

**SCHOOL OF COMPUTER SCIENCE  
COURSEWORK ASSESSMENT PROFORMA**

**MODULE & LECTURER:** CM3109, Richard Booth

**DATE SET:** 26/10/2016

**SUBMISSION DATE:** 02/12/2016 (23:59)

**SUBMISSION ARRANGEMENTS:** Submit files to Learning Central: a pdf or Word file (containing answers to tasks 1 and 3) and source code for task 2.

**TITLE:** Combinatorial Optimisation Coursework

This coursework is worth 30% of the total marks available for this module. The penalty for late or non-submission is an award of zero marks. You are reminded of the need to comply with Cardiff University's Student Guide to Academic Integrity. Your work should be submitted using the official Coursework Submission Cover sheet.

**INSTRUCTIONS**

You are required to complete 3 tasks about implementing a simulated annealing algorithm, as described in detail in the attachment. The answers to task 1 and task 3 should be submitted as a single pdf or Word file. The answer to task 2 should be submitted as source code in the programming language you choose to use.

**SUBMISSION INSTRUCTIONS**

Description		Type	Name
Cover sheet	<b>Compulsory</b>	One PDF (.pdf) file	[student number].pdf
Task 1 + 3	<b>Compulsory</b>	One PDF (.pdf) or Word file (.doc or .docx)	Text_[student number].pdf/doc/docx
Task 2	<b>Compulsory</b>	One or more source code files in programming language of choice	No restriction

## **CRITERIA FOR ASSESSMENT**

Credit will be awarded against the following criteria.

- [Correctness] Do the answers correctly address the requirements of each task? Does the provided code run correctly in Task 2?
- [Clarity] Are explanations and summaries easily understandable? Is documentation clear? Are comments provided in the code to clarify non-standard code fragments?
- [Understanding of concepts] Do the answers show an understanding of basic optimisation concepts such as neighbourhood.

## **FURTHER DETAILS**

Feedback on your coursework will address the above criteria and will be returned in approximately 3 weeks.

This will be supplemented with oral feedback via individual meetings.

# CM3109 Coursework

## Autumn semester 2016

### Draft\*

#### Notes before starting

- The following programming exercises may be done using a programming language of your own choice, though all parts should be done using the same language. If you wish to use anything other than Java or Python then please let me know well in advance so I can make sure I have everything necessary to run your program installed on my computer.
- If you use any external sources for solving your coursework (e.g. pseudocode from a website), you should add a comment with a reference to these sources and a brief explanation of how they have been used. You are allowed to discuss high-level aspects of your solution with other students, but you should disclose the names of these students in a comment.

The aim of this coursework is to use simulated annealing to solve an optimisation problem coming from the field of *computational social choice* - a hot interdisciplinary topic currently receiving a lot of attention from researchers in several different fields, including AI. Section 1 provides necessary background to the problem, with the specific coursework tasks coming in Section 2.

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\*Subject to minor cosmetic changes after a final check. Final version to be released by the end of week 5.

# 1 Background

We start by introducing the problem we're looking at, then we discuss how to represent the input and outputs to this problem, and then we introduce a particular proposal for approaching this problem which we will aim to implement via simulated annealing.

## The problem

Suppose we have a competition with  $n$  participants, which is decided by playing a *tournament* in which each participant plays one match against each of the others. (These participants could be football teams, chess players, or anything). After the tournament is finished we know both (i) the winner of each match, and (ii) the margin of victory (represented by an integer greater than 0). **To keep things simple we assume none of the matches is tied.** The question is, based on this information, which participant should be judged to be the winner of the tournament, who is the runner-up, who places third, and so on up to the worst? In other words, which *ranking* of the participants (from best to worst) appropriately reflects the relative quality of the participants? (We do not allow ties in the ranking).

## Weighted tournament graphs and rankings

We can picture such a tournament  $T$  as a directed graph whose nodes are the participants, and such that there exists a directed edge *from*  $i$  *to*  $j$  if  $i$  is the winner of the match “ $i$  versus  $j$ ”. Each such edge is labelled with a *weight*  $w(i, j)$ , which is the number representing the margin of victory. One example of such a weighted tournament graph is on the left of Fig. 1. There is an edge from  $A$  to  $C$  labelled with 14, indicating that  $A$  was the very clear winner (by margin 14) of the match “ $A$  versus  $C$ ”.  $A$  also defeated  $B$  by a margin of 5, but was defeated by  $D$  by a margin of 3. These three results would suggest that  $D$  has claims to being declared the best, but  $D$  was itself defeated quite convincingly by  $C$  (by margin of 9), so it's not clear-cut.

