



Week 5 Video 2

Relationship Mining

Causal Mining

Causal Data Mining

- These slides developed in partnership with Stephen Fancsali, Carnegie Learning, Inc.

Causal Data Mining

- Distinct from prediction or correlation mining
- The goal is not to figure out what predicts X ,
- or to figure out what is correlated to X ,
- but instead...

Causal Data Mining

find *causal* relationships in data.

- ▣ A causes B

Examples from Scheines (2007):

What features of student behavior cause learning?

What will happen when we make everyone take a reading quiz before each class?

What will happen when we program our tutor to intervene to give hints after an error?



Causal Data Mining

- Use graphs to represent causal structure
 - ▣ Frequently directed graphs without cycles
 - (Bayesian networks – see week 4 slides)
 - Nodes represent variables
 - (Directed) edges represent causal relationships

Causal Data Mining

- Algorithms infer (classes of) causal graphs that explain dependencies in observed data
 - ▣ From observed data alone, often cannot infer a *unique* causal graph.

Finding Causal Structure

- Easy to determine if you intervene
 - ▣ Some experiments are impossible, too expensive, unethical, etc.
- Can you determine this from purely correlational data?
 - ▣ Spirtes, Glymour, and Scheines say: sometimes, yes!

Example

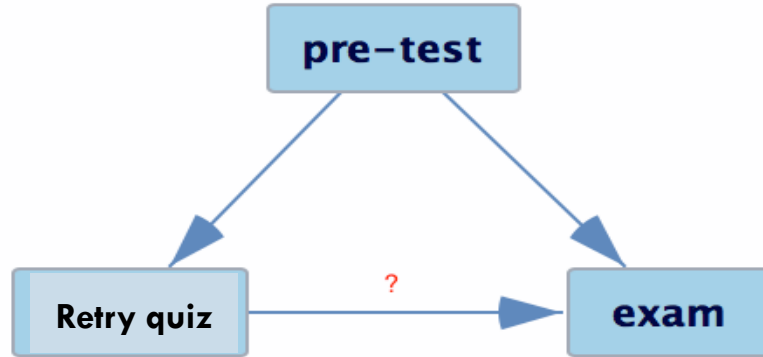
- Is repeatedly retrying quizzes harmful?
 - ▣ Does repeatedly retrying quizzes *cause* decreased learning?
- Suppose an investigator notices that repeatedly retrying quizzes and exam score are negatively associated (i.e., correlated).

Causal Graphs



- A direct causal relationship could explain this correlation...

Causal Graphs

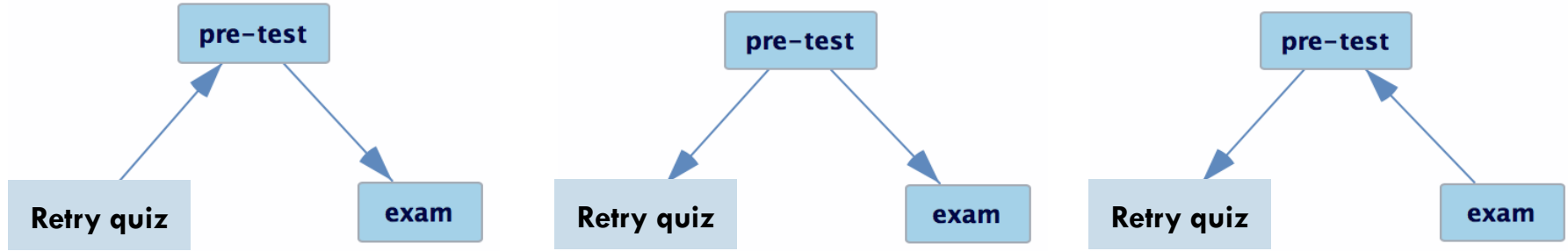


- or the correlation of *retry quiz* and *exam* might arise from a common cause, e.g., prior knowledge.
 - ▣ (or both!)

Causal Graphs

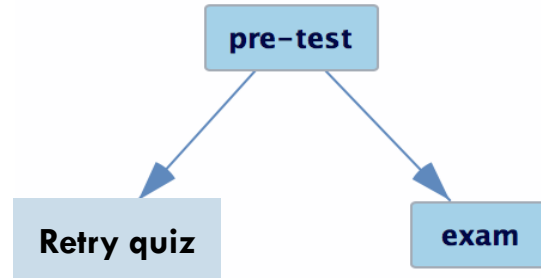
- Suppose that when we control for *pre-test*, the correlation of *retry quiz* & *exam* disappears.
 - ▣ E.g., the partial correlation is not significantly different from zero.

