

Algorithmic Information Theory:

Algorithmic Complexity /-Probability and -Induction

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This references are an overview of the field of algorithmic information theory (AIT), its most important and recent concepts. AIT arises by mixing information theory and computation theory to quantify the notions of information and complexity in an individual object, and in so doing gives also rise to an notion of randomness of individual objects. Furthermore the list contains literature of the important subfield of algorithmic probability and induction.

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