# Jumping Ship in the Multiverse of Higher Education Berkeley

Cracking the course transferring problem in a university system using NLP



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# **1. Background, Contributions and Datasets**

SCHOOL OF

**INFORMATION** 

**Background**: Higher education institutes have been promoting socioeconomic mobility by supporting student transfer from 2-year community colleges to 4-year degree granting universities, such as in the California public post-secondary system and SYS1. However,

- **Course articulation**, *i.e.*, defining which course at one institution will count as credit for an equivalent course at another institution, requires efforts from both the schools, and it is intractable when attempting to manually articulate every set of courses at every institution pair.
- The Community College Research Center reported that the <u>transferring credit</u> problem is a major contributing factor to the dismal rates of graduation among transferring students.

Goal: To enhance the process of defining and maintaining course articulations for SYS1 by leveraging enrollment patterns and course catalog information.

**Datasets**: Real-life enrollment and catalog data from a large system of colleges and universities (SYS1) with 58 campuses

Catalog	~120K courses with titles and descriptions	ei
Enrollment histories	~16.7M enrollments of 1.27M students from 2014 Fall to 2019 Spring (anonymous student ID, year, semester, course ID etc.)	- in
Mapping	~128K course articulation pairs (source course, source university,	sr

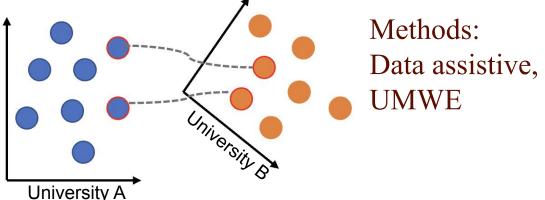
**Problem definition**: Given a course in university A, find the most similar course

2. Methodologies

in university B. <u>Two ways of modeling:</u>

1. Learn course embeddings within universities separately and align them to the same vector space.

2. Learn course embeddings across



#### Data assistive (Pardos et al., 2019):

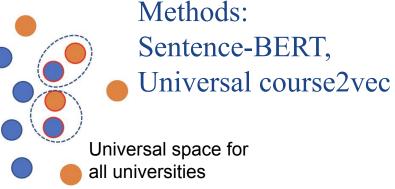
- Learn course embeddings from each university and then learn a linear transformation to map them into the same vector space based on articulation pairs.

### **UMWE (Unsupervised multilingual word** embeddings, *Chen and Claire*, 2018)

Use adversarial learning to map all the words in different languages into the same vector

