

Intelligent Systems (AI-2)

Computer Science cpsc422, Lecture 32

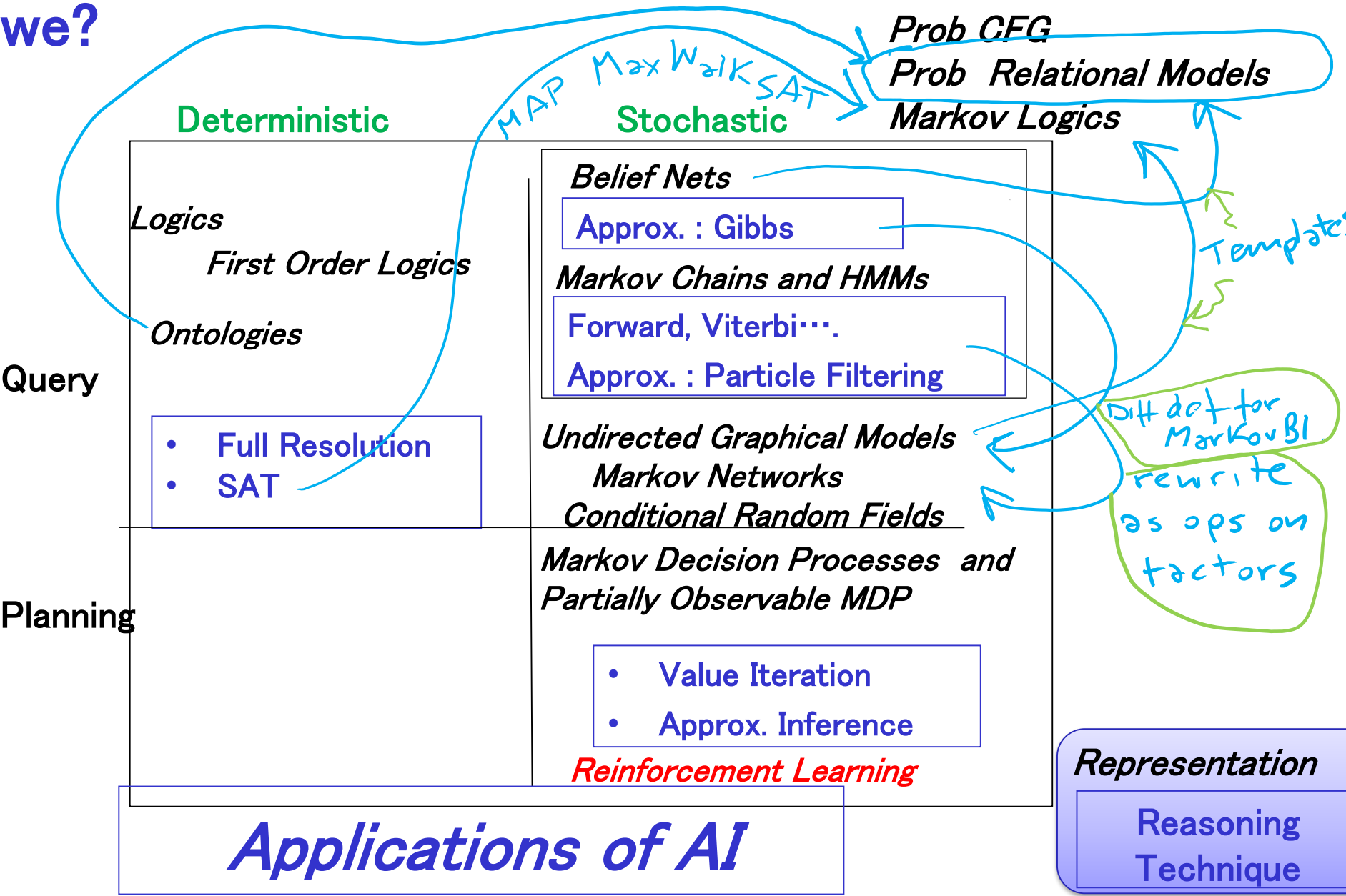
Nov, 28, 2016

Slide source: from David Page (MIT) (which were from From Lise Getoor, Nir Friedman, Daphne Koller, and Avi Pfeffer) and from Lise Getoor

422 big picture: Where are we?

StarAI (statistical relational AI)

Hybrid: Det +Sto



Combining Symbolic and Probabilistic R&R systems

(a) Probabilistic Context-Free Grammars

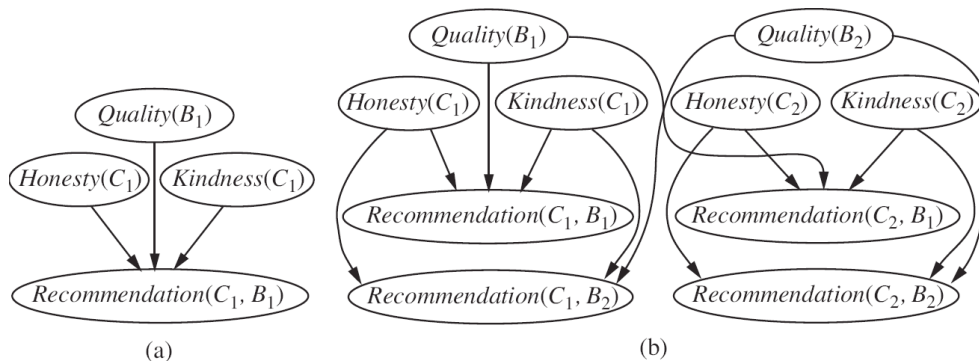
- Weights are conditional prob. on rewriting rules
- Applications: NLP parsing & Hierarchical Planning

(b) Markov Logics: weighted FOL

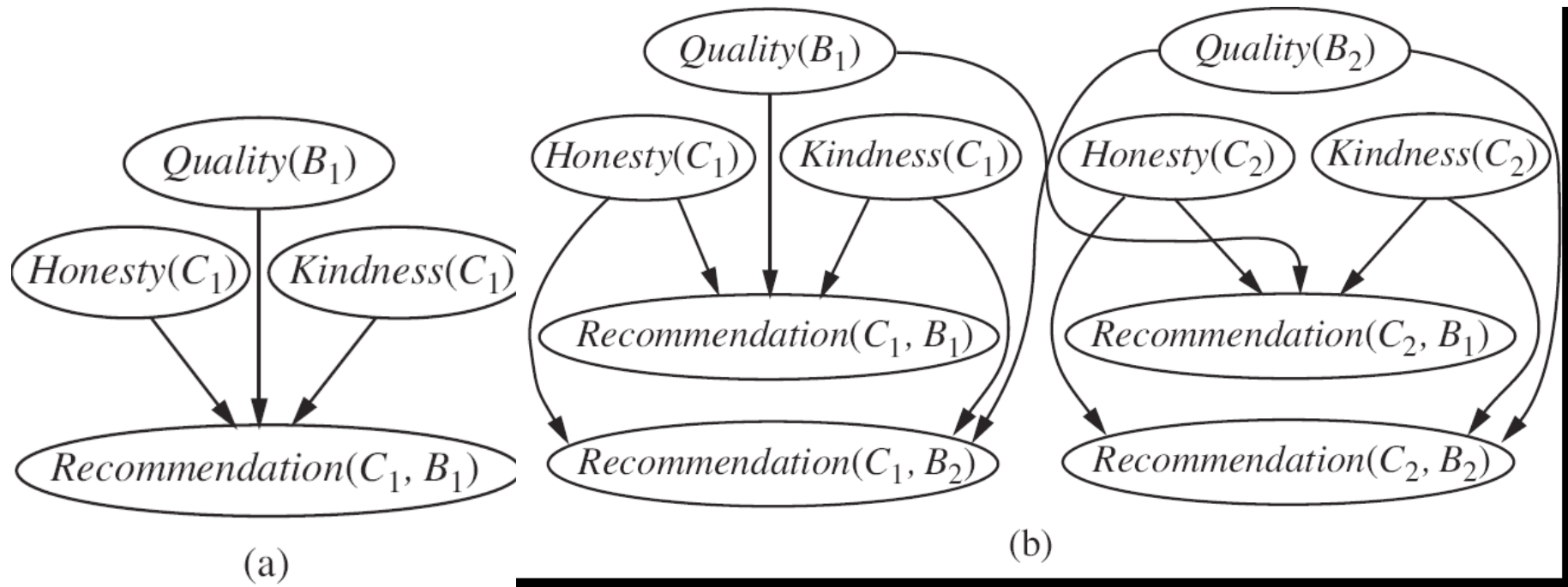
$$P(\text{world}) \propto \exp\left(\sum \text{weights of formulas it satisfies}\right)$$

(c) Probabilistic Relational models

- Probs specified on relations



Intuition for Prob. Relational models



A **customer** C_1 will / will not *recommend* a **book** B_1 depending on the book *quality*, and the customer *honesty* and *kindness*

When you have two customers and two books.....

