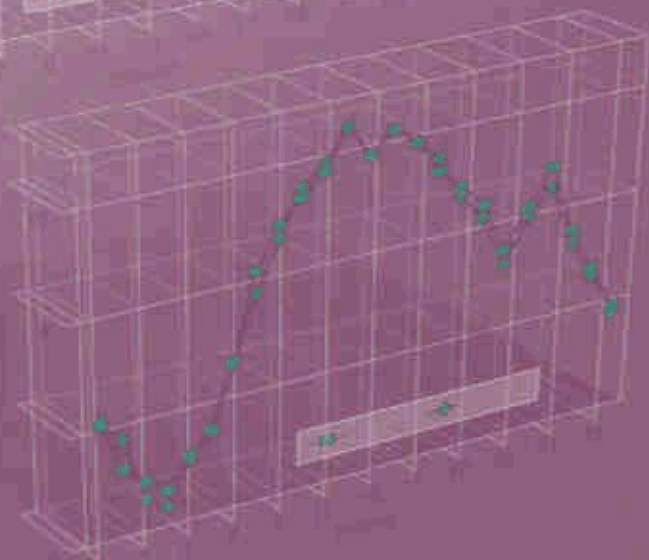
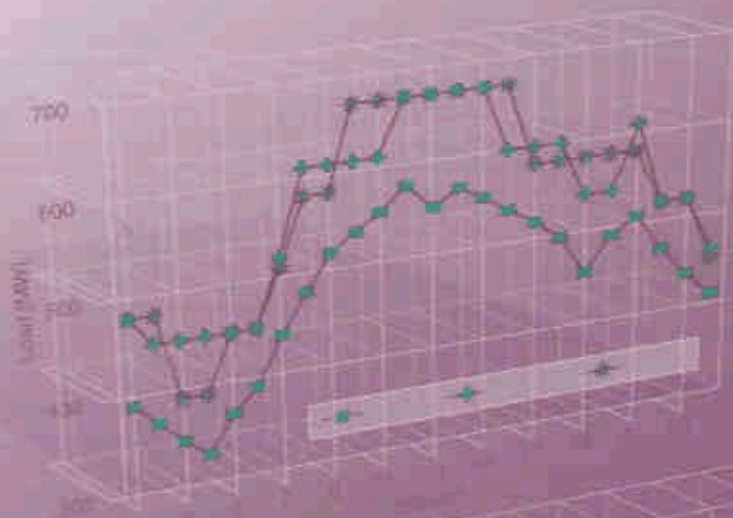


# MODERN HEURISTIC OPTIMIZATION TECHNIQUES

THEORY AND APPLICATIONS TO POWER SYSTEMS

EDITED BY

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POWER  
ENGINEERING

Mohamed E. El-Hawary, Series Editor

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## 19 Hybrid Systems

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