

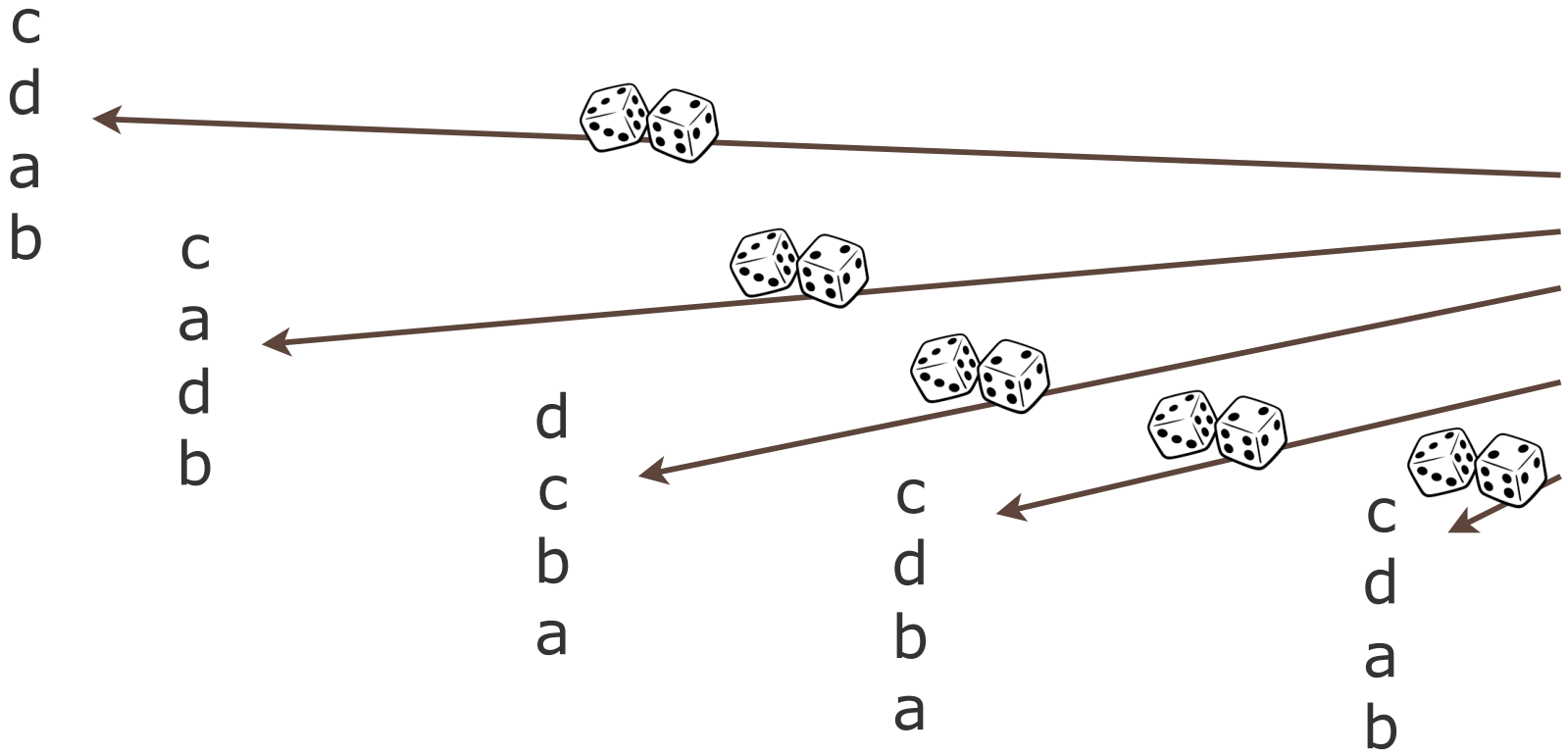
Algorithms For Democratic Decision-Making