drawn on white Board

Lecture Overview

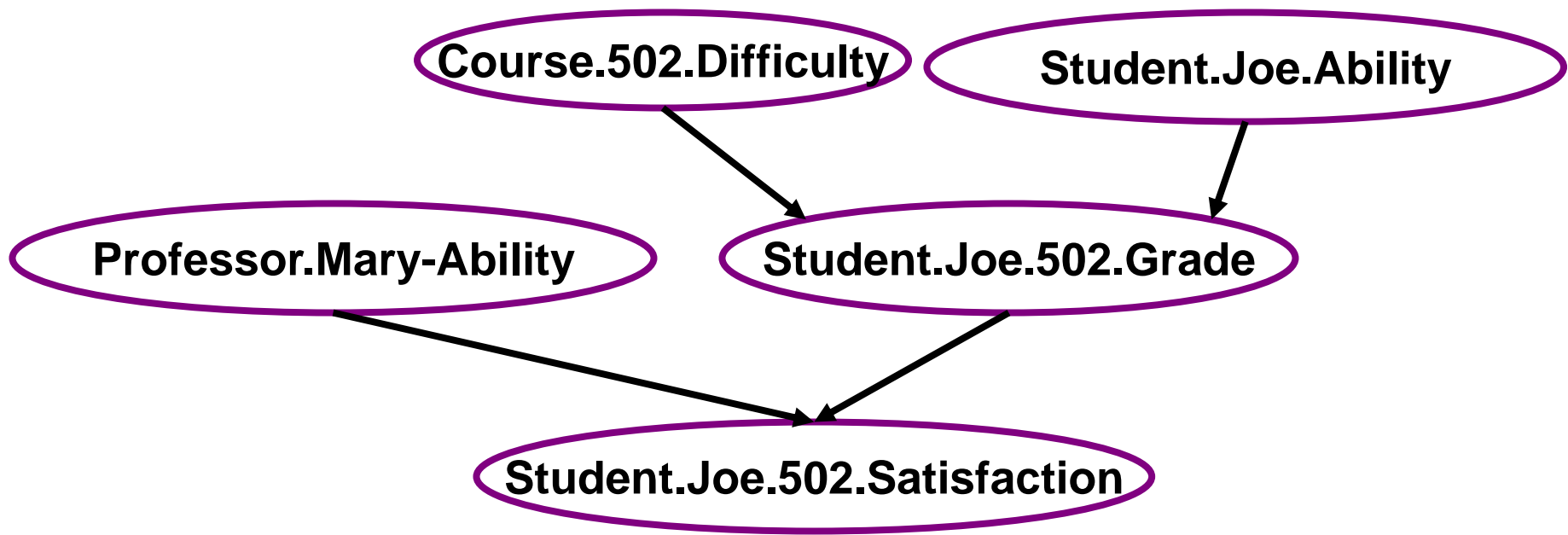
- **Motivation and Representation**
- Semantics of Probabilistic Relational Models (PRMs)
 - Classes and Relations
 - Attributes and Reference Slots
 - Full Relational Schema and its Instances
 - Fixed vs. Probabilistic Attributes
 - Relational Skeleton and its Completion Instance
 - Inverse Slot and Slot chain

Motivation for PRMs

- Most real-world data are stored in relational DBMS
- Combine advantages of relational logic & Bayesian networks:
 - natural domain modeling: objects, properties, relations;
 - generalization over a variety of situations;
 - compact, natural probability models.
- Integrate uncertainty with relational model:
 - properties of domain entities can depend on properties of related entities;
 - uncertainty over relational structure of domain.

Limitations of Bayesian Networks

A Bayesian networks (BNs) represents a pre-specified set of attributes/variables whose relationship to each other is fixed in advance.



How PRMs extend BNs?

1. PRMs conceptually extend BNs to allow the specification of **a probability model for classes of objects** rather than a fixed set of simple attributes
2. PRMs also allow **properties of an entity to depend probabilistically on properties of other related entities**

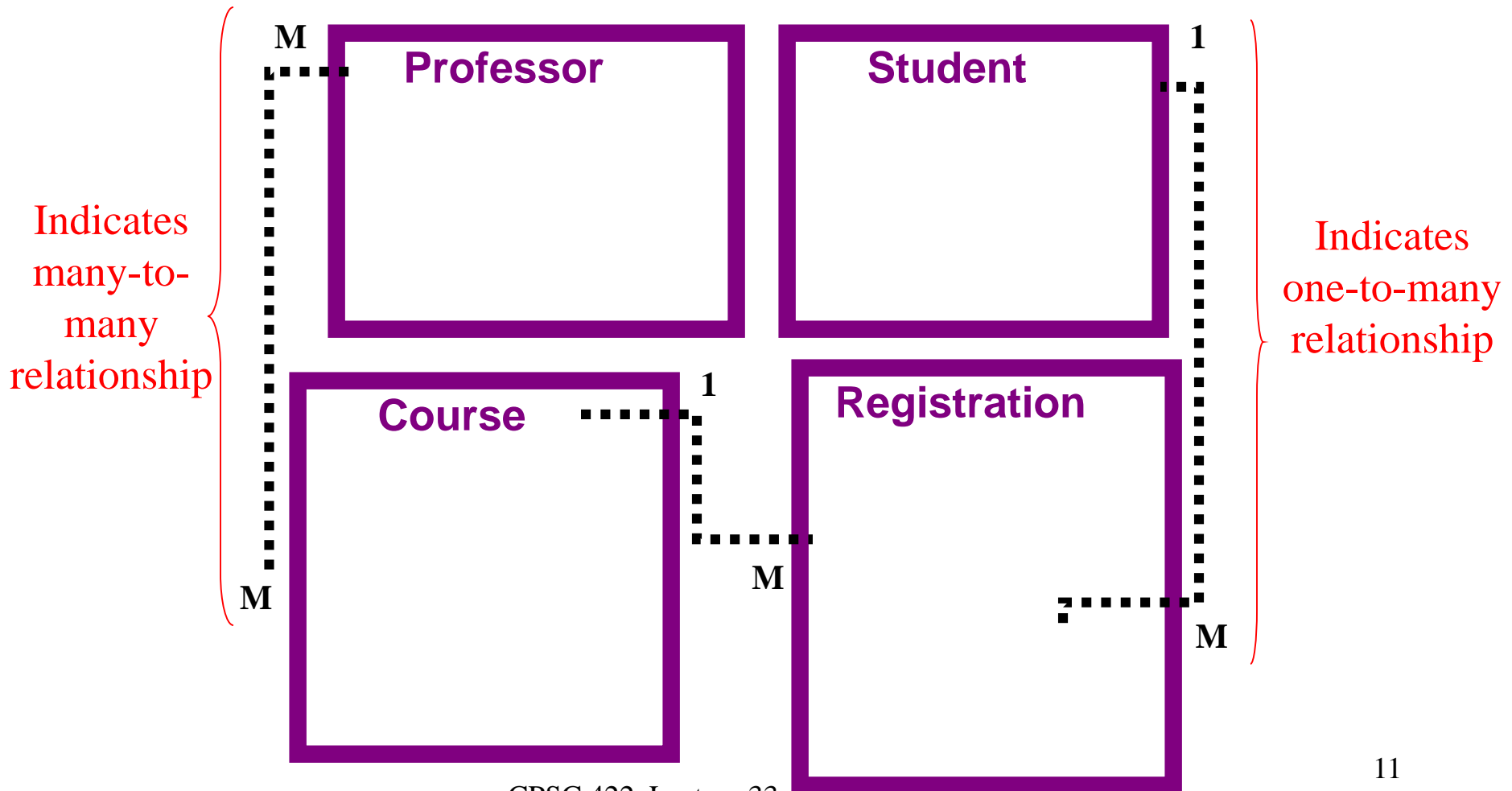
Lecture Overview

- Motivation and Representation
- **Semantics of Probabilistic Relational Models (PRMs)**
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Mapping PRMs from Relational Models

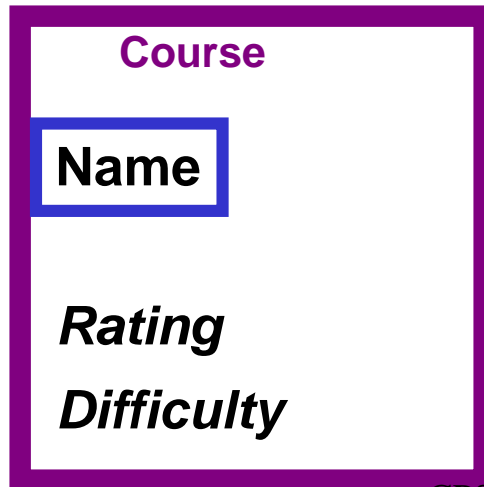
- The representation of PRMs is a direct mapping from that of relational databases
- A **relational model** consists of a set of *classes* X_1, \dots, X_n and a set of *relations* R_1, \dots, R_m , where each relation R_i is typed

