

TA Assignment Problem

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1 Preamble

For this problem we decided on OR-Tree based search and Set based search. Our reasoning was as follows. We immediately had a solution using an OR-Tree based model that covered every combination of TA assignments to time slots. We figured using a greedy heuristic for our *altern* function we could likely come up with an optimal or slightly sub optimal solution with relative ease. Our second choice as set based search was very natural in that set based search allows for so much flexibility. On top of this, the F_{werth} function of set based search parallels with the soft(and hard) constraints detailed in the problem description. We were uninspired when we considered the option of using an AND-Tree based search model simply because it's difficult to apply a *div* function to the structure of this problem. Finally the main reason for selecting these search paradigms was professor kremer confirming our instincts and telling us we made the right choice.

2 OR-Tree Based Search

Let L be the set of all tutorials or labs and T be set of all TA's. Then the problem is defined as follows:

1. $\forall x \in L, \exists y \in T$ such that $\text{instructs}(y, l)$.
2. All hard constraints are satisfied¹

2.1 Explanation of Model

In the input we're given the following a set of time slots². These time slots consist of a set of days MWF, MW, TR and a time in 24 hour format. Additionally were given a set of TA's that have their own schedule that must be adhered to. We're looking for the existence of as assignment(and if it exists, the optimal solution) such that all time slots are filled with available TA's. Enter OR-Tree based search.

The algorithm works as follows

1. Start with the empty set
2. Pick a TA with the greatest number of constraints to follow.³
3. For each time slot that the chosen TA is available to teach - add a branch to the root node. At the end of each branch are the nodes with a tuple that explains:

¹See appendix

²Time slots will be denoted $TS = (D, T)$ where $D \in \{MWF, MW, TR\}$

³This helps us not get forced into a corner further down the tree

- (a) the goodness value of this pair
 - (b) the time slot
 - (c) the TA
4. Sort the branches from the root node by goodness value in ascending order⁴
 5. Set the left hand node to be your root node and recurse
 6. When there are no more TA's available the left hand arm of the is a solution
 7. While there is still time available for processing - Go back to the root of the tree and this time choose 2nd child of the $root = \emptyset$

2.2 Mathematical Description of the Model

Let L be the set of labs

Let L' be the set of labs without a TA assignment yet

Let T be the set of all TA's

Let $I : T \times L$ be a set of ordered pairs representing an assignment of a TA to a lab.

2.2.1 Defining the problem instance

We defined our problem instance as follows

$$pr = \langle (l_1, t_1), (l_2, t_2), \dots, (l_n, t_n) \rangle$$

Where $(l_i, t_i) \in I$ and $pr : Prob = sequence\ I$

States can be described by

$$S = OTree(Prob, \{yes, no, ?\}, b_1 \dots b_n)$$

2.2.2 Evaluating Solutions

We define the function $f_{leaf} : S \times Env \rightarrow \mathbb{N}$

$$f(s) = \sum_{t \in T} Cv(\langle (l_1, t_1) \dots (l_n, t_n) \rangle)$$

Where $Cv : Prob \rightarrow \mathbb{N}$ is a function that takes in a sequence of assignments and returns the sum of constraints being violated.

⁴The largest goodness value possible is 0

2.2.3 When is a branch solvable or unsolvable

A pr is unsolvable when

$$Erw_{v,wt}((pr, ?), (pr, no)) \iff (l, t) \in I \wedge violates - hard - constraint((l, t))$$

I.e. for every assignment of a TA to a specific lab, a hard constraint is violated.

A pr is solved when

$$Erw_{v,wt}((pr, ?), (pr, yes)) \iff \neg Erw_{v,wt}((pr, ?), (pr, no)) \wedge |L'| = \emptyset$$

I.e. the problem is not unsolvable and all the labs have been assigned a TA.

2.2.4 Branching Definition

Define *Altern* in the following way

$$Altern(s, s') \text{ such that } (|s| + 1 = |s'|) \wedge getLab(last(s')) \in L' \wedge getTA(last(s')) \in T$$

Where $getLab() : I \times L \rightarrow L$ returns the lab from a tuple in I . Similarly for $getTA()$

I.e. the sequence s' is longer than s and we have assigned a TA to a lab which hasn't had any assignment prior.

$Erw((pr, ?), (pr, ?, (pr_1, ?), (pr_2, ?), \dots, (pr_n, ?)))$ if pr_i is generated out of assigning a valid TA to pr_i such that this assignment doesn't violate any hard constants

2.3 Demonstration of a small search instance

3 Set-Based Search

3.1 Explanation of Model

The Set based search algorithm works as follows.