Causal Graphs



- Three causal graphs can explain this conditional independence equally well...

Causal Graphs



- but only one is compatible with background knowledge
 - ▣ *pre-test* is prior to behavior in a tutor and a final *exam*.

Big idea

- Infer class of graphs that can represent the full pattern of such (in)dependencies among measured variables.

Causal Data Mining

- TETRAD is a key software package used to study this
- <http://www.phil.cmu.edu/projects/tetrad/>

TETRAD

- Implements multiple algorithms for inferring causal structure from data
 - ▣ Different algorithms are applicable given particular assumptions.

Assumptions guide algorithm choice

- Are there unmeasured common causes?
- Linear relationships between variables?
- Are underlying dynamics acyclic or cyclic?
- Distribution of variables: Gaussian vs. non-Gaussian
- See TETRAD User Guide for detailed discussion....

Math & Assumptions

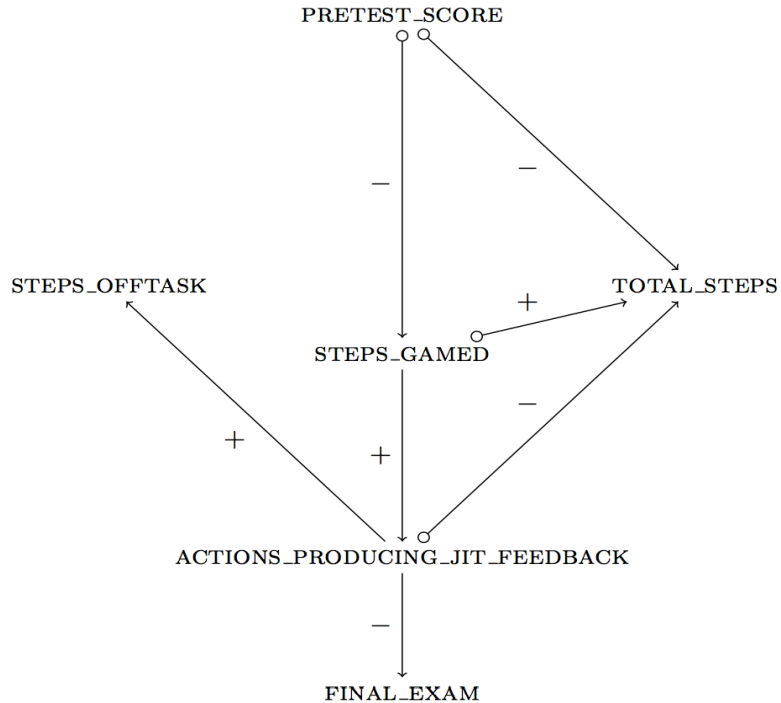
- See

Scheines, R., Spirtes, P., Glymour, C., Meek, C., Richardson, T. (1998) The TETRAD Project: Constraint Based Aids to Causal Model Specification. *Multivariate Behavioral Research*, 33 (1), 65-117.