University Domain Example - Classes and relations



Mapping PRMs from Relational Models: attributes

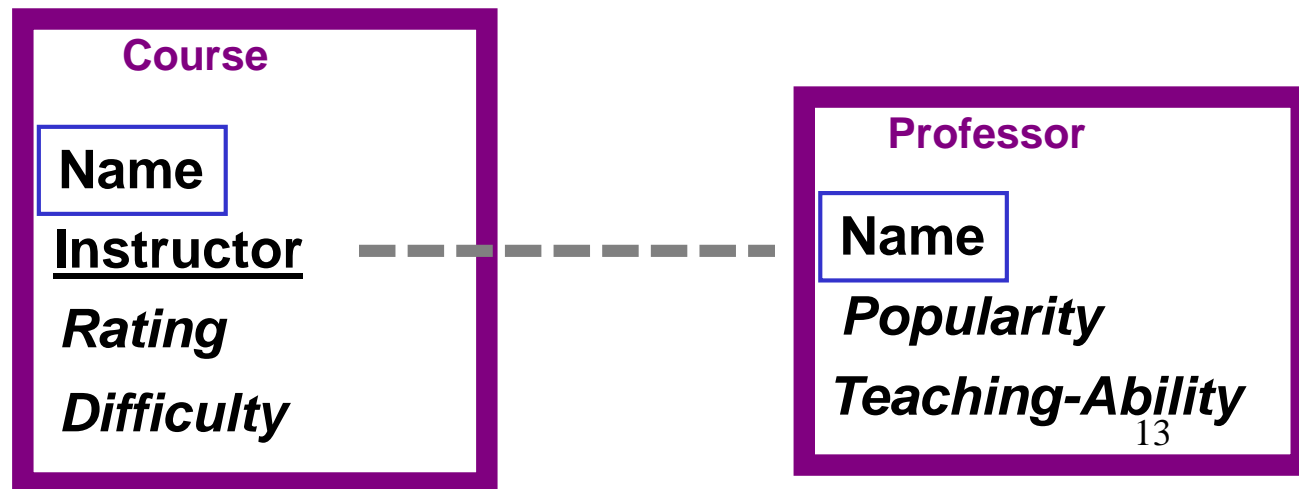
- Each class or entity type (corresponding to a single relational table) is associated with a set of **attributes** $\mathcal{A}(X_i)$ (at least one of which is a **primary key**)



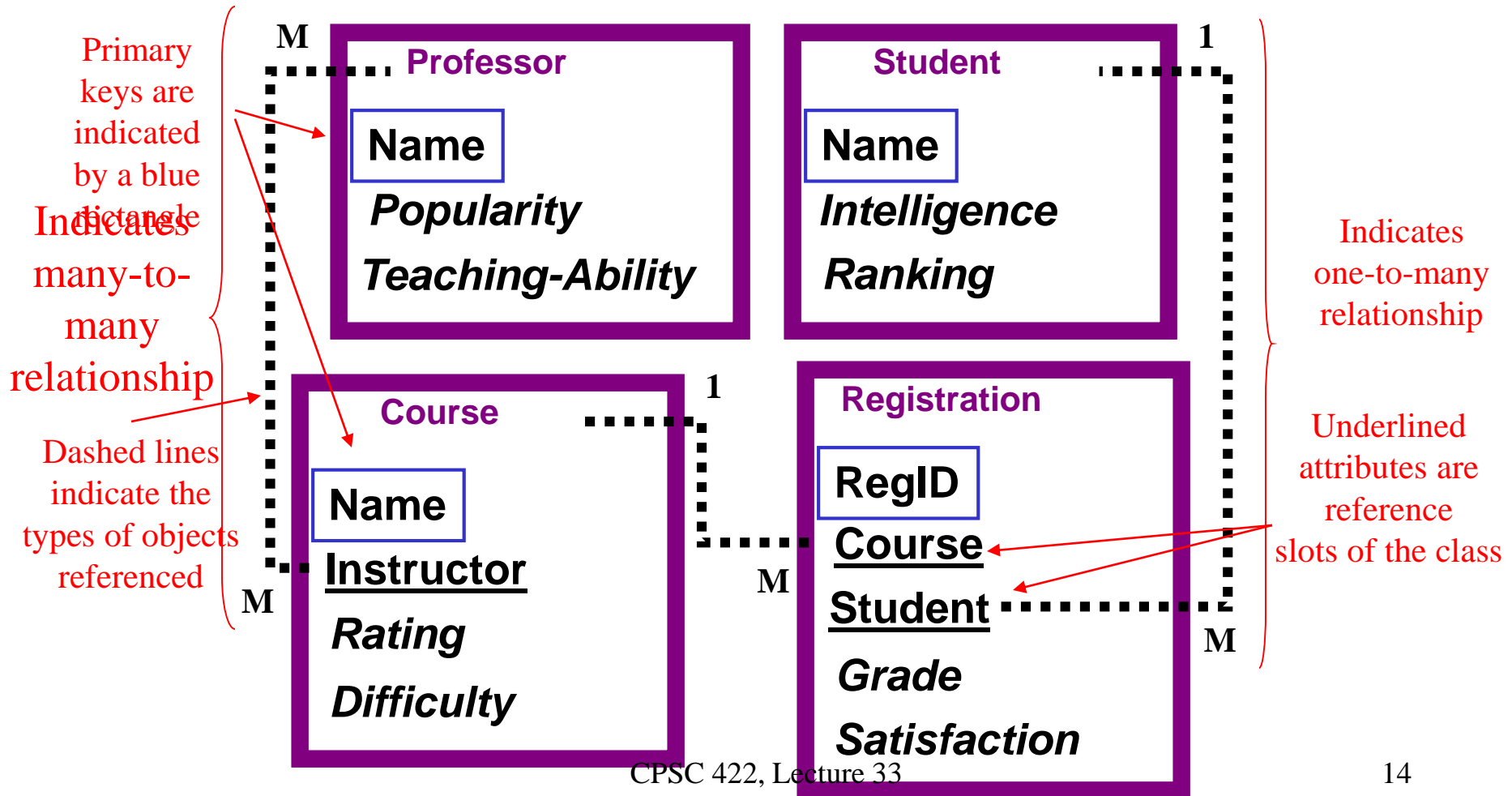
Mapping PRMs from Relational Models: reference slot

- Each class or entity type is also associated with a set of *reference slots* $\mathcal{R}(X)$
- correspond to attributes that are *foreign keys* (key attributes of another table)
- $X.p$, is used to denote reference slot p of X .

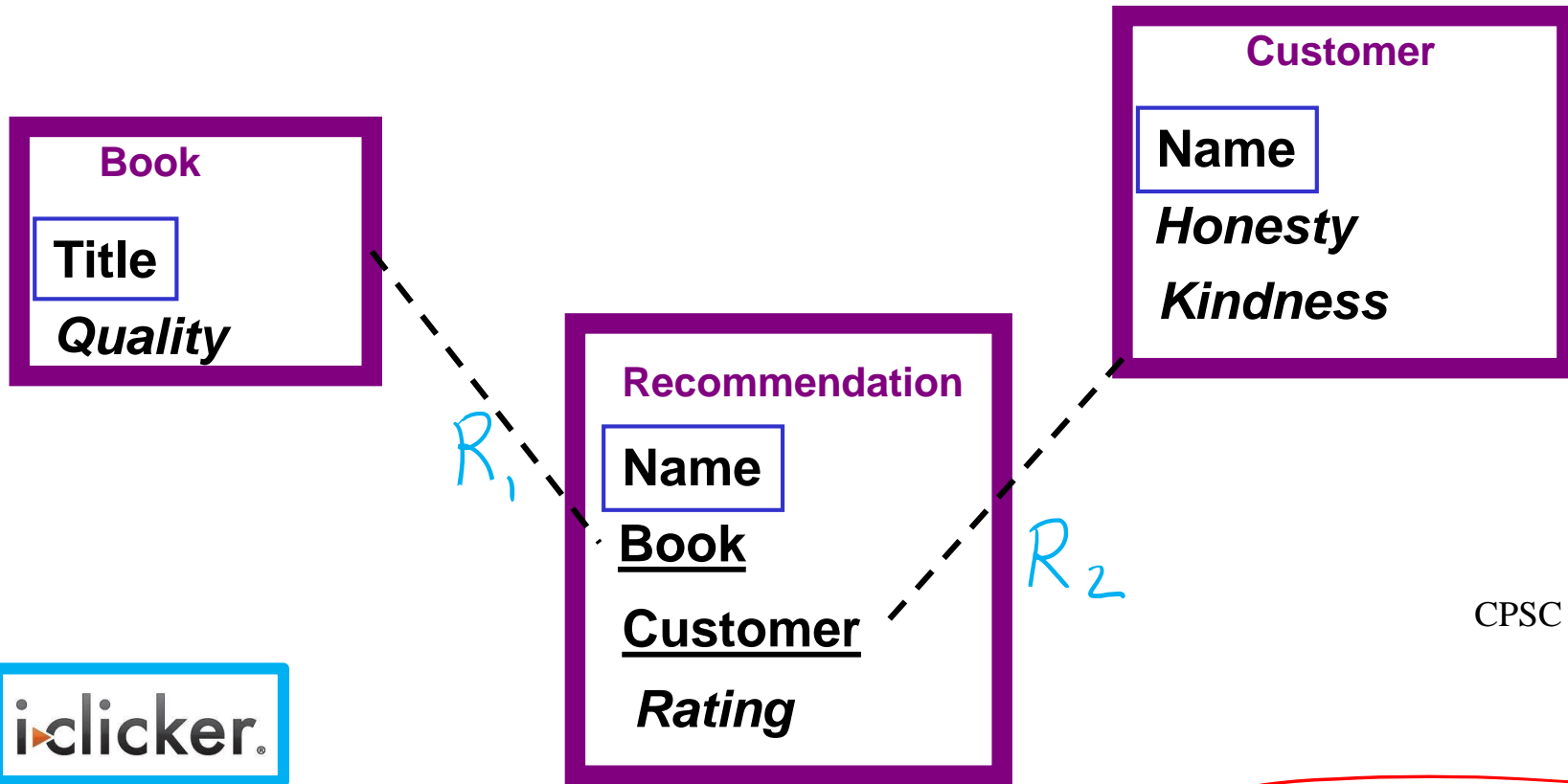
Course.Instructor



University Domain Example - Full Relational Schema



Book Recommendation Domain - Full Relational Schema



CPSC 422, Lecture 33

iclicker.

