



شبكة المعلومات الجامعية

بِسْمِ اللَّهِ الرَّحْمَنِ الرَّحِيمِ





شبكة المعلومات الجامعية



شبكة المعلومات الجامعية

التوثيق الالكتروني والميكرو فيلم

جامعة عين شمس

التوثيق الالكتروني والميكرو فيلم

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لم ترد بالأصل



جامعة أسيوط



كلية العلوم

Meta-heuristic for Combinatorial Optimization

A THESIS

Submitted to the Faculty of Science
Assiut University

In Partial Fulfillment for the Degree of Master of Science
Mathematics (Computer Science)

BY

Sara Salah Mohammed

B. Sc. Computer Science
Faculty of Science, Assiut University

Supervised by

Prof. Dr. Hassan M. H. El-
Hawary

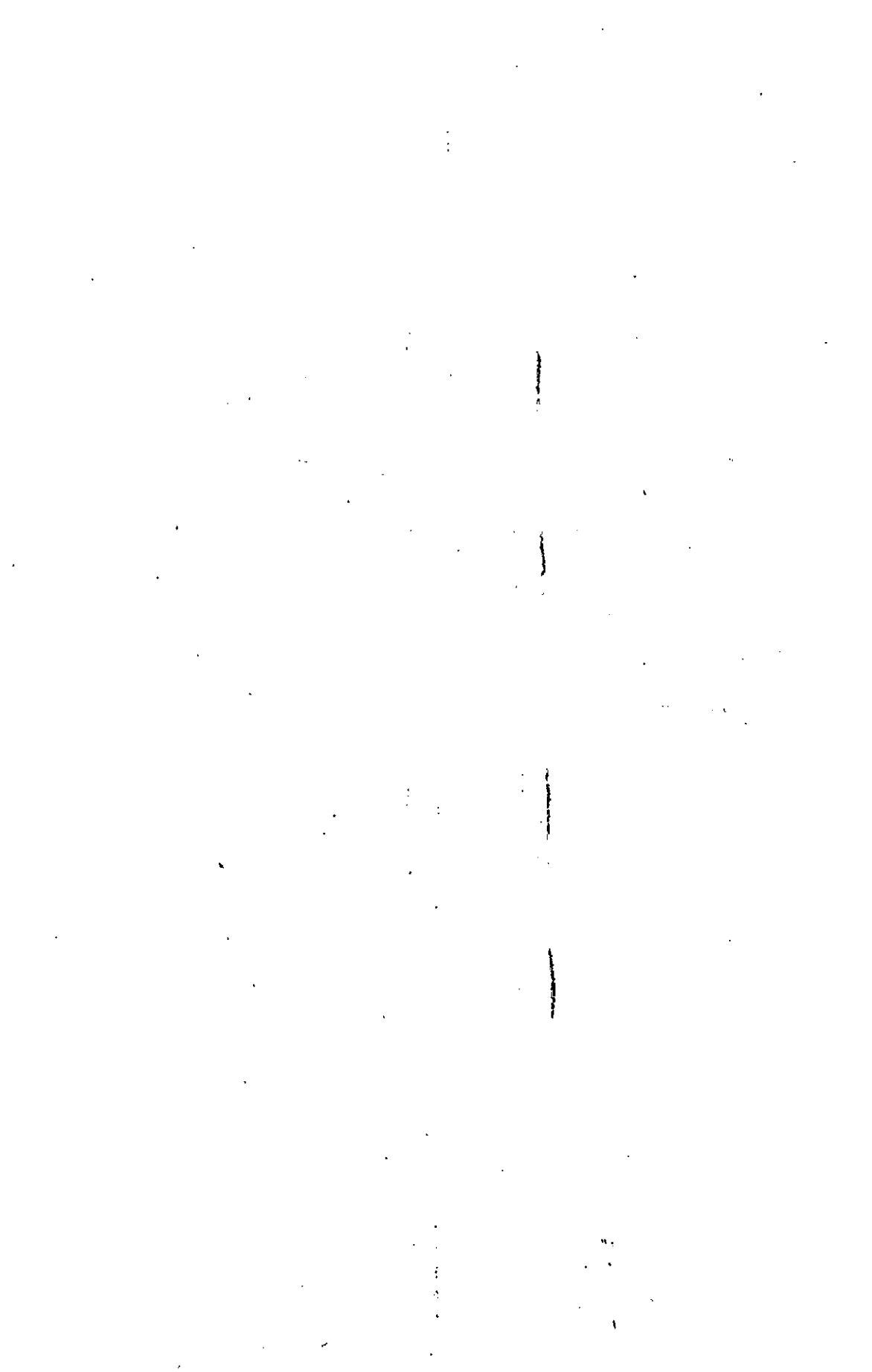
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Assiut University