Another way to represent a weighted tournament graph (more amenable for representation in a computer) is as an  $n \times n$  matrix  $[a_{ij}]$  whose rows and columns represent the participants  $\{i \mid i = 1, \dots, n\}$  and whose  $(i, j)$  entry  $a_{ij}$  equals 0 if  $i$  was defeated by  $j$ , and gives the weight  $w(i, j)$  if  $i$  defeated  $j$ .

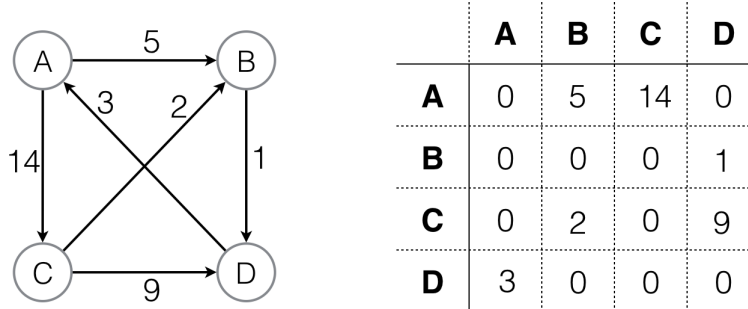


Figure 1: An example tournament graph with  $n = 4$  participants  $A, B, C, D$  on the left, with its matrix representation on the right

For example, the matrix on the right of Fig. 1 is the matrix representation of the tournament graph to its left.

Concerning the output of our problem, i.e., rankings, the easiest way to represent a ranking is as a list  $[i_1, i_2, \dots, i_n]$  with the first entry  $i_1$  representing the best participant, the second  $i_2$  representing the second best, and so on. For instance, continuing the above example, the ranking which places  $D$  first followed by  $A, B$  and  $C$ , in that order, would be represented as  $[D, A, B, C]$ .

## Kemeny rankings

In the above problem, we have to choose the ranking that “best fits” the results from a given weighted tournament  $T$ . Several different proposals have been made for how to precisely formalise what “best fit” means. We are going to focus on just one of them, which is based on the intuition of minimising the *disagreements* with  $T$ . Let’s say a particular ranking  $R$  *disagrees with  $T$  on an edge  $(x, y)$*  if  $x$  defeats  $y$  in  $T$  but  $y$  is ranked above  $x$  in  $R$ . We can then measure how well  $R$  fits  $T$  by adding up all the weights of all the edges for which  $R$  disagrees with  $T$ . This results in the *Kemeny score*<sup>1</sup> of  $R$  (with respect to  $T$ ), denoted by  $c(R, T)$  and given by the following formula:

$$c(R, T) = \sum \{w(x, y) \mid R \text{ disagrees with } T \text{ on } (x, y)\}.$$

For example if  $T$  is the weighted tournament in Fig. 1 and  $R = [D, A, B, C]$  then  $R$  disagrees with  $T$  on  $(B, D)$ ,  $(C, D)$  and  $(C, B)$ , which have weights

<sup>1</sup>Named after John Kemeny.

1, 9 and 2 respectively. Hence  $c(R, T) = 1 + 9 + 2 = 12$ . The ranking that best fits  $T$  is then taken to be any ranking  $R$  that has the lowest possible Kemeny score. Thus we arrive at the following optimisation problem:

**Optimisation problem.** Given a weighted tournament  $T$ , find a ranking  $R$  that minimises the Kemeny score  $c(R, T)$ .

A ranking that minimises the Kemeny score (with respect to a given tournament  $T$ ) is called a *Kemeny ranking* (with respect to  $T$ ). Note that some tournaments can have more than one Kemeny ranking. The tournament in Fig. 1 has just one Kemeny ranking, namely  $[A, C, B, D]$ .

## 2 Coursework questions

The tasks below require you to write a program (in a programming language of your choice) that is able to read a specific weighted tournament from a separate file (to be described below), and return a Kemeny ranking for this tournament - or at least an approximation of one - using simulated annealing (see Lecture 7, slide 13 for the pseudocode of the simulated annealing algorithm).