Glymour, C. (2001) *The Mind's Arrows*

Examples in EDM

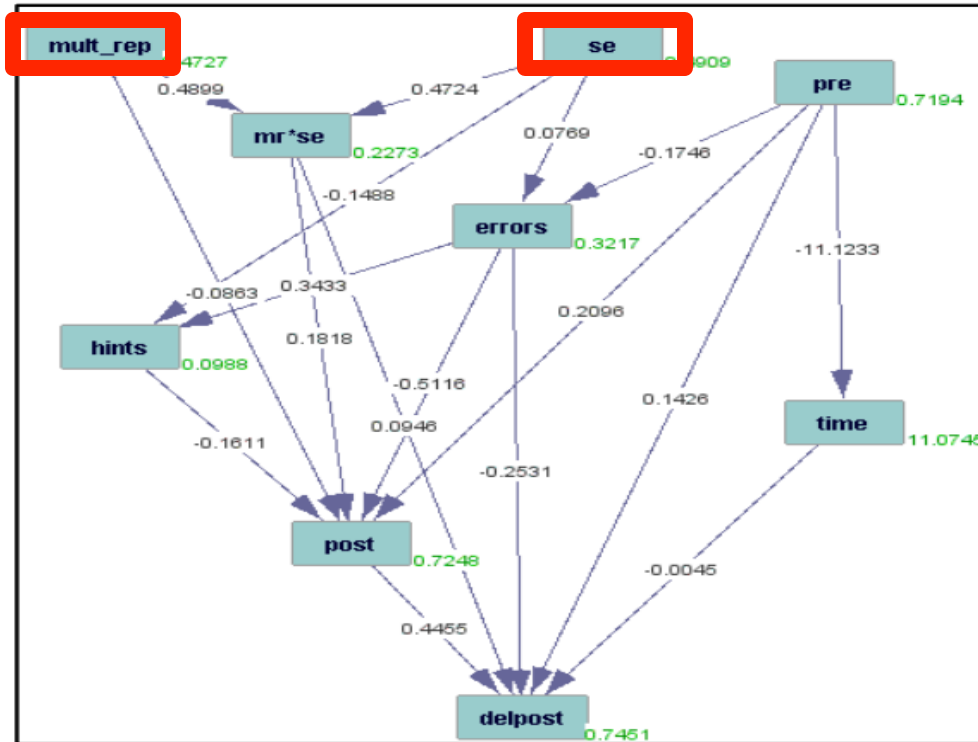
Fancsali (2013) Example



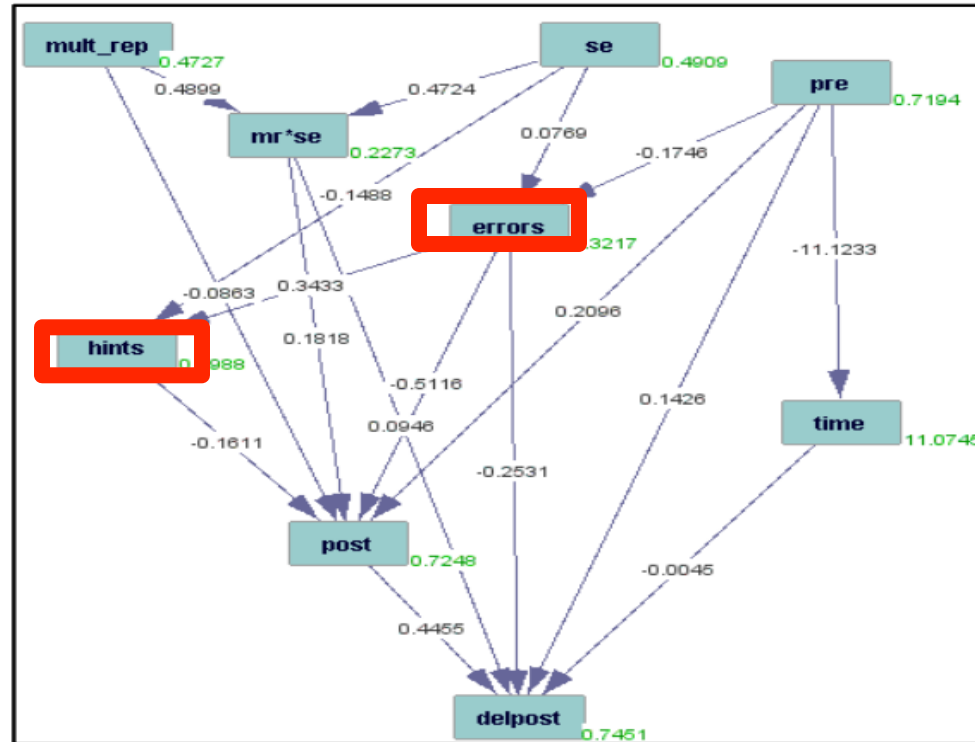
This example uses an algorithm that allows for unmeasured common causes of measured variables.

- pretest_score* → *total_steps* can signify
- (1) *pretest_score* is a cause of *total_steps*;
 - (2) *pretest_score* & *total_steps* share a common cause;
 - (3) both!