Jamie Tucker-Foltz • Yale University • Spring 2026

Lecture 7: **The Epistemic Approach**

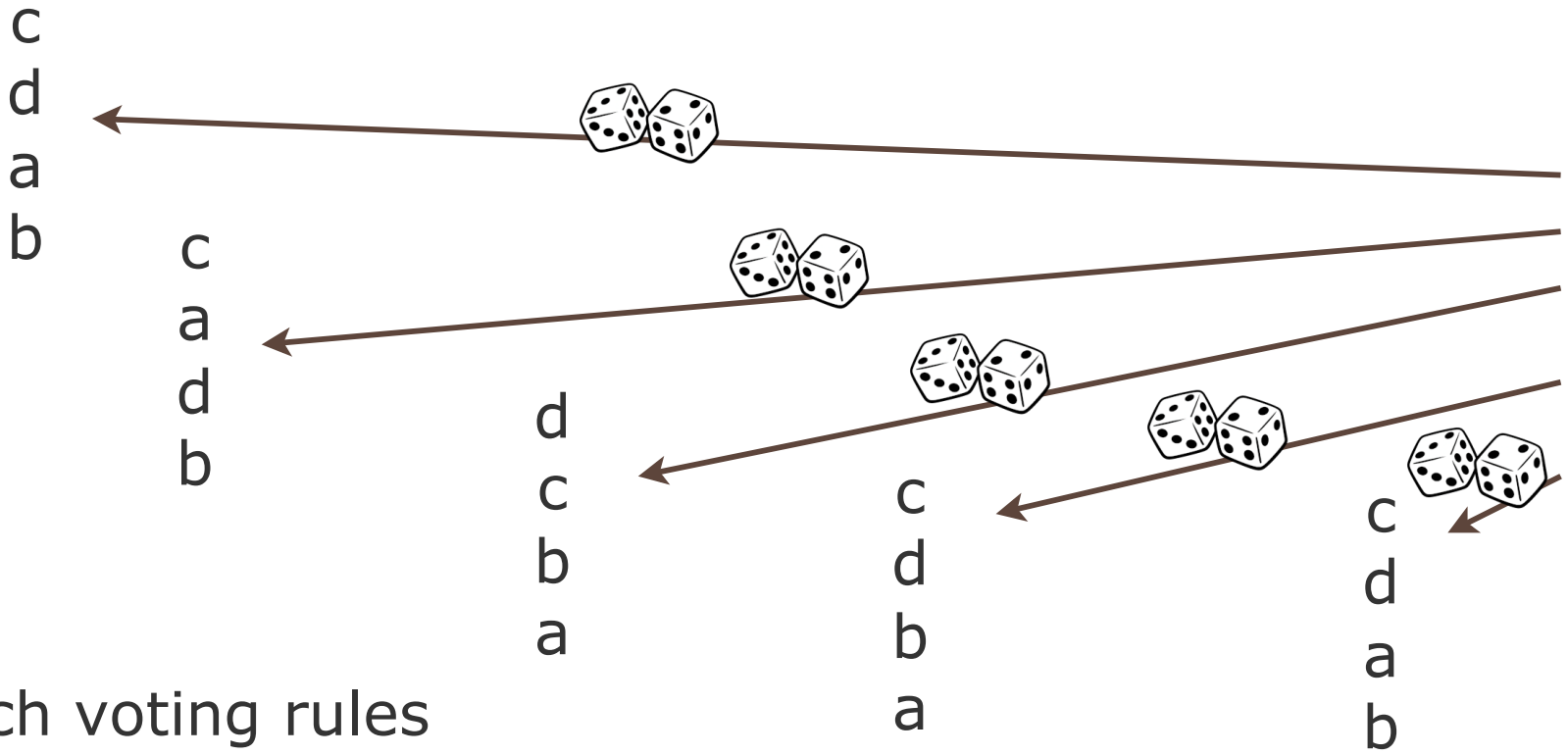
Introduction

Alternative perspective on voting:
There is a *ground truth* ordering of alternatives.



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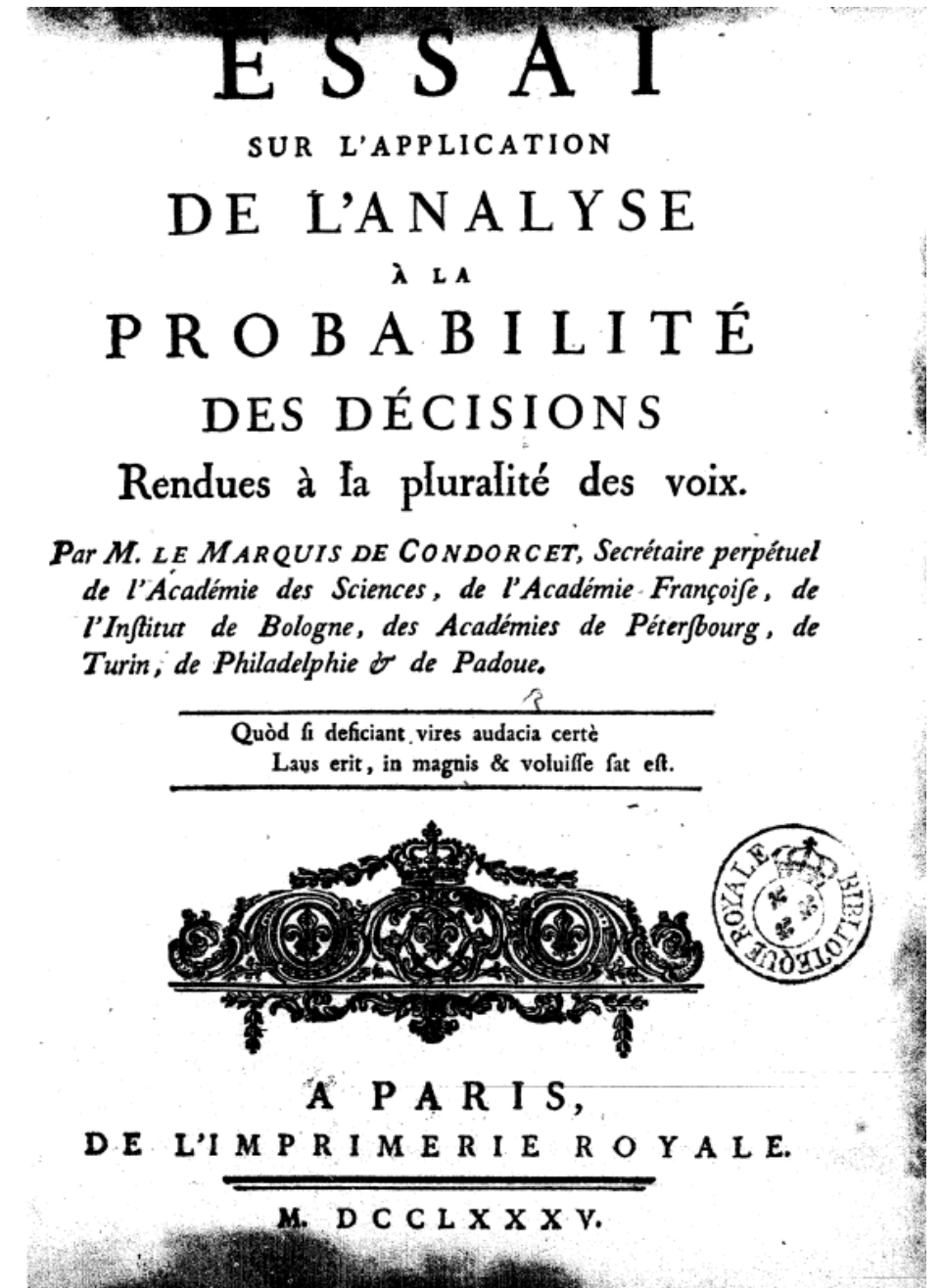
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Which voting rules
recover the ground truth?