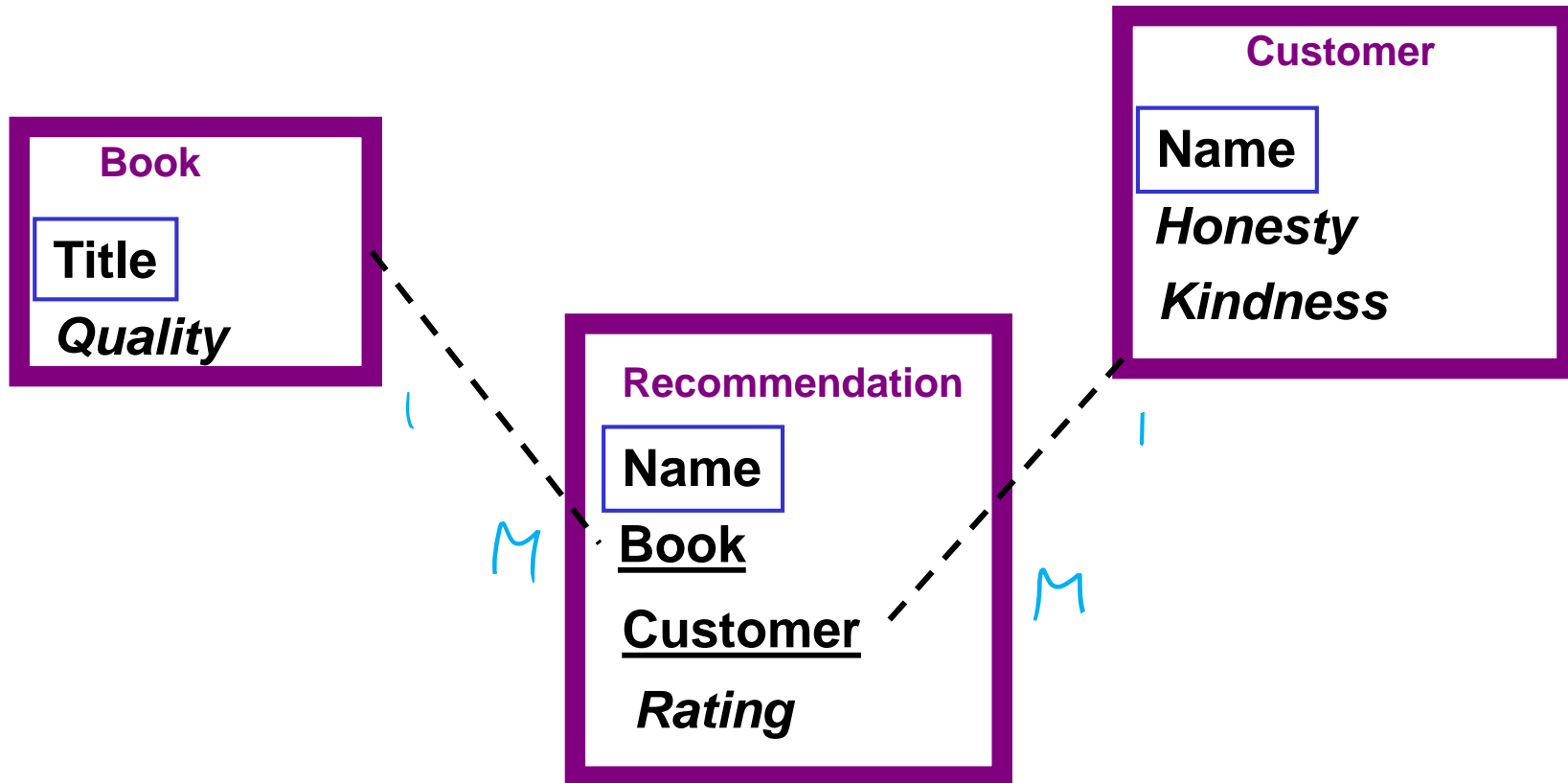
A. $R_1 (1:M), R_2 (M:M)$

B. $R_1 (M:M), R_2 (1:M)$

C. $R_1 (1:M), R_2 (1:M)$

D. $R_1 (M:M), R_2 (M:M)$

Book Recommendation Domain - Full Relational Schema



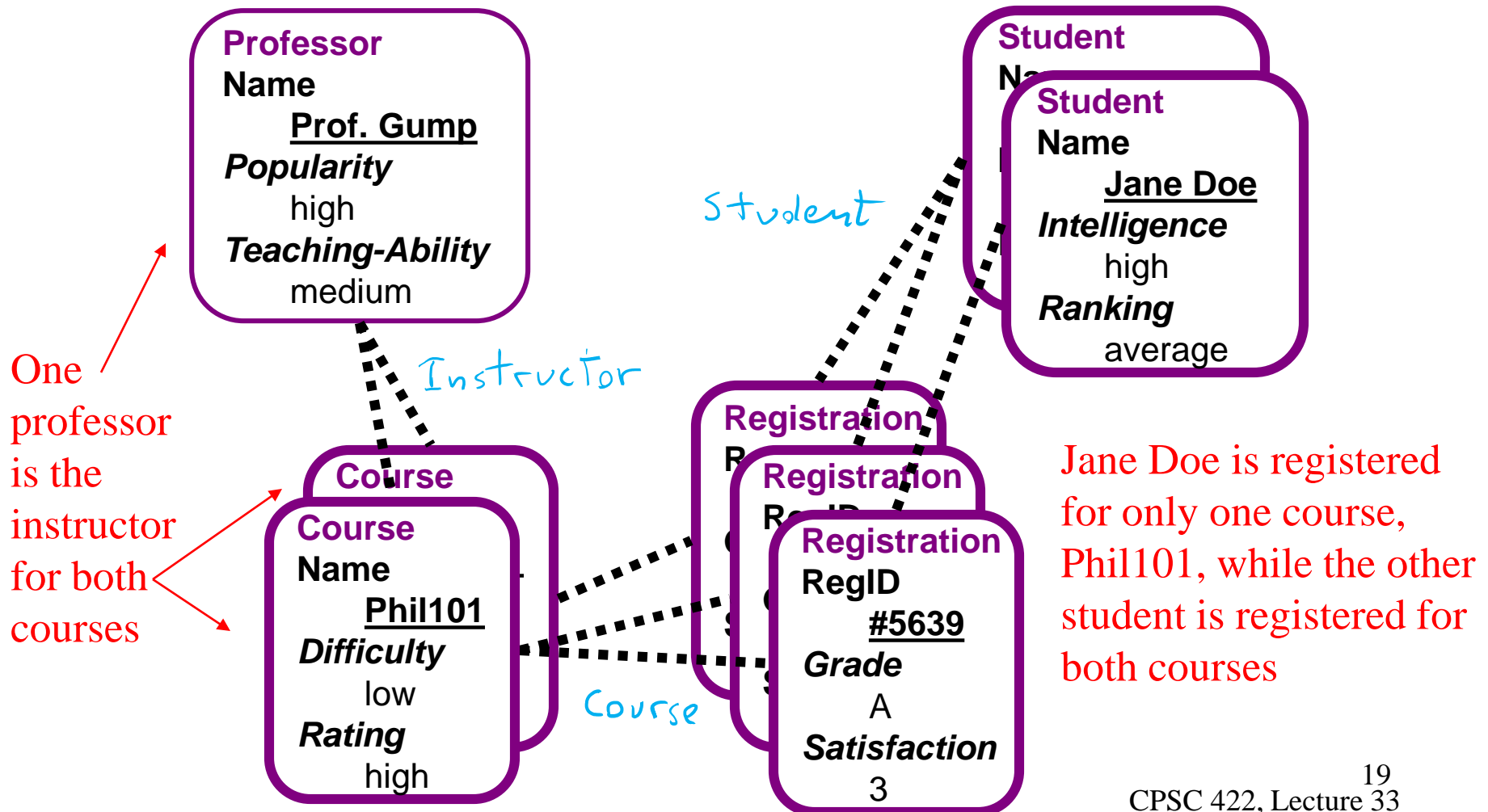
PRM Semantics: Attribute values

- Each attribute $A_j \in \mathcal{A}(X_i)$ takes on values in some fixed domain of possible values denoted $V(A_j)$. We assume that value spaces are finite
- Attribute A of class X is denoted $X.A$
- E.g., $V(\text{Student.Intelligence})$ might be $\{ \text{high}, \text{low} \}$