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universities directly

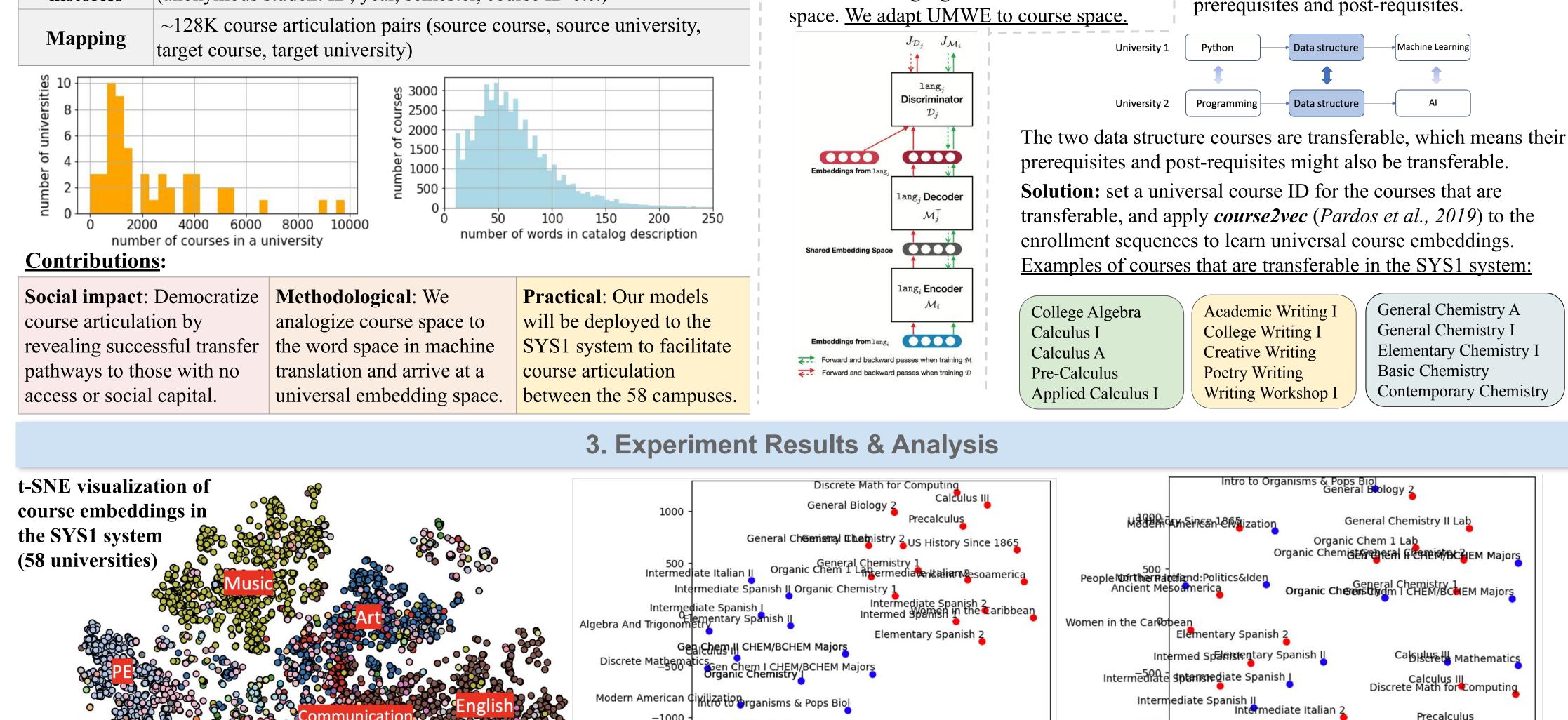


## Sentence-BERT (Nils and Iryna, 2019):

- a siamese BERT network to derive semantically meaningful sentence embeddings. We applied it to course catalog and generated course catalog embeddings.

#### Universal course2vec (proposed):

- Assumption: similar courses in different universities might also have similar prerequisites and post-requisites.



The courses in the two universities are initially separated. After the model was trained, the courses are blended to the same vector space where similar courses across universities became closer.

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Experiment results (Recall) on predicting the course mappings (5 fold cross validation)

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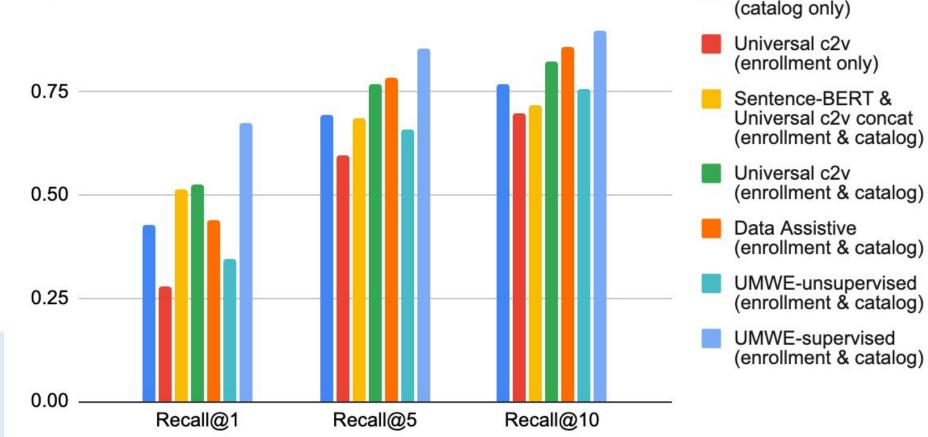
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Courses with the same subject are clustered together (the same color). Similar subjects are also closer to each other, for example, Communication is in the midst of Art, English, and Business; Chemistry is closer to Biology and Physics than other social related courses.



method performed the best among all the methods. - Combining the (enrollment & catalog) catalog and (enrollment & catalog) enrollment data together boosted (enrollment & catalog) the performance to a large degree.

Precalculus

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- The UMWE

supervised

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Intermediate ItaliaAlgebra And Trigonometry

Sentence-BERT

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