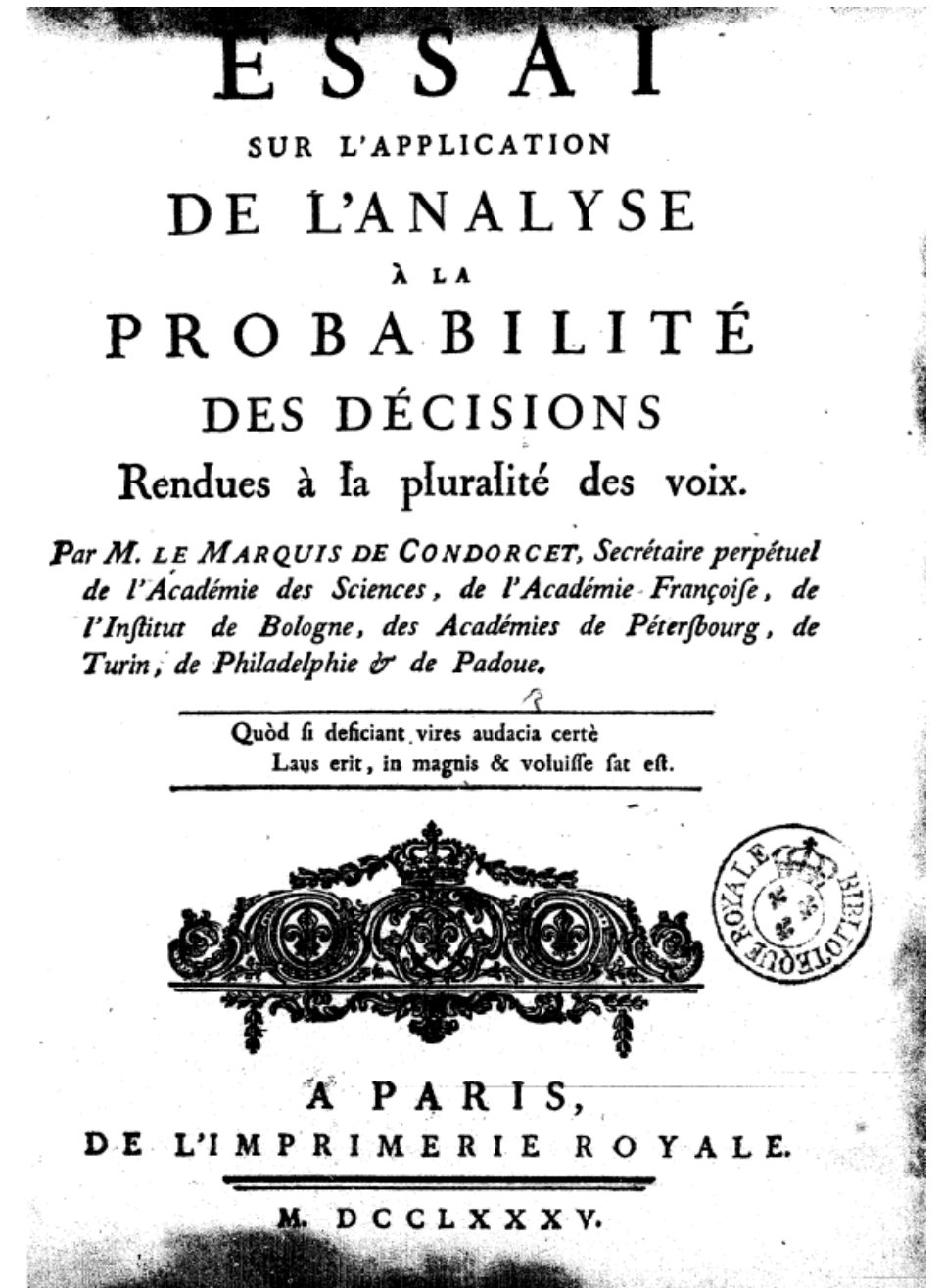
Condorcet's Jury Theorem



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"Essay on the Application of Analysis to the Probability of Majority Decisions"