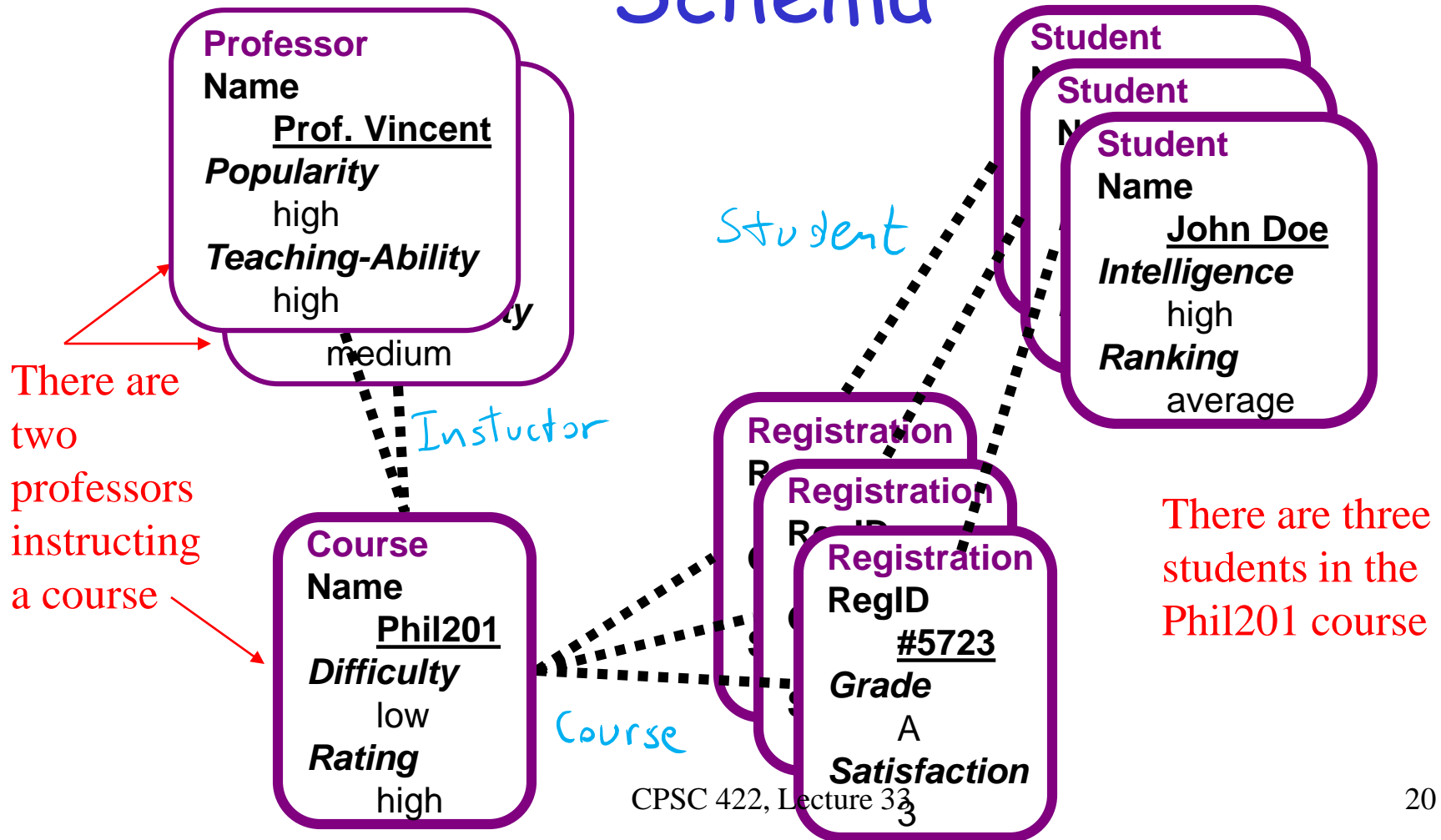
PRM Semantics: Instance of Schema

- An *instance* I of a schema/model specifies a set of objects x , partitioned into classes; such that there is
 - a value for each attribute $x.A$
 - and a value for each reference slot $x.p$

University Domain Example - An Instance of the Schema



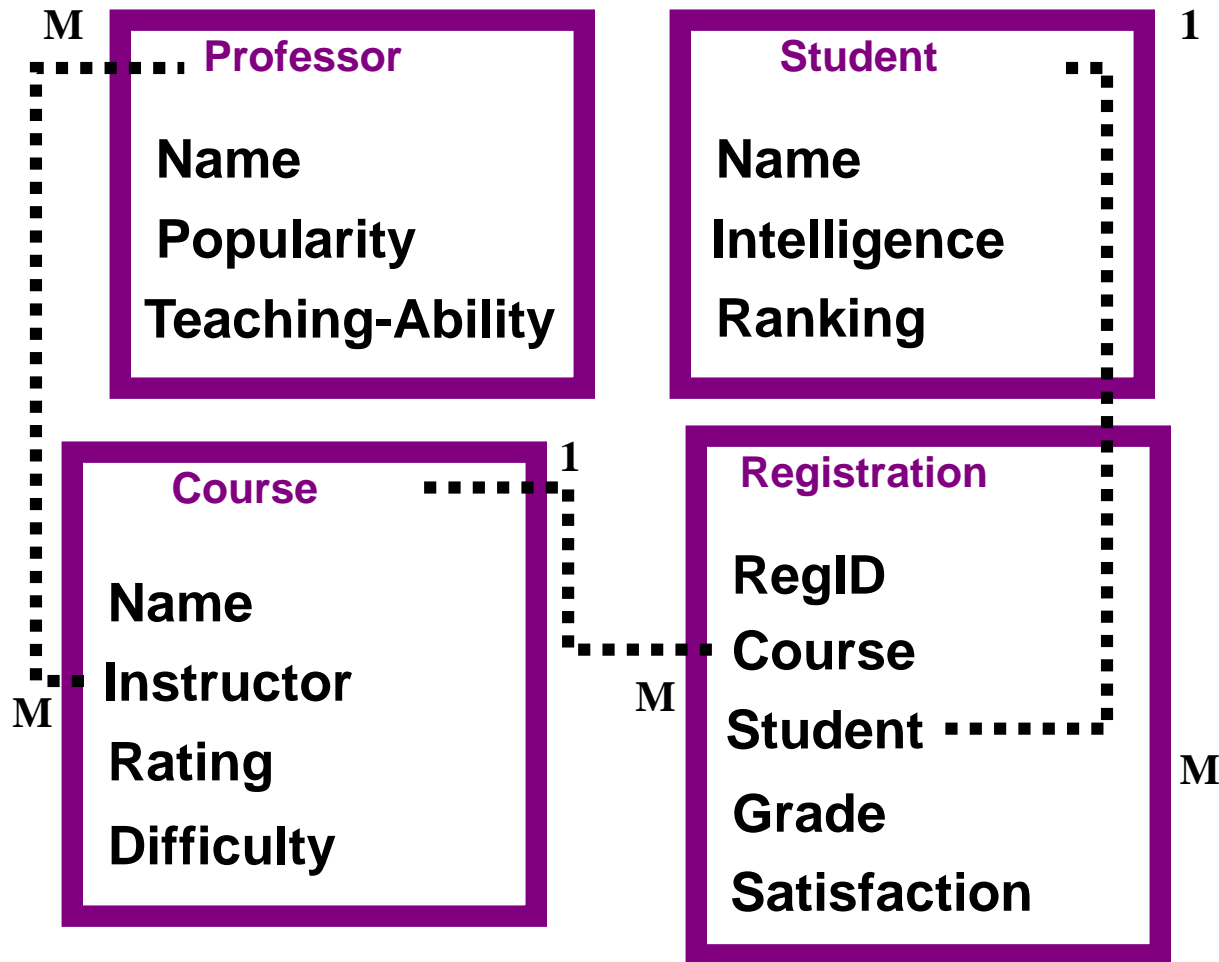
University Domain Example - Another Instance of the Schema



PRM Semantics: fixed vs. prob. attributes

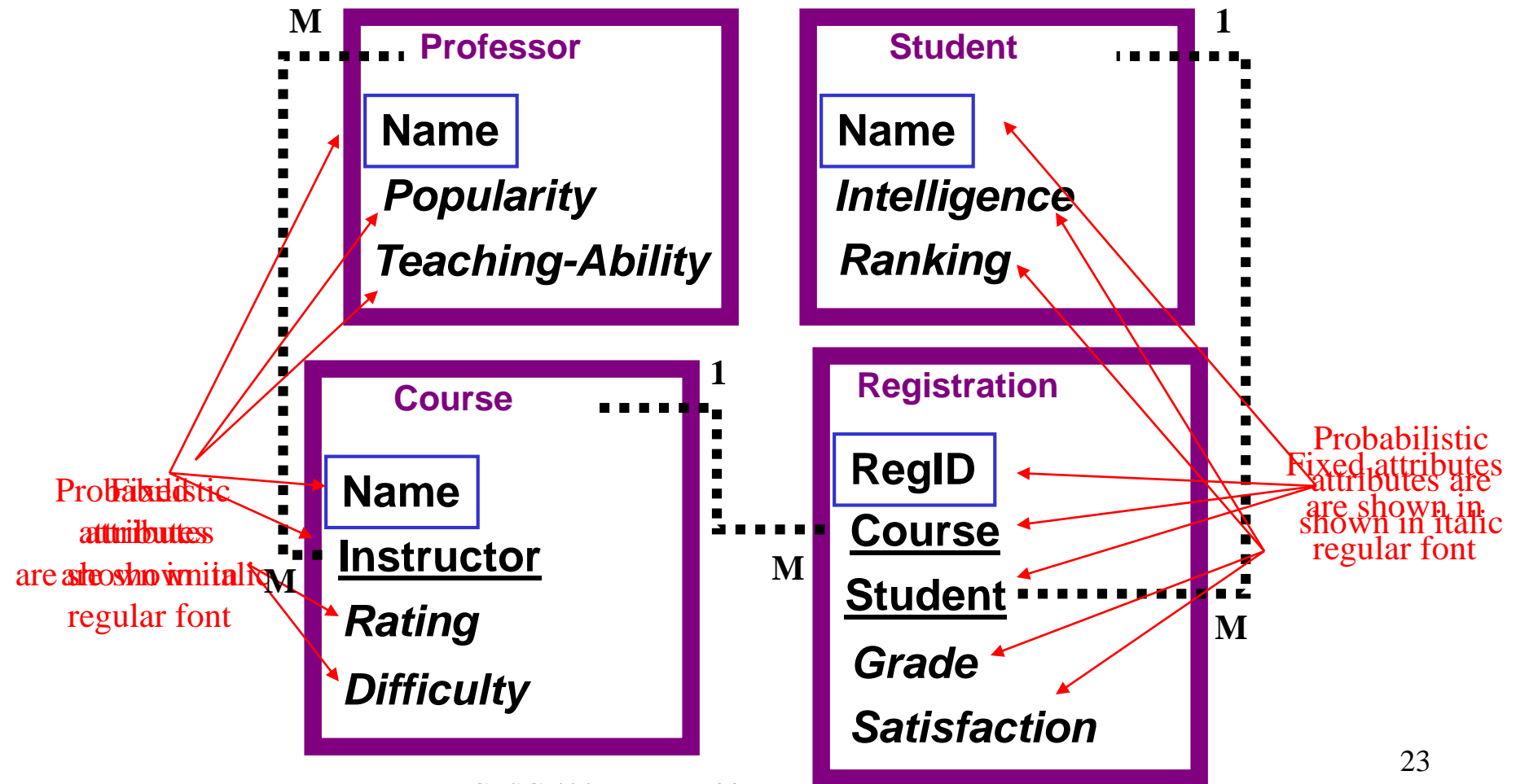
- Some attributes, such as *Name* or *Social Security Number*, are fully determined. Such attributes are labeled as **fixed**. Assume that they are known in any instantiation of the schema
- The other attributes are called **probabilistic**. We may be uncertain about their value

