

## AXIAL THREE-INDEX ASSIGNMENT AND TRAVELLING SALESMAN PROBLEMS: FAST APPROXIMATE ALGORITHMS AND THEIR PROBABILISTIC ANALYSIS

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### 1. Introduction

*Three-index axial assignment problem* (3-AAP) (see [1]–[3]) is formulated as follows: Minimize the linear form

$$\sum_{i=1}^n c_{i, \pi(i), \sigma \pi(i)} \quad (1)$$

on the set of substitutions  $\pi, \sigma$  from symmetric group  $S_n$  of order  $n$ , where  $c_{ijk}$  are given real numbers,  $1 \leq i, j, k \leq n$ .

Along with this well-known assignment problem we consider the following new generalization of the classical travelling salesman problem.

*Three-index axial travelling salesman problem* (3-ATSP) represents a minimization of (1) with the additional condition

$$\pi, \sigma, \sigma \pi \in P_n, \quad (2)$$

where  $P_n$  is a set of all cyclic substitutions (i. e., consisting of one cycle) from  $S_n$ .

Theoretical results in [4]–[6] imply that both the problems are *NP*-complex; moreover 3-ATSP is *MAX SNP*-complex. This circumstance stimulated the authors to construct for these problems approximate polynomial algorithms on random inputs.

Fast approximate algorithms and probabilistic distributions on which these algorithms turn out optimal, represent a significant interest in the discrete optimization (see [7]–[14]).

We denote by  $f_A$  and  $f^*$  an approximate (obtained by algorithm  $A$ ) value and an optimal value, respectively, of the objective function on a certain input. Recall that, in the case of concrete input data (problem's input), they usually speak on an individual problem. By a mass problem (or simply problem) we shall mean a definite set of individual problems.

In following [8], we shall say that an algorithm  $A$  has estimates  $(\varepsilon_A, \delta_A)$  on a class of inputs of the problem under consideration if the inequality is valid

$$\Pr\{f_A > (1 + \varepsilon_A)f^*\} \leq \delta_A,$$

where  $\varepsilon_A$  is the estimate of the relative error of solution obtained by the algorithm  $A$ ,  $\delta_A$  is the probability that algorithm  $A$  fails (i. e., the quantity  $\delta_A$  can be treated as the part of cases when the algorithm does not guarantee an error lying within the limits  $\varepsilon_A$ ).

It seems of interest to clarify the behavior of the estimates  $\varepsilon_A$  and  $\delta_A$  under increase of problem's dimension.

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