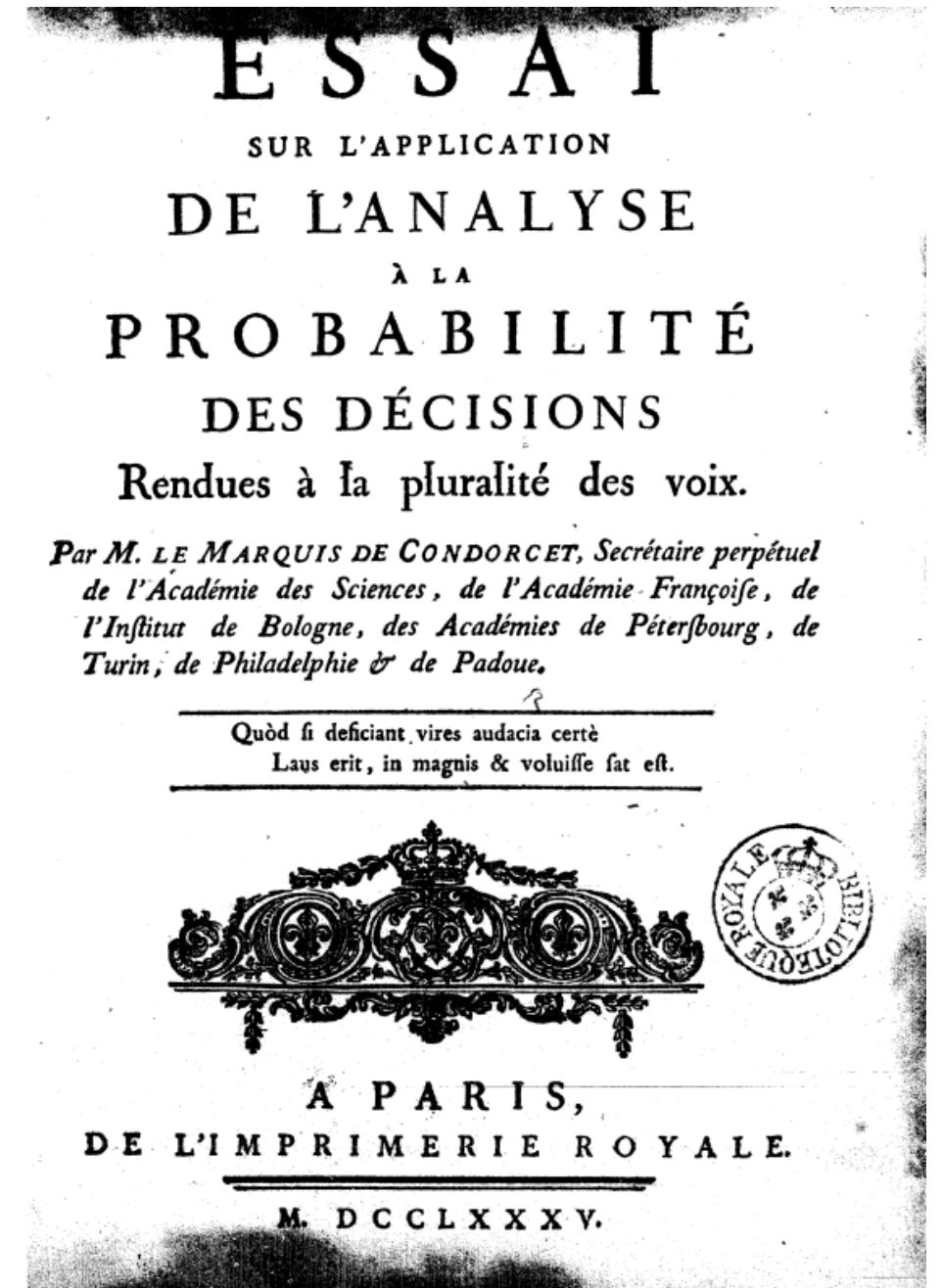


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Model: An odd number n of voters must decide among $m = 2$ alternatives, one of which is *correct*. Each voter prefers the correct candidate with probability $p > 1/2$.