1. Start with an initial state S_0 of approximately fifty facts. This can be generate via random walks of our OR-Tree based tree
2. Take the best fact in S_0 and set it to max
3. Modify each fact by swapping TA assignments such that it enhances or doesn't change the f_{werth} value of functions
4. Update the value of max
5. While there is still time left, go back to step 3

3.2 Mathematical Description of the Model

3.2.1 Model

- Let L be the set of labs
- Let T be the set of all TA's
- Let $I : T \times L$ be a set of ordered pairs representing an assignment of a TA to a lab.
- Let a fact $f : \{f' | f' \in I\}$ such that no hard constraints are violated
- Let F be a set of facts
- Let a state $S \subseteq 2^F$
- Let $A = (S, T)$
- Let $T : S \times S$ and $T = \{(s, s') | \exists A \rightarrow B \in Ext \wedge A \subseteq s \wedge s' = (s - A) \cup B\}$

Define $Ext : \{A \rightarrow B | A, B \subseteq F\}$ where $B = A \cup C$ where C is generated by specifying an allowable time and calling *Generate* then *Combine* until that time is exceeded or until $|C| = |A|$ so that $|B| = 2|A|$. The operations used are defined as:

1. *Generate* - Do a random walk through the defined in ???. The random walk does not compare leafs, it only tries paths at random until a solution is found. If no solution is found within the allowed time, this operation fails.
2. *Combine* - First, map each element in a fact f from a 2-tuple to a 3-tuple $(t, l) \rightarrow (t, l, b = time(l))$. Then, for each b such that $\exists (t', l', b) \in f$, match each instance of t' and l' once at random. This does not change the times that any TA teaches, it only changes which labs a TA is teaching. If the result violates any hard constraints, this operation fails. For the implementation, we will consider lazy evaluation of hard constraints.

3.2.2 Process

We define our process $P : (A, Env, K)$ for the set based search. The model A has already been defined. It is assumed that the environment Env is unchanging so $K : S \times Env \rightarrow S$ is just $K : S \rightarrow S$. The control K is ??? from rubric ???.

f_{wert} is defined as $-\sum_i penalty_i(f)$ where $penalty_i$ is defined in Table ?? as a function of a fact which is either the penalty value from the table or zero if the penalty does not apply.

f_{select} is defined as a tournament. The number of facts is “culled” down to a specified number N . This is done by repeating the following operation $|A| - N$ times. At random, two facts in $f_1, f_2 \in F$ are selected. A random number $0 < r < 1$ is generated. If $r < \frac{f_{1wert}}{f_{1wert} + f_{2wert}}$, f_1 is removed from A , otherwise f_2 is removed from A . For the implementation,

3.3 Demonstration of a small search instance

Appendices

Hard Constraints

1. every TA is assigned at most MAX_LABS labs
FORALL ta:TA . lab-count(ta) \leq MAX_LABS)
2. every TA is assigned at least MIN_LABS labs (if the TA *has* a lab assignment)
FORALL ta:TA . lab-count(ta) \neq 0 then lab-count(ta) \geq MIN_LABS)
3. no lab has more than one TA assigned to it
FORALL course:Course, lab:Lab | has-lab(course,?,lab) . \neg EXISTS ta1,ta2:TA
| ta1 \neq ta2 . instructs(ta1,lab) \wedge instructs(ta2,lab)
4. every lab has a TA assigned to it
FORALL course:Course, lab:Lab | has-lab(course,?,lab). EXISTS ta:TA .
instructs(ta,course,lab)
5. no TA is assigned simultaneous labs
FORALL ta:TA, c1,c2:Course, b1,b2:Lab | (c1=c2 \Rightarrow b1 \neq b2) \wedge instructs(ta,c1,b1) \wedge instructs(ta,c2,b2) . \neg EXISTS t1,t2 | at(c1,b1,t1) \wedge at(c2,b2,t2) . conflicts(t1,t2)
6. no TA is assigned a lab that conflicts with his/her own courses
FORALL ta:TA, course:Course, lab:Lab | instructs(ta,course,lab) .
((\neg EXISTS c:Course, lec:Lecture | taking(ta,c,lec) . EXISTS t1,t2 |
at(course,lab,t1) \wedge at(c,lec,t2)) . conflicts(t1,t2)) \wedge
((\neg EXISTS c:Course, b:Lab | taking(ta,c,b) . EXISTS t1,t2 | at(course,lab,t1)
 \wedge at(c,b,t2)) . conflicts(t1,t2)))

where:

lab-count(TA) is a function that returns the number of labs a TA instructs.