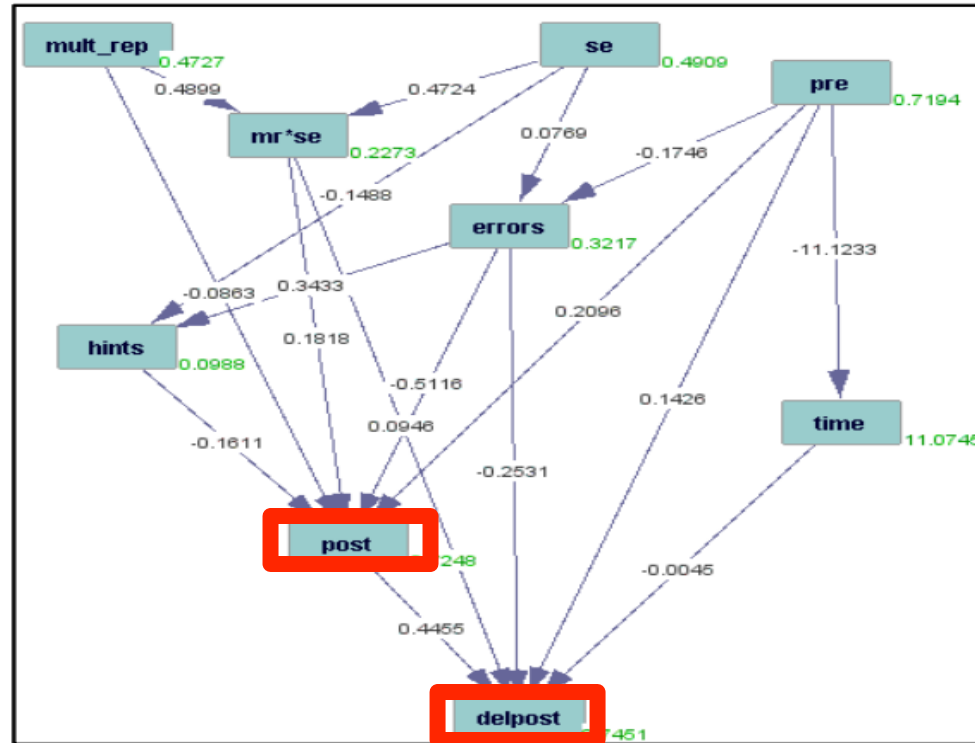
Rau & Scheines (2012)



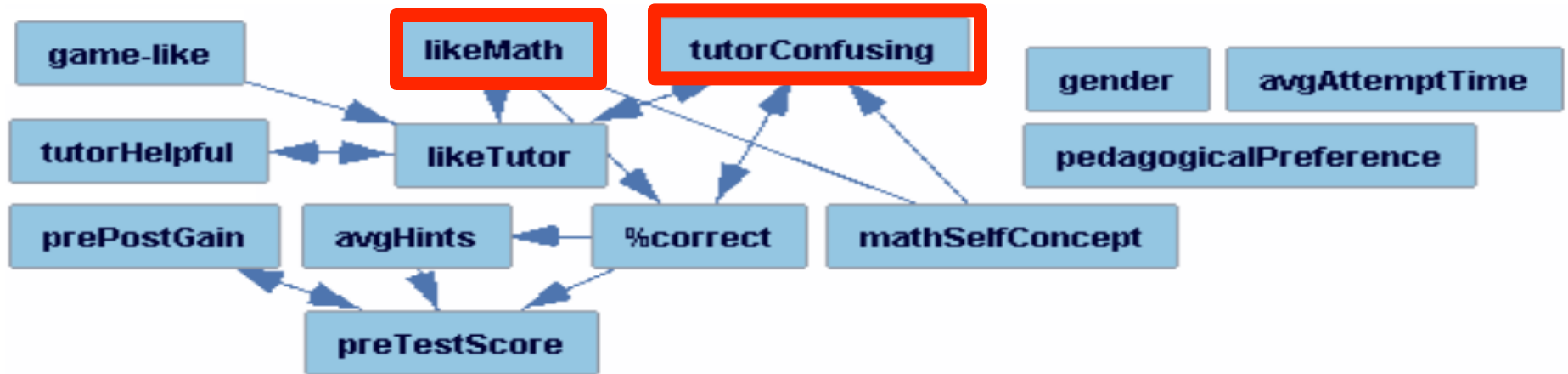
Rau & Scheines (2012)