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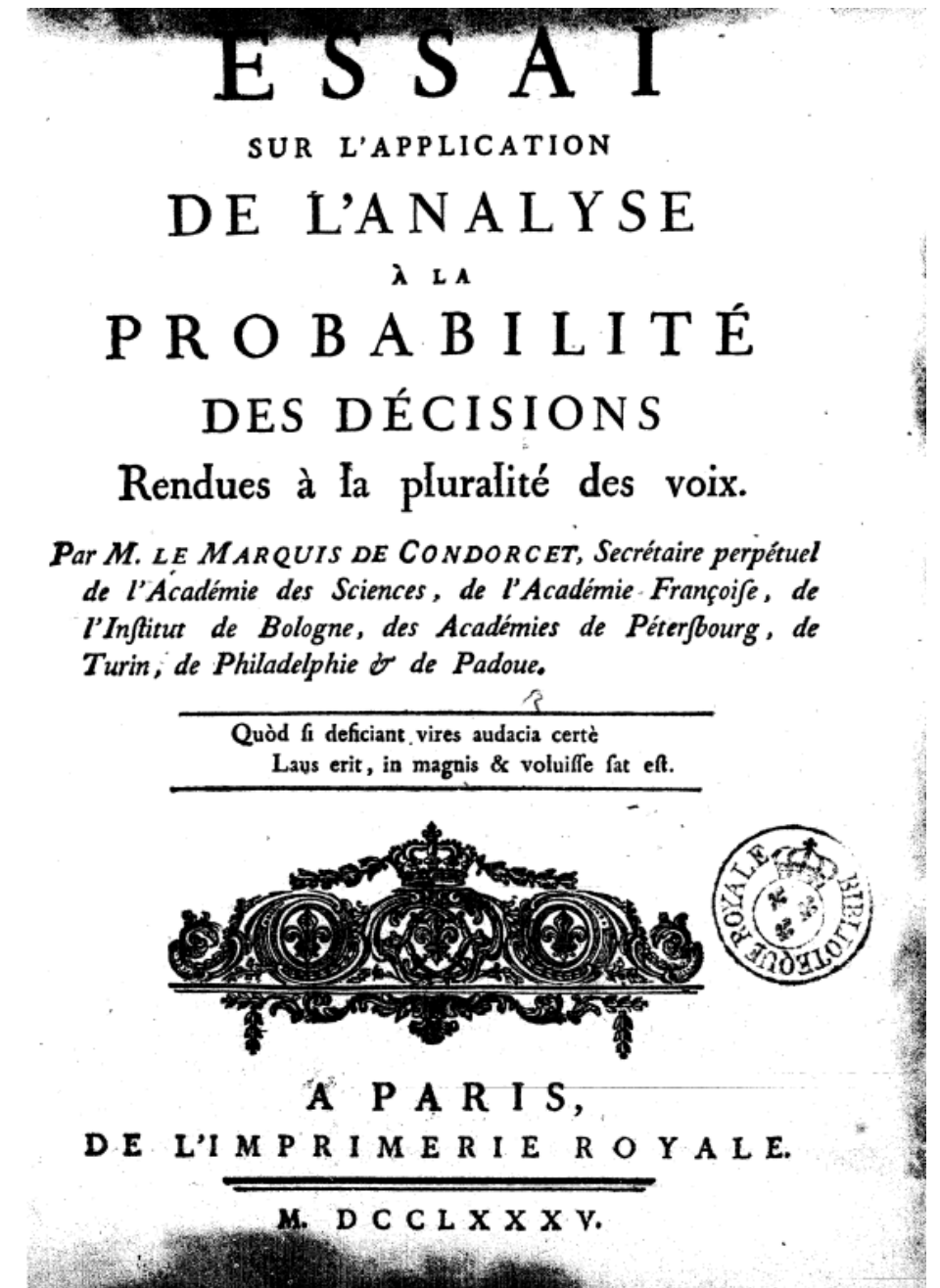


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Model: An odd number n of voters must decide among $m = 2$ alternatives, one of which is *correct*. Each voter prefers the correct candidate with probability $p > 1/2$.

Theorem (CJT)

The probability that majority rule elects the correct candidate is increasing in n and tends to 1 as $n \rightarrow \infty$.



Proof of Condorcet's Jury Theorem (limiting case)

Let X_i be a random variable that is 0 or 1 depending on whether voter i votes incorrectly or correctly.

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$$\Pr[Z] = \Pr \left[\sum_{i=1}^n X_i \leq n/2 \right]$$

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Hoeffding's Inequality

The probability that the sum of indep. random variables $a_i \leq X_i \leq b_i$ deviates from its mean by t in one direction is

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- True probability is 97.8%

What about $m \geq 3$?

The *Kendall tau distance* between two rankings $\sigma_1, \sigma_2 \in \Sigma([m])$ is

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Kemeny rule: Output the Most Likely Explanation (MLE) for the ground truth, i.e., the π of maximum probability under the Mallows model.

Computing the MLE

$$\Pr[\sigma_1, \sigma_2, \dots, \sigma_n] = \prod_{i=1}^n \frac{\phi^{d_{KT}(\sigma_i, \pi)}}{\sum_{\sigma' \in \Sigma([m])} \phi^{d_{KT}(\sigma', \pi)}}$$

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► How does the Kemeny ranking compare a vs b?



Respond at:
pollev.com/jtuckerfoltz255 or
bit.ly/jtfpoll or
 text jtuckerfoltz255 to 37607

