

Unveiling the Blue Text Analytics machine learning project

- *Presented by : Zelbery Bedard*

AGENDA

1. Overview of BTA concepts
2. Discussion of Natural Language Processing (NLP) applied to BTA

Shift in habits...



ABC UNIVERSITY **A 1.** Course-Instructor Survey

COURSE UNIQUE NUMBER
 1 2 3 4 5 6 7 8 9 0

INSTRUCTOR'S NAME AND SURNAME

INSTRUCTIONS
 Enter your instructor's name and the course information in the box to the left and right.

14. Overall, the instructor was A B C D E F G

15. Overall, the course was A B C D E F G

YOU MAY CONTINUE COMMENTS ON THE REVERSE SIDE



Open-ended feedback

Digital Quantity
 Quality Contextual

Course Component Evaluation 2018 for 401791 African Cultures and Societies

The course was intellectually challenging

- Strongly Disagree
- Disagree
- Neutral
- Agree
- Strongly Agree

The instructor was well-prepared for class

- Strongly Disagree
- Disagree
- Neutral
- Agree
- Strongly Agree

How could the learning experience of this course be improved?

Ditch the power points, or at least only use them as an aide. Focus on your lecture style. You describe the texts—a good professor synthesizes the material, gives insight. It is frankly somewhat offensive that Dr. Thompson does not take more care in her lectures. She does very little exegesis. This is high school level teaching.

Format lab 1 differently. 4 people doing different parts of the incubation doesn't make sense when we should all have a comprehensive understanding of the process, also sometimes the manual was confusing in terms of where to put what in which tube, like remove and put in a tube but it doesn't specify if it goes in the same one as the other stuff or if its just extra.

BLUE TEXT ANALYTICS

BTA is
Efficient,
insightful
analysis on
large volumes
of open-ended
feedback



BTA Dictionary

A sampling of BTA

Themes and synonymous expressions

APPROACHABLE:

approachable, calm, easy to talk to, laidback, good natured, is open to, open minded, relaxed

UNAPPROACHABLE:

aloof, chilly, detached, didn't care, distant, frigid, haughty, heartless, indifferent, intimidating, not easy to talk, overbearing, standoffish, stuck up, taciturn, too formal, unapproachable, unresponsive, unsympathetic

CLEAR:

accurate, clear, exact, explicit, precise, straightforward

UNCLEAR:

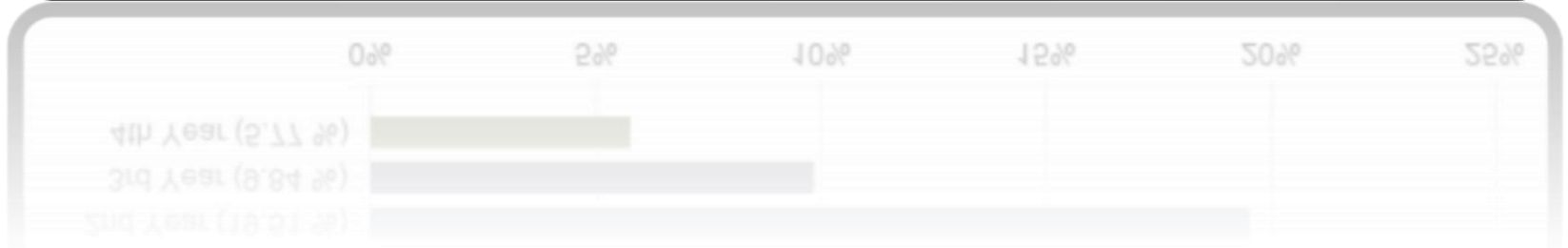
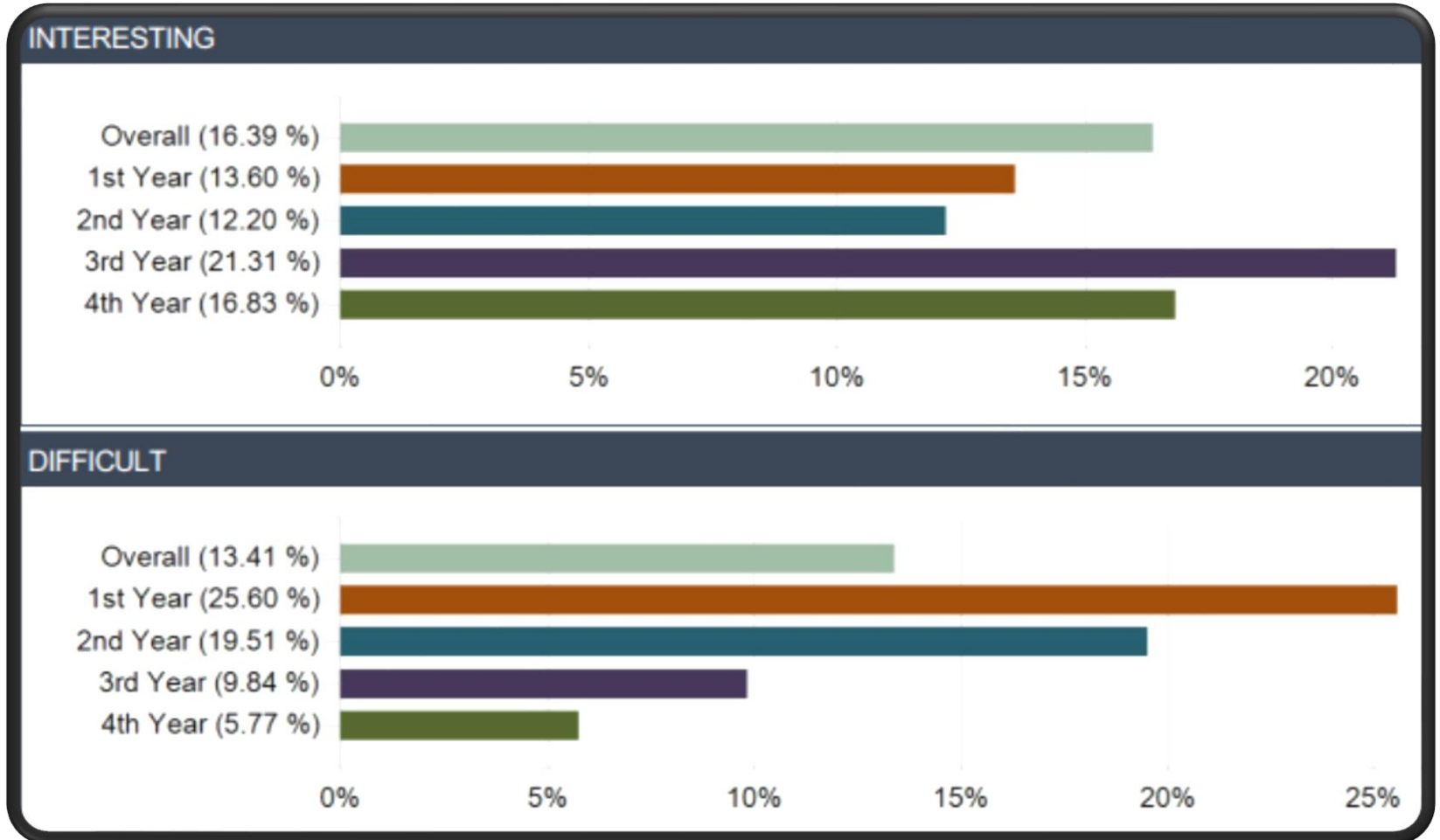
ambiguity, ambiguous, baffled, confusing, bewildered, blur, cryptic, enigmatic, vague, foggy, fuzzy, hazy, illegible, imprecise, in the dark, incomprehensible, inconsistent, lack of clarity, lost, muddy, not detailed, not explicit, obscure, opaque, puzzled, unclear, not well described

BTA Analysis

Themes compared across demographics

Comparing:

- a) Top 2 Attributes mentioned in the program
- b) Student year of study demographic data



BTA Analysis

Themes compared across a rating question

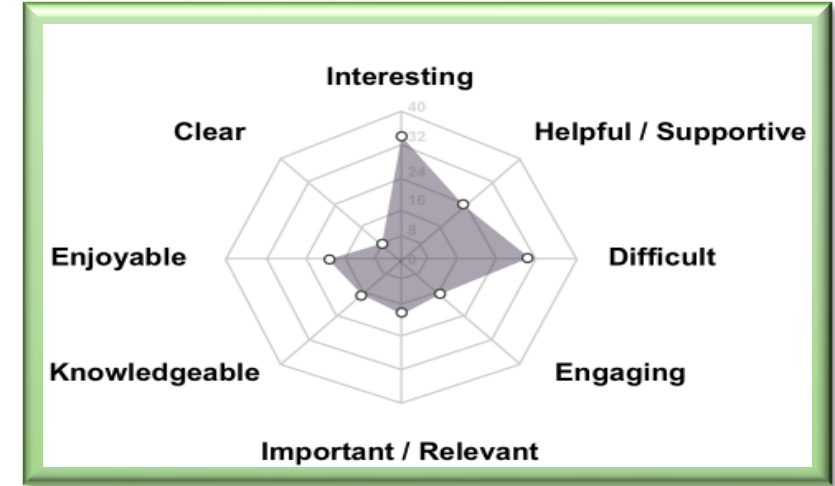
Comparing:

- a) Top 8 Attributes mentioned in the program
- b) "Overall what did you think of this course?" rating question