This coursework will use a particular weighted tournament containing real data (`ice_dance_1998.wtg`) from the 1998 Olympic Ice Dancing Championship, for which there were a total of 24 participants. This file, along with an explanation of its data format (see `How_to_read_tournament_dataset.pdf`) can be downloaded from Learning Central.

### First part: Setting SA parameters

As discussed during lecture, simulated annealing has a number of parameters that need to be fixed by the programmer in advance (see Lecture 7, slide 17). Some will be fixed below, but for others you will be free to experiment.

- ***initial solution.*** For the initial solution you must always use the ranking  $[1, 2, 3, 4, \dots, 23, 24]$ , using the numbering of candidates specified in `ice_dance_1998.wtg`.
- ***cooling schedule.*** For initial temperature  $TI$  and temperature length  $TL$  you are free to experiment with different values. Your only re-

striction regarding cooling ratio  $f$  is that  $f(T)$  should be of the form  $f(T) = a \times T$ , where  $a$  is some fixed parameter between 0 and 1 (which you are free to choose).

- **cost function.** The cost of a solution (i.e., possible ranking)  $R$  is the Kemeny score  $c(R, T)$ , defined in Section 1 above, where  $T$  is the tournament loaded from `ice_dance_1998.wtg`.
- **stopping criterion.** The algorithm should STOP if a pre-specified number `num_iterations` of iterations have passed without accepting a new solution. You are free to experiment with different values of `num_iterations`.

The only parameter not yet specified is the neighbourhood function, which brings us to...

**TASK 1.** Write down a definition of a *neighbourhood*  $N$  that is specific to this problem, i.e., specify the conditions for when any given ranking  $R_2$  is a neighbour of another  $R_1$ , and illustrate your definition with an example. Explain the rationale for your choice of definition (by, for example, showing that the cost of a neighbouring solution  $R_2$  can easily be computed from the cost of current solution  $R_1$ ).

**(4 marks for definition, 3 each for example and explanation - Total 10 marks)**

## Second part: Implementation

After having set up the parameters, you must now implement the algorithm.

**TASK 2.** Write a program implementing the simulated annealing algorithm to find a ranking that minimises the Kemeny score with respect to the given weighted tournament  $T$  specified in `ice_dance_1998.wtg`. You must use the parameters in accordance with the first part above, and use the neighbourhood you defined in Task 1. Your program should have the following behaviour. Successful completion results in the marks indicated:

1. It should be executable via the command line, taking the file `ice_dance_1998.wtg` as an argument. **(2 marks)**
2. Upon running, it should first prompt the user for the value of the parameter  $m$  needed for step 4 below. **(2 marks)**

3. Upon running, it should convert the tournament into a data structure suitable for the algorithm (you may use a matrix, as detailed in Section 1 above, or any other appropriate data structure). **(2 marks)**
4. During runtime, after every  $m$  iterations (using the value of  $m$  specified by the user in step 2 above), it should print to the screen the Kemeny scores of both the current solution  $R^{\text{now}}$ , and the best solution found so far  $R^{\text{best}}$ . **(2 marks)**
5. After running, your program should print to the screen the following information:
  - The best ranking it found (positions 1 to 24), in the form of a table, using the real names of the candidates, which are included in `ice_dance_1998.wtg`. **(2 marks)**
  - The Kemeny score of this best ranking. **(2 marks)**
  - The algorithm's runtime. **(2 marks)**

**(Total 14 marks)**

### Third part: Selecting best parameters and discussion

As stated in the first part, you are free to try out different values for the parameters  $TI$ ,  $TL$  and the “ $a$ ” in  $f(T) = a \times T$ , as well as the number `num_iterations` in the stopping criterion. The final task is about experimentally tuning to the best choice of parameters and reflecting on your results.

**TASK 3.** Give the values of the parameters  $TL$ ,  $TI$  and  $a$  (in  $f(T) = a \times T$ ), as well as `num_iterations`, which seem to give the best solutions for your program. For this “best” choice of parameters provide screenshots of the output of your program. Write a short summary of your results, indicating the range of different values you tried for the parameters, which parameters' variation had the biggest effect on the output solution, etc. (maximum 300 words).

**(3 marks for screenshots, 3 marks for summary - Total 6 marks)**