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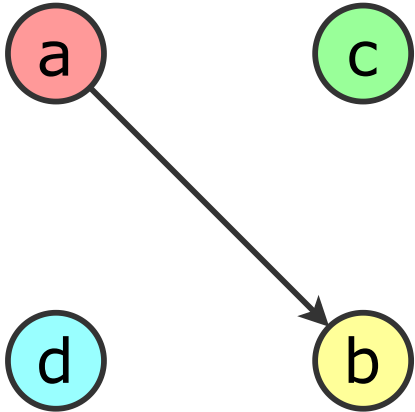
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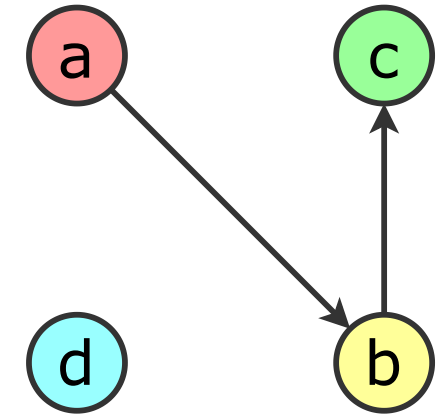
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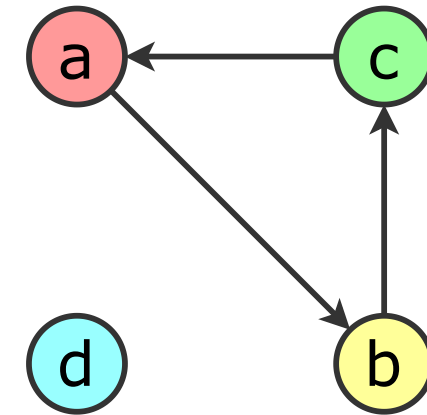
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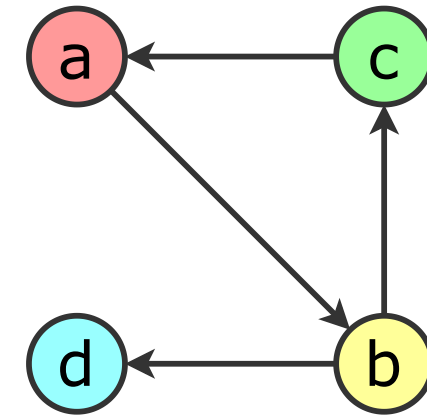
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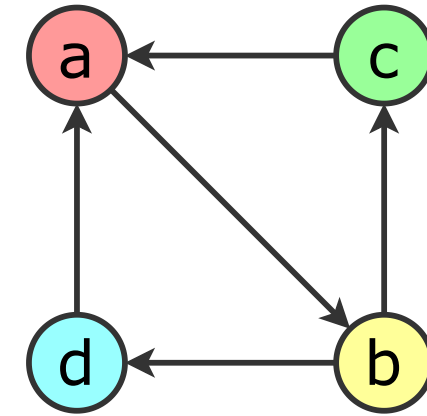
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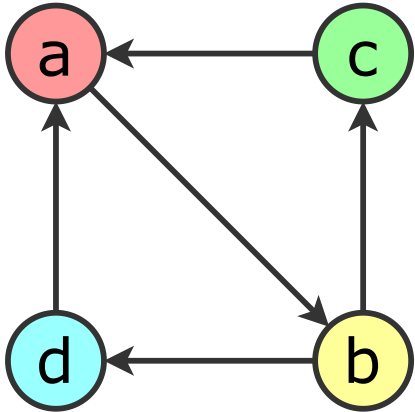
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Proposition

There is a ranking that disagrees with at most k edges



There is a set of at most k edges whose removal makes the graph acyclic.