University Domain Example - fixed vs. probabilistic attributes



Which ones are fixed? Which are probabilistic?

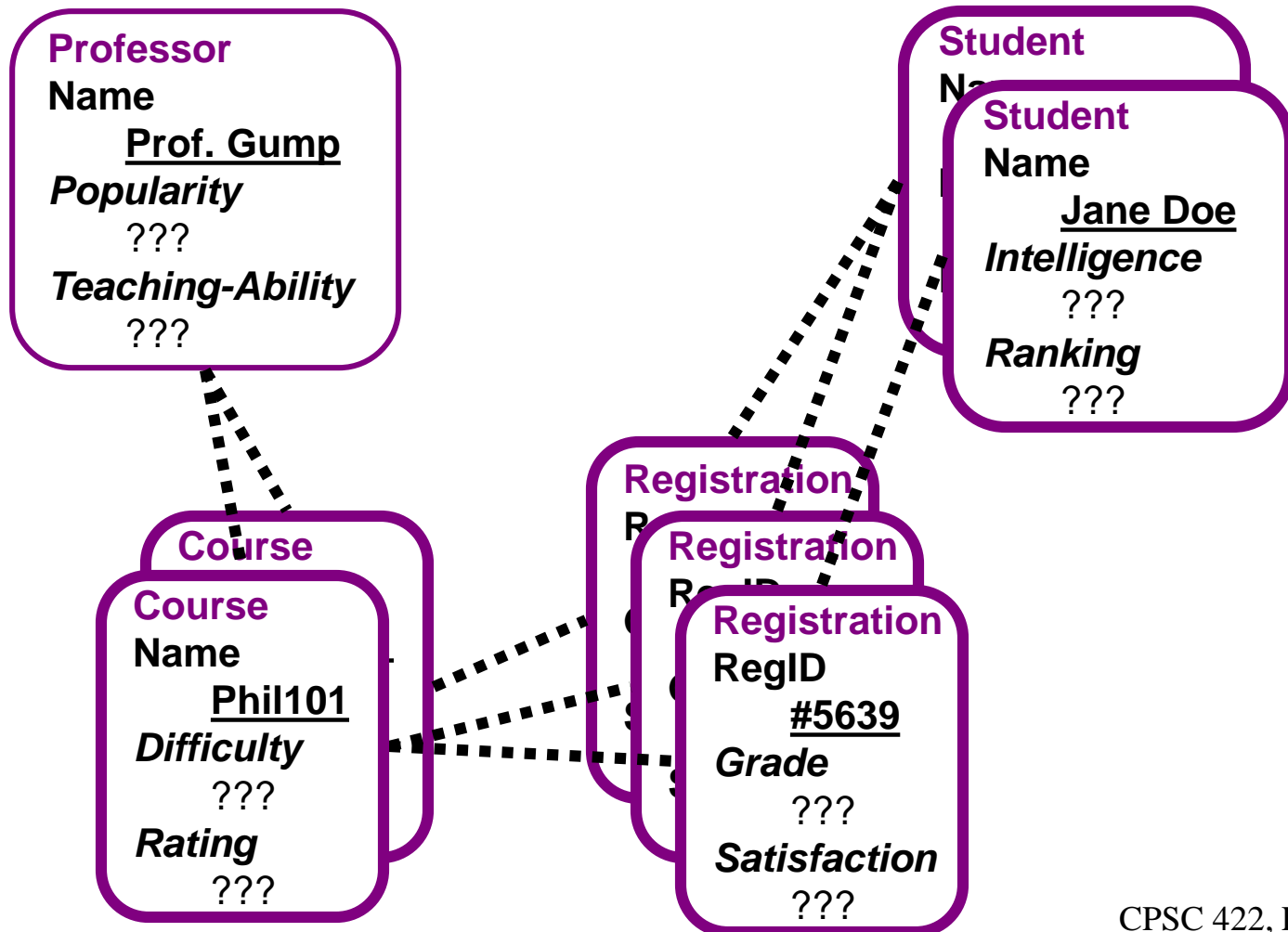
University Domain Example - fixed vs. probabilistic attributes



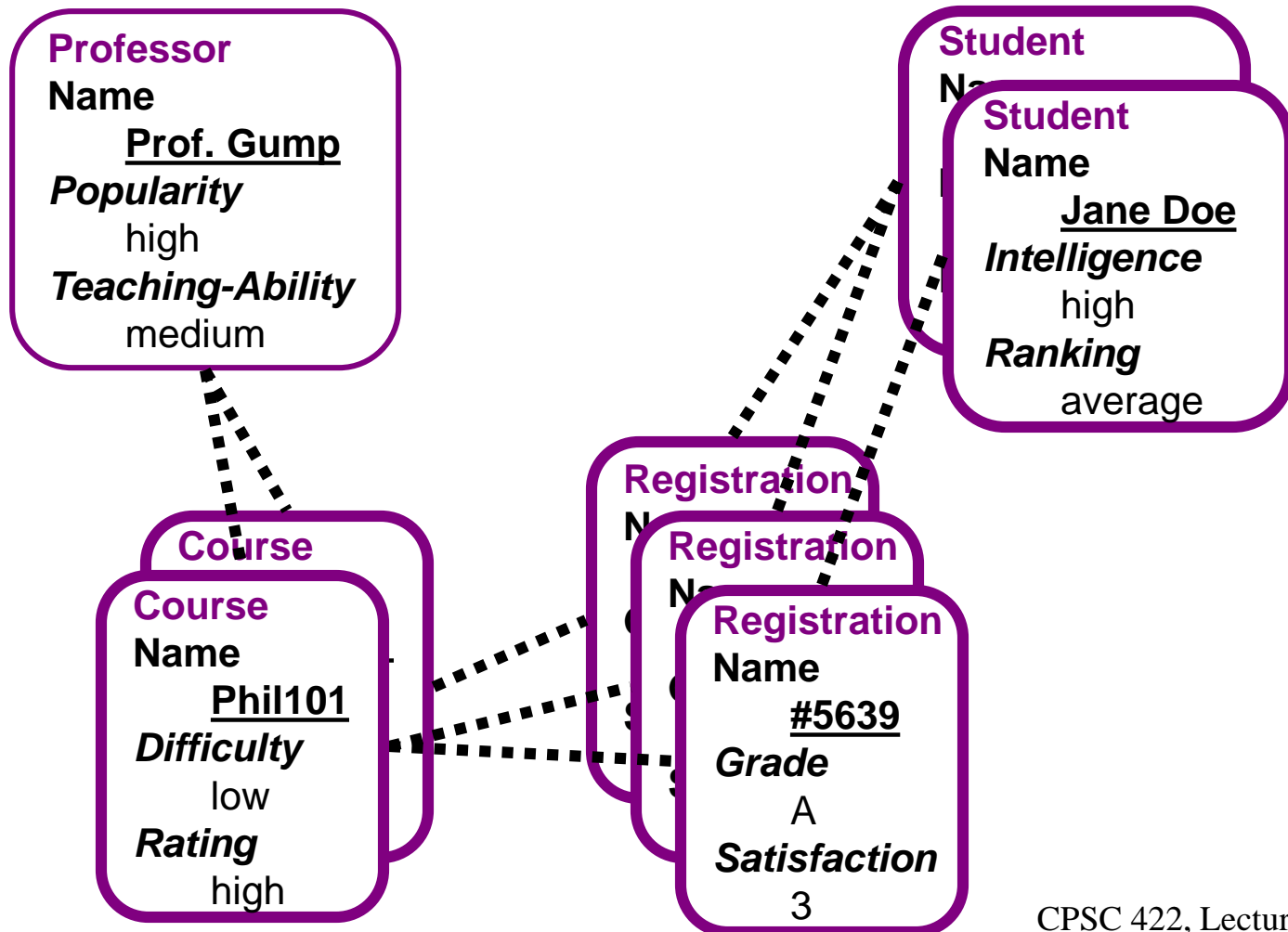
PRM Semantics: Skeleton Structure

- A **skeleton structure** σ of a relational schema is a partial specification of an instance of the schema. It specifies
 - set of objects for each class,
 - values of the fixed attributes of these objects,
 - relations that hold between the objects
- The values of probabilistic attributes are left unspecified
- A **completion** I of the skeleton structure σ extends the skeleton by also specifying the values of the probabilistic attributes
 - \Rightarrow possible world...