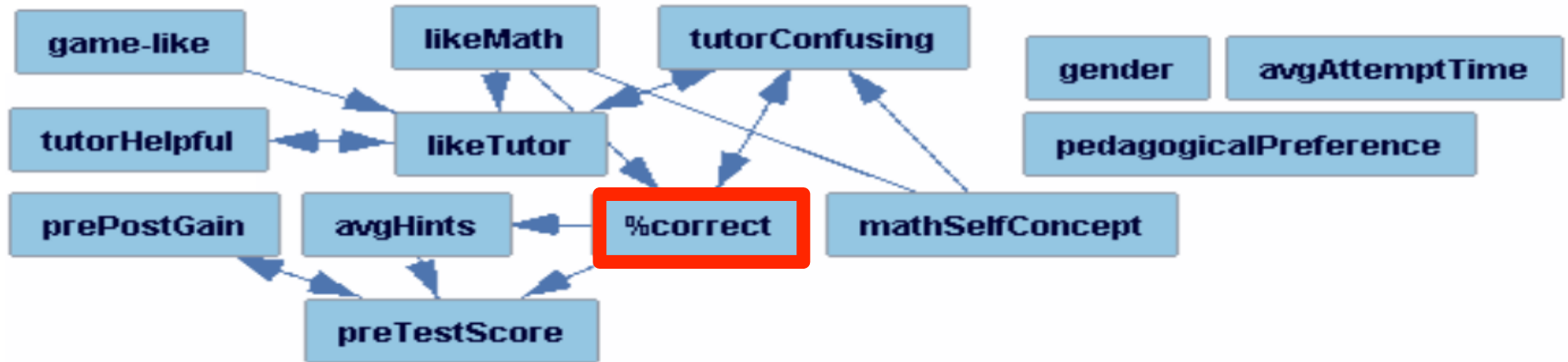
Rau & Scheines (2012)



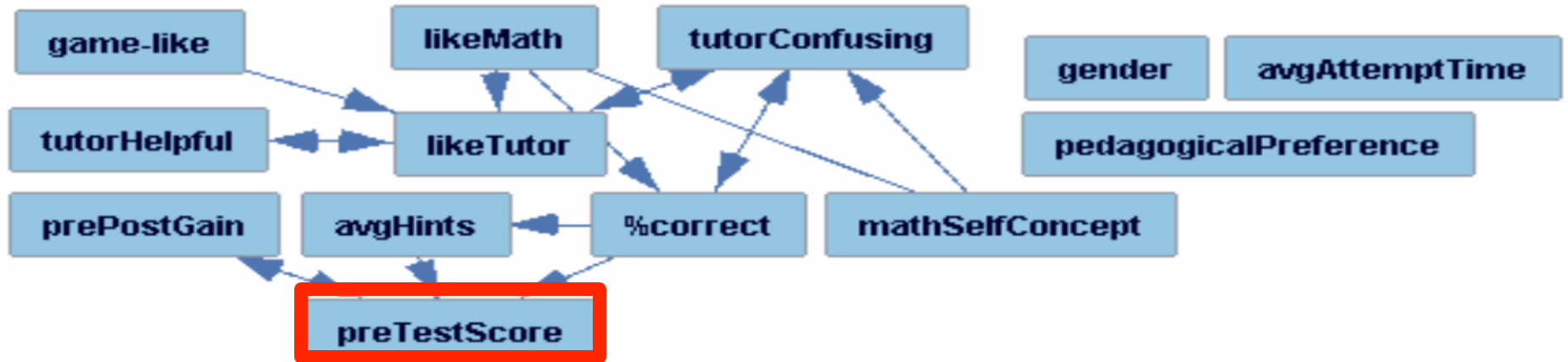
Rai et al. (2011)