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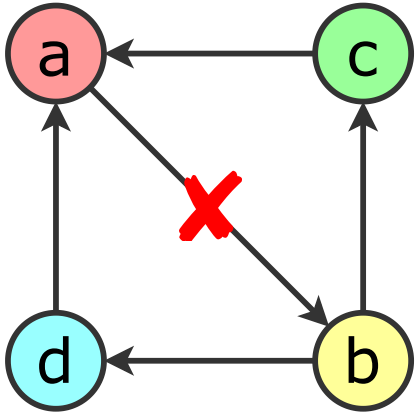
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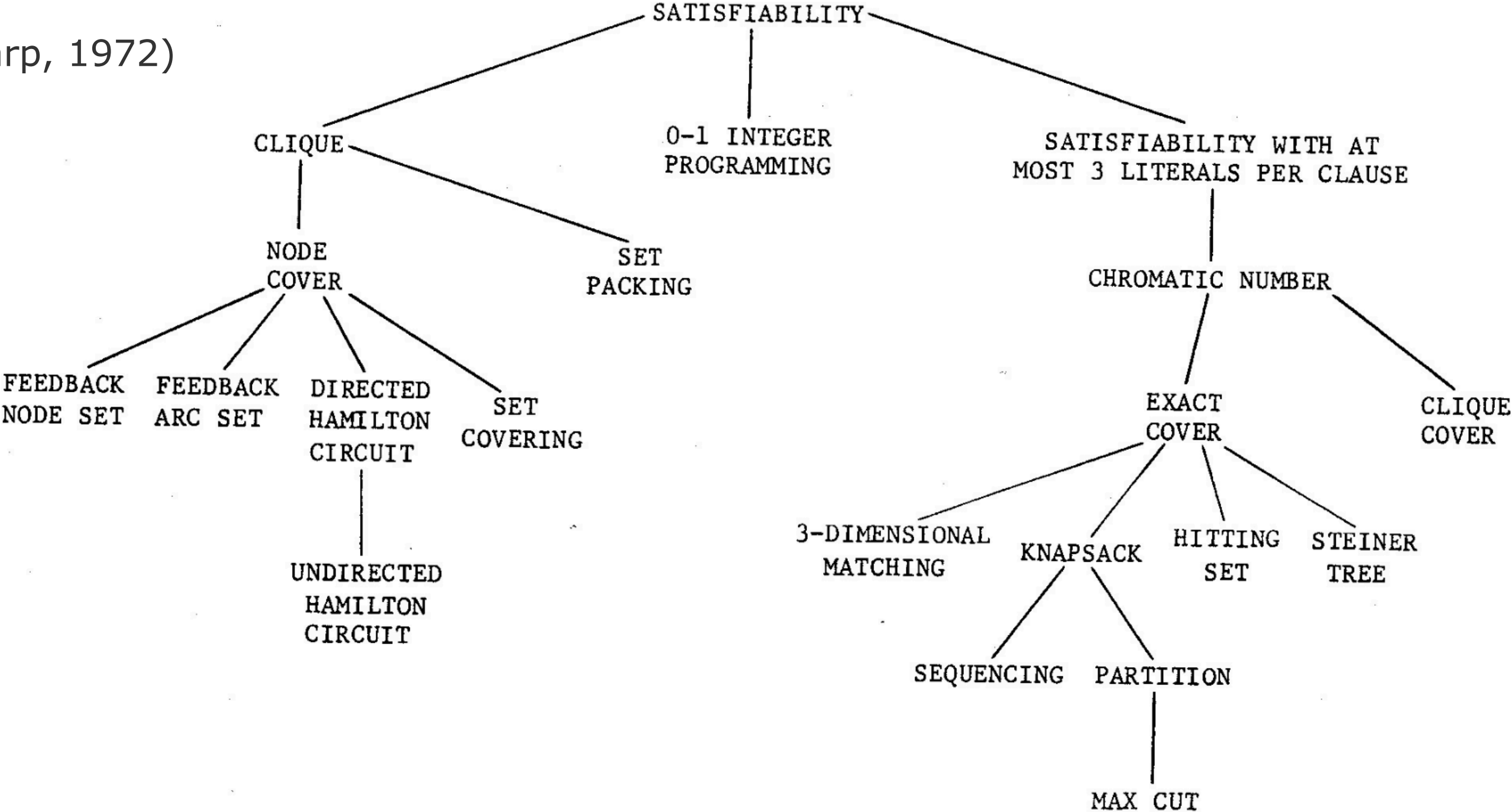
Solution: $b > \{c, d\} > a$

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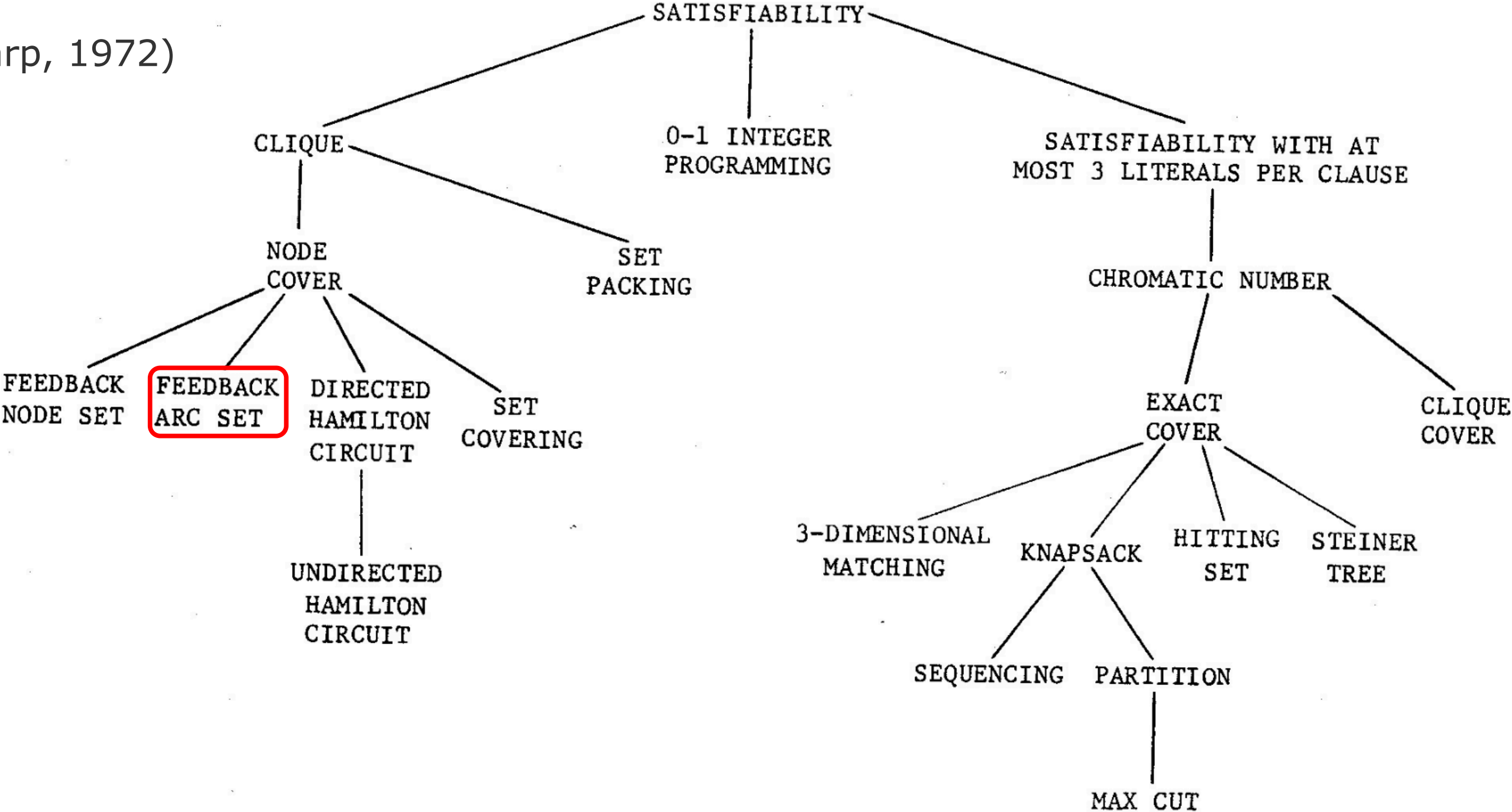
Kemeny is NP-complete