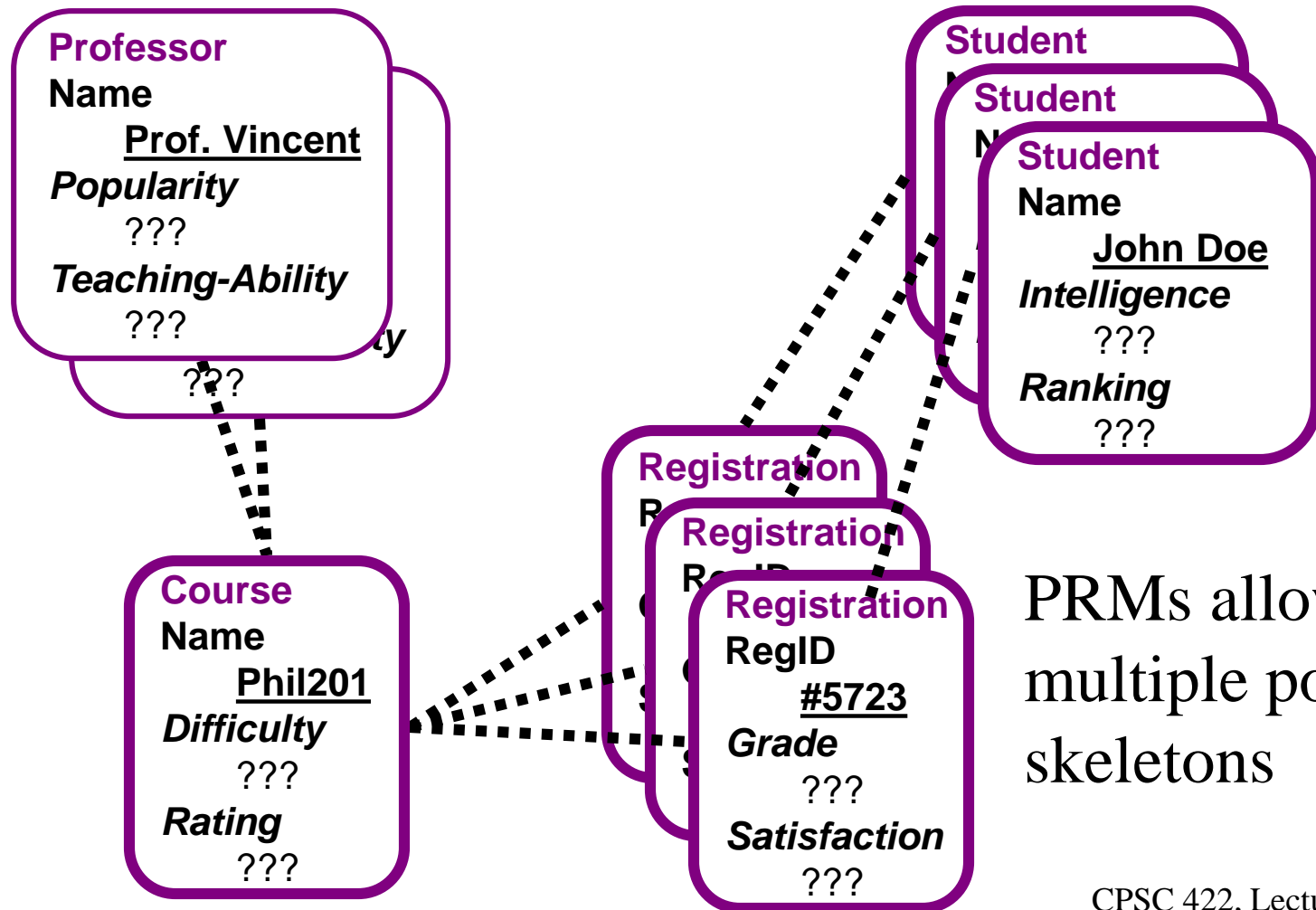
University Domain Example - Relational Skeleton



