) \right) + \right. \\ & \left(\sum_{dc} \beta_{0dc}^{pw} pea_{dc} \right) + \left(\sum_{m=1,dc}^{M-1} \beta_{mdc}^{pw} (\hat{pw}_{m-1dc} + pwt_{m-1d-1c} - pwt_{m-1dc}) \right) + \\ & \left. \left(\sum_{dc} \beta_{Mdc}^{pw} \left(\sum_{M-1}^M \hat{pw}_{mdc} + pwt_{md-1c} - pwt_{mdc} \right) \right) + \right. \end{aligned} \quad (36)$$

$$\begin{aligned} & \left(\sum_{tdc} \beta_{t0dc}^{ps} 0 \right) + \left(\sum_{mdc} \beta_{Tmdc}^{ps} 0 \right) + \left(\sum_{t=1m=1dc}^{T-1M-1} \beta_{tmdc}^{ps} (\hat{ps}_{t+1m-1dc} + pst_{t+1m-1d-1c} - pst_{t+1m-1dc}) \right) + \\ & \left. \left(\sum_{t=1Mdc}^{T-1} \beta_{tmdc}^{ps} \left(\sum_{M-1}^M (\hat{ps}_{t+1mdc} + pst_{t+1md-1c} - pst_{t+1mdc}) \right) \right) \right) \geq v(\vec{ul}, \vec{pw}, \vec{ps}) \quad \forall \vec{S} \vec{A} \end{aligned}$$

$$\vec{\beta} \geq 0 \quad \forall \beta \quad (37)$$

And all auxiliary constraints in "Auxiliary Variable Definition"

5.2 Expectation ADP Model

Steps for rearranging and converting the constraint (36) into expectation:

- $E[ued_p] = 0$
- $E[pwt_{mdc}] = ptp_{mdc} * \hat{pw}_{mdc}$
- $E[pst_{tmdc}] = ptp_{mdc} * \hat{ps}_{tmdc}$

$$\beta^0 : \quad (1 - \gamma) \beta^0 \quad (38)$$

$$\beta_p^{ul} : \quad \sum_p \beta_p^{ul} (ul_p - \gamma(\hat{ul}_p)) \quad (39)$$

$$\beta_{mdc}^{pw} : \quad \sum_{dc} \beta_{0dc}^{pw} (pw_{0dc} - \gamma(pea_{dc})) + \sum_{m=1dc}^{M-1} \beta_{mdc}^{pw} (pw_{mdc} - \gamma(\hat{pw}_{m-1dc} + E[pwt_{m-1d-1c}] - E[pwt_{m-1dc}])) + \quad (40)$$

$$\begin{aligned} & \sum_{dc} \beta_{Mdc}^{pw} (pw_{Mdc} - \gamma \sum_{M-1}^M (\hat{pw}_{mdc} + E[pwt_{md-1c}] - E[pwt_{mdc}])) \\ & \beta_{tmdc}^{ps} : \quad \sum_{tdc} \beta_{t0dc}^{ps} (ps_{t0dc}) + \sum_{mdc} \beta_{Tmdc}^{ps} (ps_{Tmdc}) + \\ & \sum_{t=1m=1dc}^{T-1M-1} \beta_{tmdc}^{ps} (pst_{tmdc} - \gamma(\hat{ps}_{t+1m-1dc} + E[pst_{t+1m-1d-1c}] - E[pst_{t+1m-1dc}])) + \\ & \sum_{t=1dc}^{T-1} \beta_{tmdc}^{ps} (pst_{tmdc} - \gamma \sum_{M-1}^M (\hat{ps}_{t+1mdc} + E[pst_{t+1md-1c}] - E[pst_{t+1mdc}])) \end{aligned} \quad (41)$$

Let's say $E[V]$ is the addition of all parts above. Then the ADP model converted to expectation would look as follows:

$$\max_{\vec{\beta}} \left(\sum \beta^0 + \sum \beta^{ut} E[ut] + \sum \beta^{pw} E[pw] + \sum \beta^{ps} E[ps] \right) \quad (42)$$

Subject to:

$$E[V] \leq c(p\vec{w}, \vec{s}\vec{c}, r\vec{s}\vec{c}, \vec{u}\vec{v}) \quad \forall \vec{S}\vec{A} \quad (43)$$

5.3 ADP Master Problem

Converting to Dual

$$\min_w \sum_{\vec{S}\vec{A}} w(\vec{S}\vec{A}) c(\vec{S}\vec{A}) \quad (44)$$

Subject To:

$$\beta^0 : \sum_{\vec{S}\vec{A}} \vec{w}(1 - \gamma) = 1 \quad (45)$$

$$\beta^{ul} : \sum_{\vec{S}\vec{A}} \vec{w} \left(ul_p - \gamma(\hat{u}l_p) \right) \geq E[ul_p] \quad \forall p \quad (46)$$

$$\beta^{pw} : \sum_{\vec{S}\vec{A}} \vec{w} \left(pw_{0dc} - \gamma(pea_{dc}) \right) \geq E[pw_{0dc}] \quad \forall dc \quad (47)$$