(Karp, 1972)



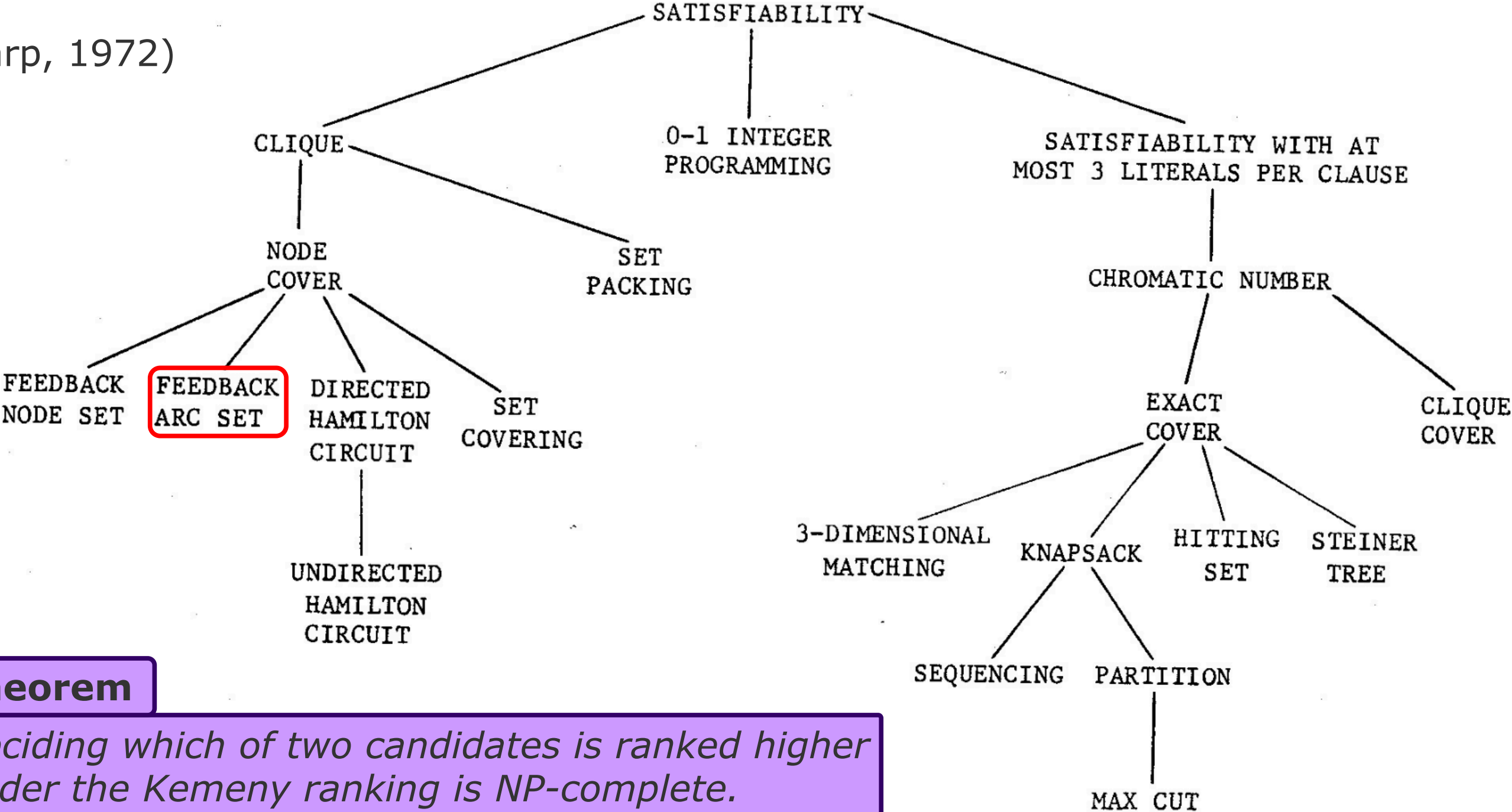
Kemeny is NP-complete

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Kemeny is NP-complete

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Theorem

Deciding which of two candidates is ranked higher under the Kemeny ranking is NP-complete.