University Domain Example - The Completion Instance I

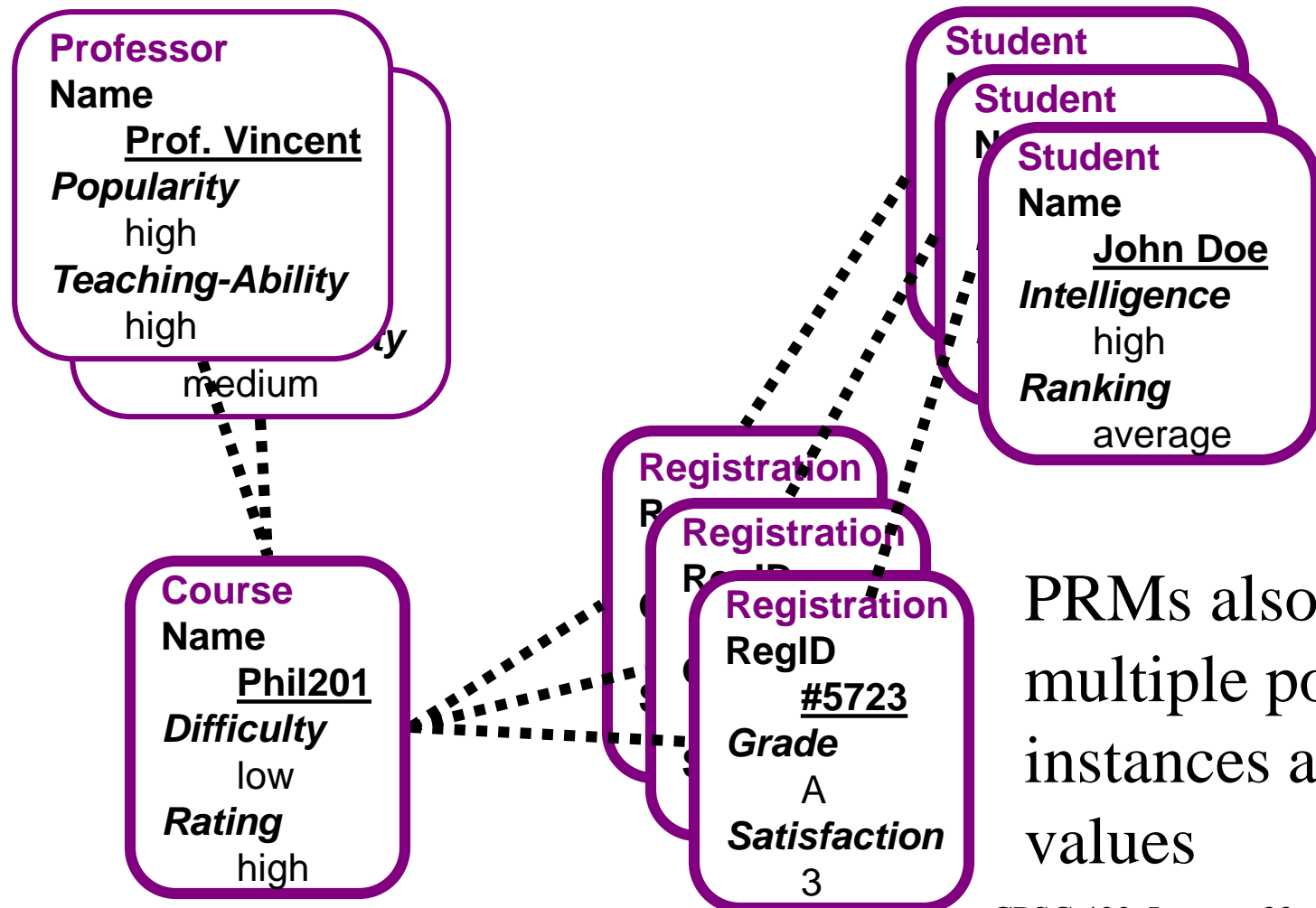


University Domain Example - Another Relational Skeleton



PRMs allow multiple possible skeletons

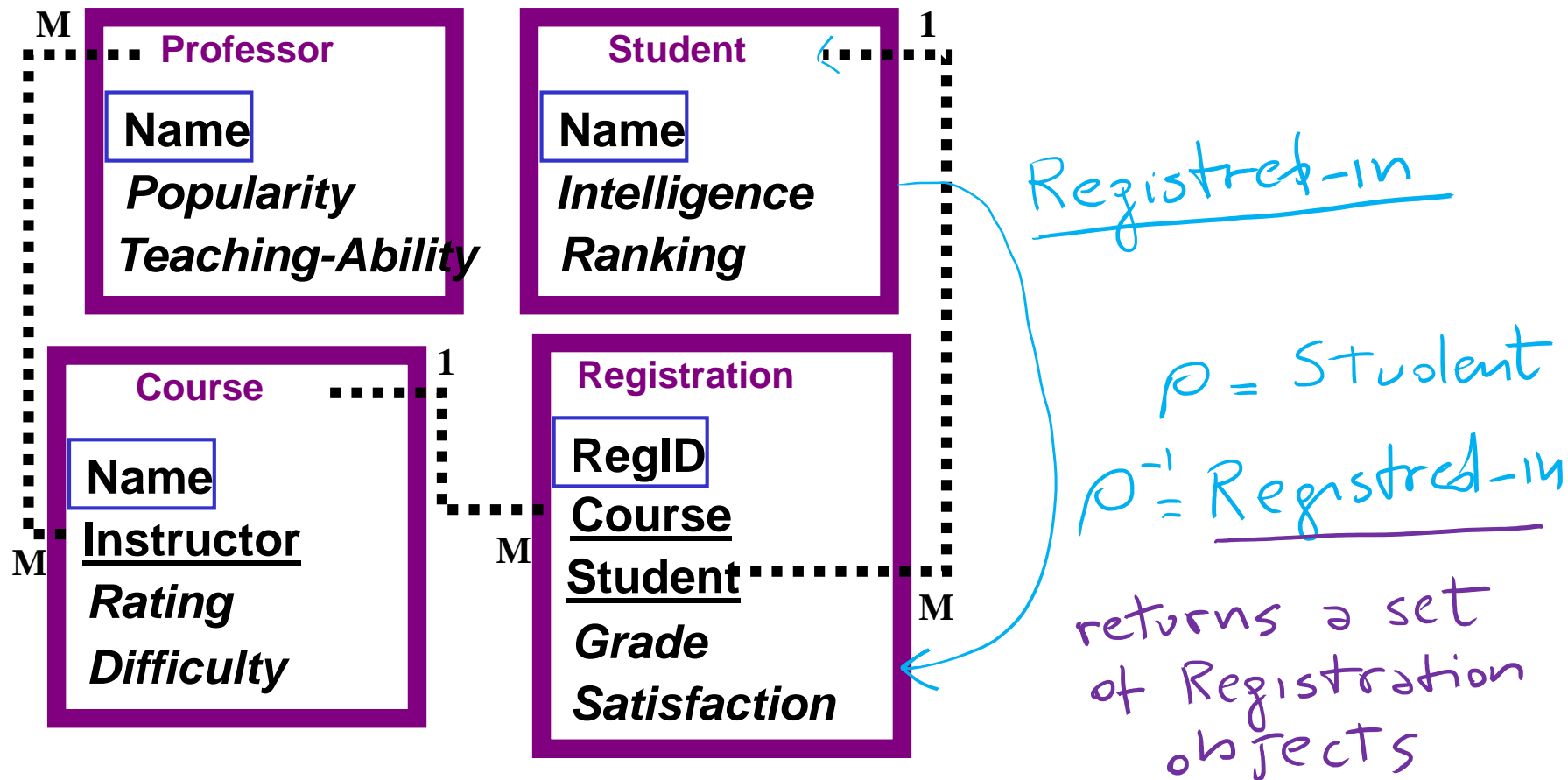
University Domain Example - The Completion Instance I



PRMs also allow
multiple possible
instances and
values

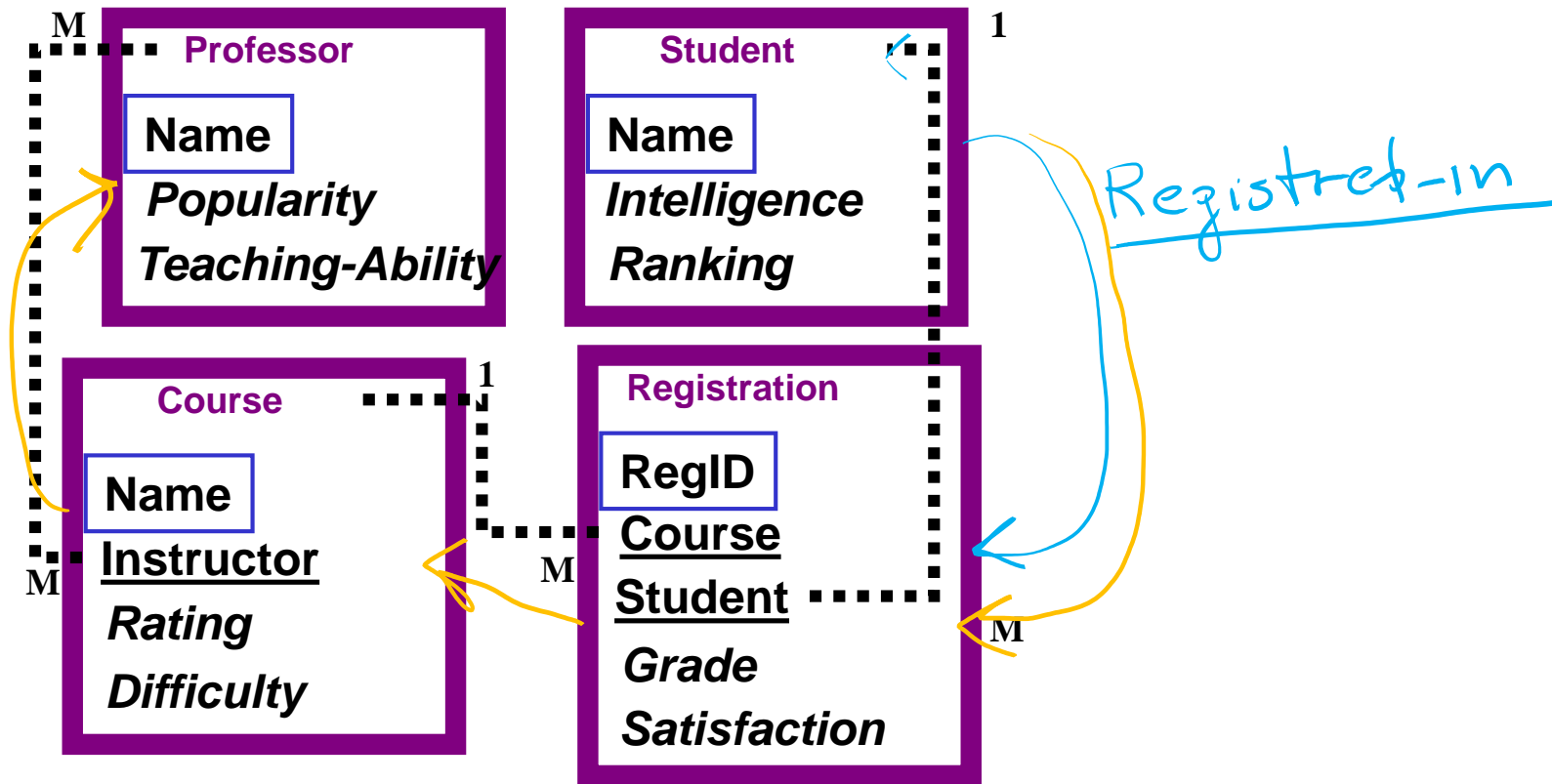
PRM Semantics: inverse slot

- For each reference slot ρ , we define an **inverse slot**, ρ^{-1} , which is the inverse function of ρ



PRM Semantics: slot chain

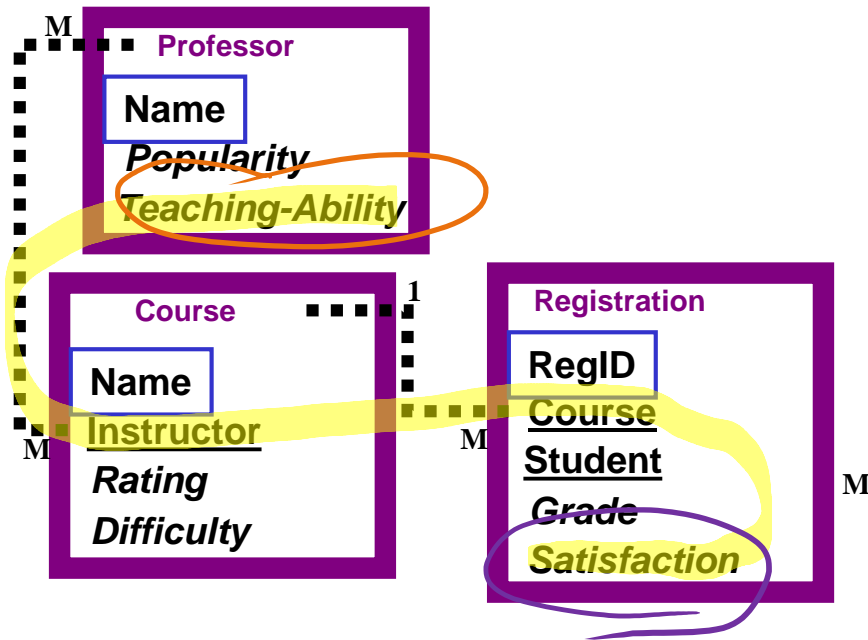
A **slot chain** $\tau = \rho_1 \dots \rho_m$ is a sequence of reference slots that defines functions from objects to other objects to which they are indirectly related.



Student.Registered-In.Course.Instructor
can be used to denote..... a student's set of instructors

Slot chains will allow us...

To specify probabilistic dependencies between attributes of related entities



... Course . Instructor ...

Learning Goals for today's class

You can:

- **Explain the need for Probabilistic relational model**
- **Explain how PRMs generalize BNs**
- **Define a Full Relational Schema and its instances**
- **Define a Relational Skeleton and its completion Instances**
- **Define an inverse slot and an slot chain**

Next class on Wed

Finish Probabilistic Relational Models

- Probabilistic Model
- Dependency Structure
- Aggregation
- Parameters
- Class dependency Graph
- Inference

Keep working on **Assignment-4**
Due Dec 2