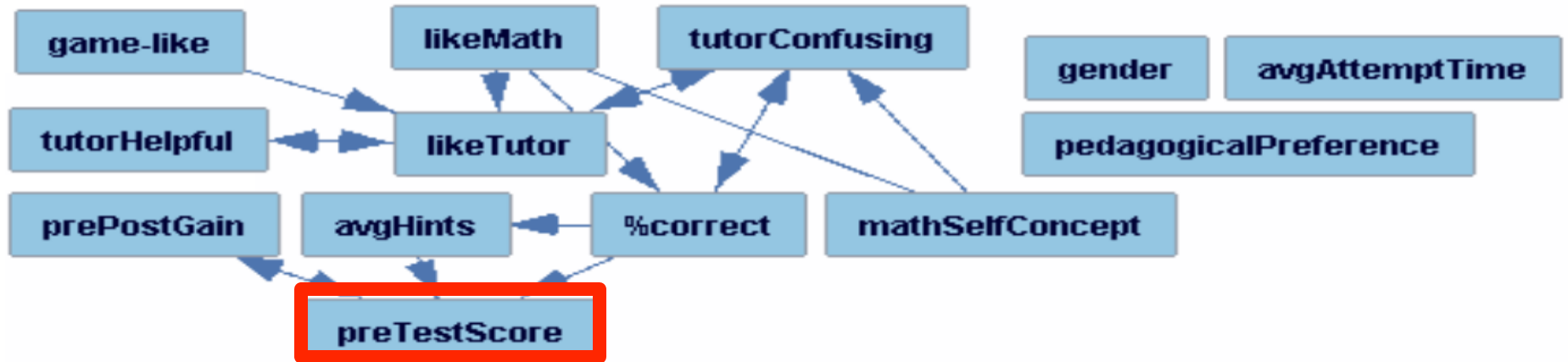
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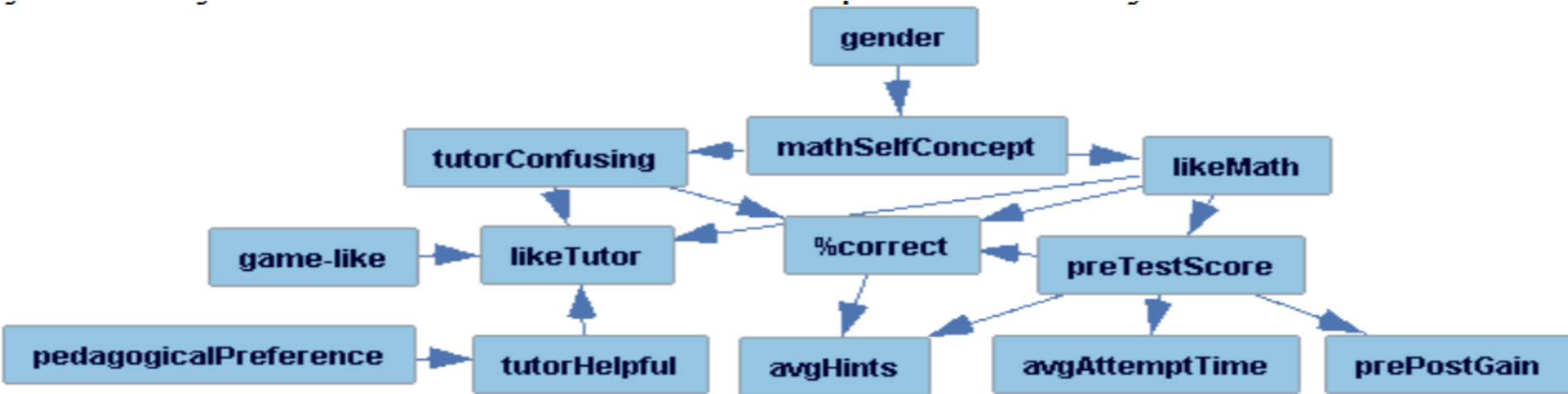
Wait, what?



Solution

- Use domain knowledge to constrain search.
- The future can't cause the past.
 - ▣ cf. example of *pre-test* being prior to *retry quiz* & *exam*.

Result

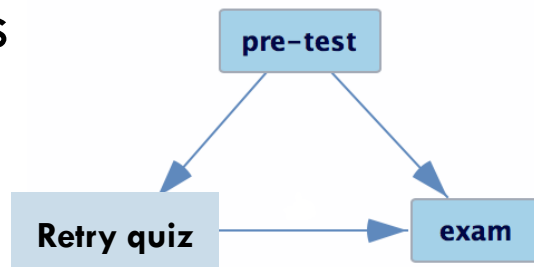


Important

- Important to use causal modeling algorithms correctly!
 - ▣ Which assumptions are reasonable?
 - ▣ The future can't cause the past
 - Except in movies

Important

- Are variables good proxies for what we intend to study (especially if “latent”)?
 - ▣ Suppose *pre-test* isn’t an appropriate measure of prior knowledge.
 - ▣ *pre-test* might not “screen off” *retry quiz* & *exam*, so we might still think that *retry quiz* causes decreased learning (*exam*).



Causal Modeling

- A powerful tool
- But needs to be used carefully!

Next lecture

- Association rule mining