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Approval Sheet

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Degree : Master of Science Mathematics (Computer Science)
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- Simulation
- Special Course In Computer Science
- Artificial Neural Network
- Option Course
- Numerical Analysis
- Data Base
- Practical Course

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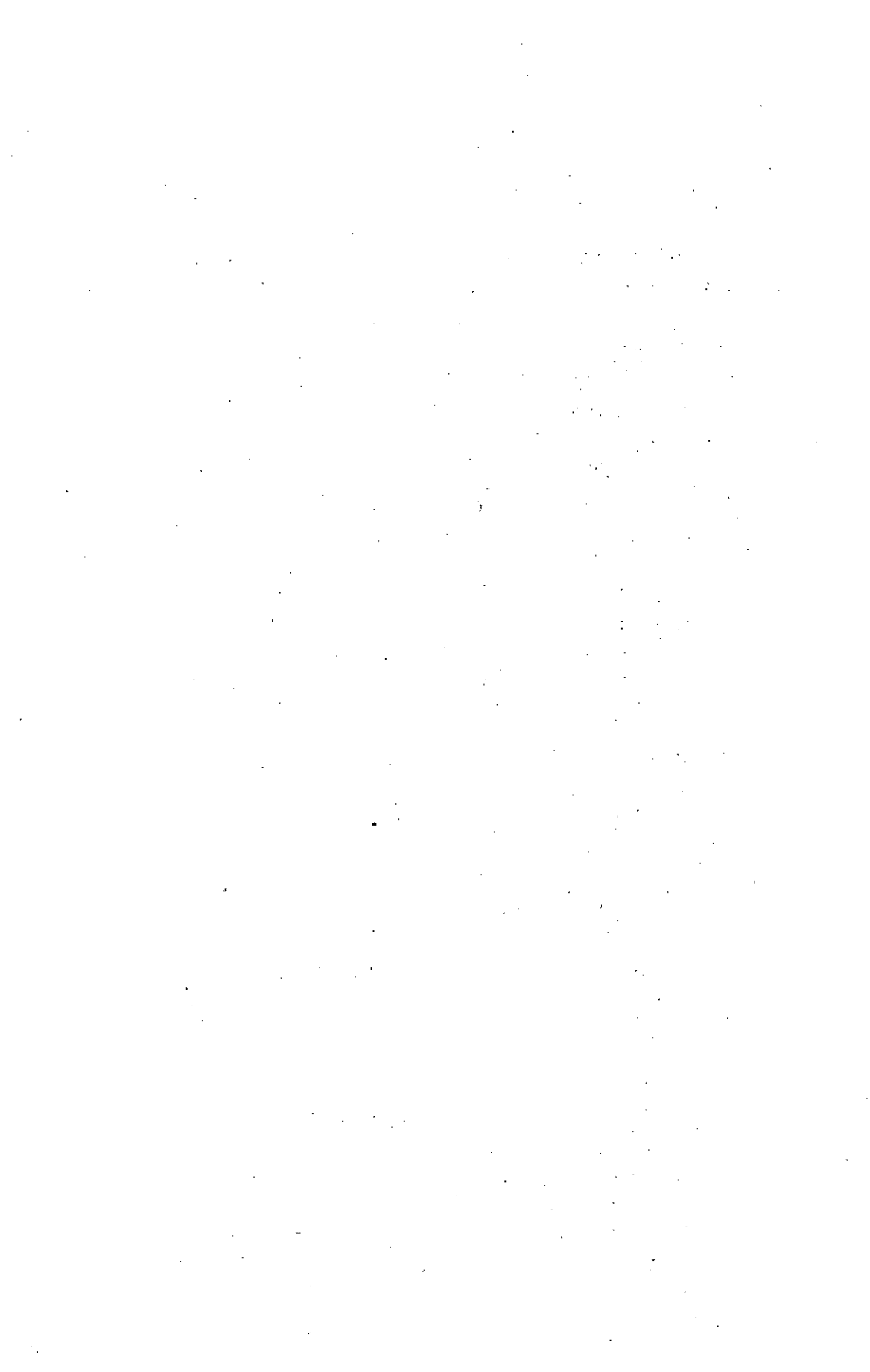


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