$$\sum_{\vec{S}\vec{A}} \vec{w} \left(pw_{mdc} - \gamma(\hat{p}w_{m-1dc} + E[pwt_{m-1d-1c}] - E[pwt_{m-1dc}]) \right) \geq E[pw_{mdc}] \quad \forall m \in \{1 \dots M-1\} dc \quad (48)$$

$$\sum_{\vec{S}\vec{A}} \vec{w} \left(pw_{Mdc} - \gamma \sum_{M-1}^M (\hat{p}w_{mdc} + E[pwt_{md-1c}] - E[pwt_{mdc}]) \right) \geq E[pw_{Mdc}] \quad \forall dc \quad (49)$$

$$\beta^{ps} : \sum_{\vec{S}\vec{A}} \vec{w} \left(ps_{t0dc} \right) \geq E[ps_{t0dc}] \quad \forall tdc \quad (50)$$

$$\sum_{\vec{S}\vec{A}} \vec{w} \left(ps_{Tmdc} \right) \geq E[ps_{Tmdc}] \quad \forall mdc \quad (51)$$

$$\sum_{\vec{S}\vec{A}} \vec{w} \left(ps_{tmdc} - \gamma(\hat{p}s_{t+1m-1dc} + E[ps_{t+1m-1d-1c}] - E[ps_{td}]) \right) \geq E[ps_{tmdc}] \quad \forall \{1 \dots T-1\} \{1 \dots M-1\} dc \quad (52)$$

$$\sum_{\vec{S}\vec{A}} \vec{w} \left(ps_{tMdc} - \gamma \sum_{M-1}^M (\hat{p}s_{t+1mdc} + E[ps_{t+1md-1c}] - E[ps_{td}]) \right) \geq E[ps_{tMdc}] \quad \forall \{1 \dots T-1\} dc \quad (53)$$

$$\text{State Action Bounds:} \quad w \geq 0 \quad \forall w \quad (54)$$

5.4 ADP Pricing Problem

$$\min_{(\vec{u}\vec{e}, \vec{u}\vec{u}, \vec{p}\vec{w}, \vec{p}\vec{e}, \vec{p}\vec{s}) \in S, (\vec{s}\vec{c}, r\vec{s}\vec{c}) \in A} c(\vec{p}\vec{w}, \vec{s}\vec{c}, r\vec{s}\vec{c}, \vec{u}\vec{v}) - E[V] \quad (55)$$

Subject to: constraints in sections "State-Action Constraints" and "Auxiliary Variable Definition"

6 Solution Explanation

6.1 Algorithm for solving

To get $\vec{\beta}$ values, which will be used to generate an action follow steps below:

1. Perform a monte-carlo simulation (following some arbitrary policy) to get $E[ue], E[uu], E[pw], E[pw]$
2. Create an initial feasible set of state-action pairs - \vec{w}
3. Solve model in section "ADP Master Problem" where each state-action pairs in \vec{w} corresponds to a variable and parameters for all the constraints for a specific action.

4. Solve model in section "ADP Pricing Problem", where duals from problem in step 3 correspond to $\vec{\beta}$ values.
 - If objective function is less than 0, add solution as a single state-action pair to \vec{w} and go to step 3
 - If objective function is greater than 0, continue to next step
5. Duals from problem in step 3 correspond to final $\vec{\beta}$ values

6.2 Generating an Action

Once $\vec{\beta}$ values have been approximated, you may use the model below to generate a recommended action for a specific state.

$$\begin{aligned}
 & \min_{\vec{A}} c(\vec{S}, \vec{A}) + \gamma v(S') \\
 v(S') = & \left(\beta^0 + \left(\sum_p \beta_{1p}^{ue} (uen_d + ue_{1p} - \hat{u}u_{1p} + uv_{1p}) \right) + \left(\sum_{t=2,p}^T \beta_{tp}^{ue} uen_d \right) + \right. \\
 & \left(\sum_{tp}^{T-1} \beta_{tp}^{uu} (\hat{u}u_{t+1,p} + \sum_{mdc} E[pst_{tmdc}](U_{pd+1c} - U_{pdc})) \right) + \\
 & \left(\sum_{dc} \beta_{0dc}^{pw} pea_{dc} \right) + \left(\sum_{m=1,dc}^{M-1} \beta_{mdc}^{pw} (\hat{p}w_{m-1dc} + E[pwt_{m-1d-1c}] - E[pwt_{m-1dc}]) \right) + \\
 & \left(\sum_{dc} \beta_{Mdc}^{pw} \left(\sum_{M-1}^M \hat{p}w_{mdc} + E[pwt_{md-1c}] - E[pwt_{mdc}] \right) \right) + \\
 & \left(\sum_{t=1m=1dc}^{T-1M-1} \beta_{tmdc}^{ps} (\hat{p}s_{t+1m-1dc} + E[pst_{t+1m-1d-1c}] - E[pst_{t+1m-1dc}]) \right) + \\
 & \left. \left(\sum_{t=1Mdc}^{T-1} \beta_{tmdc}^{ps} \left(\sum_{M-1}^M (\hat{p}s_{t+1mdc} + E[pst_{t+1md-1c}] - E[pst_{t+1mdc}]) \right) \right) \right)
 \end{aligned} \tag{56}$$

Subject to: constraints in section "State-Action Constraints"