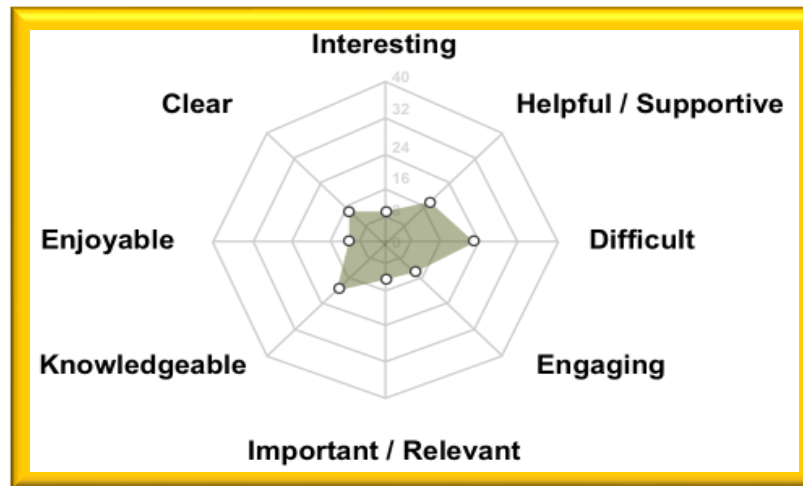
Very Good



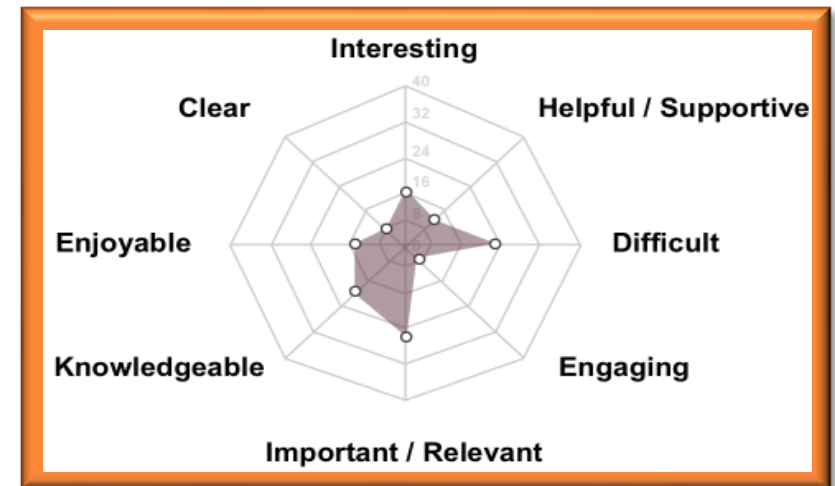
Good



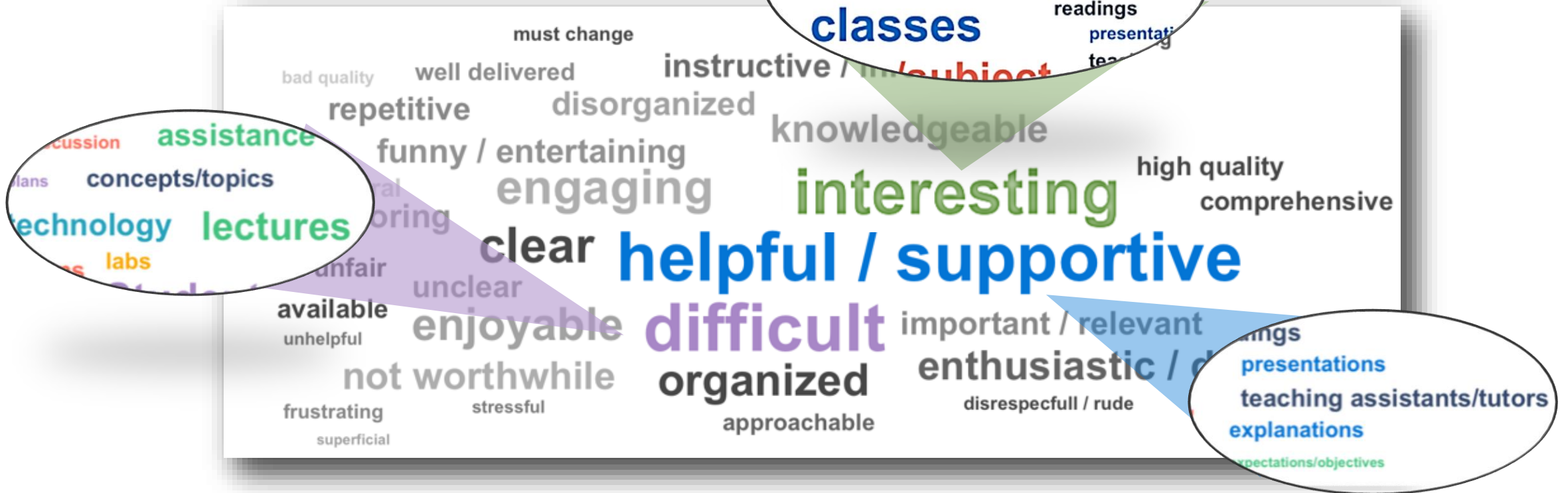
Fair



Poor



What's next for BTA?



What is *difficult*?

Who or what is *helpful / supportive*?

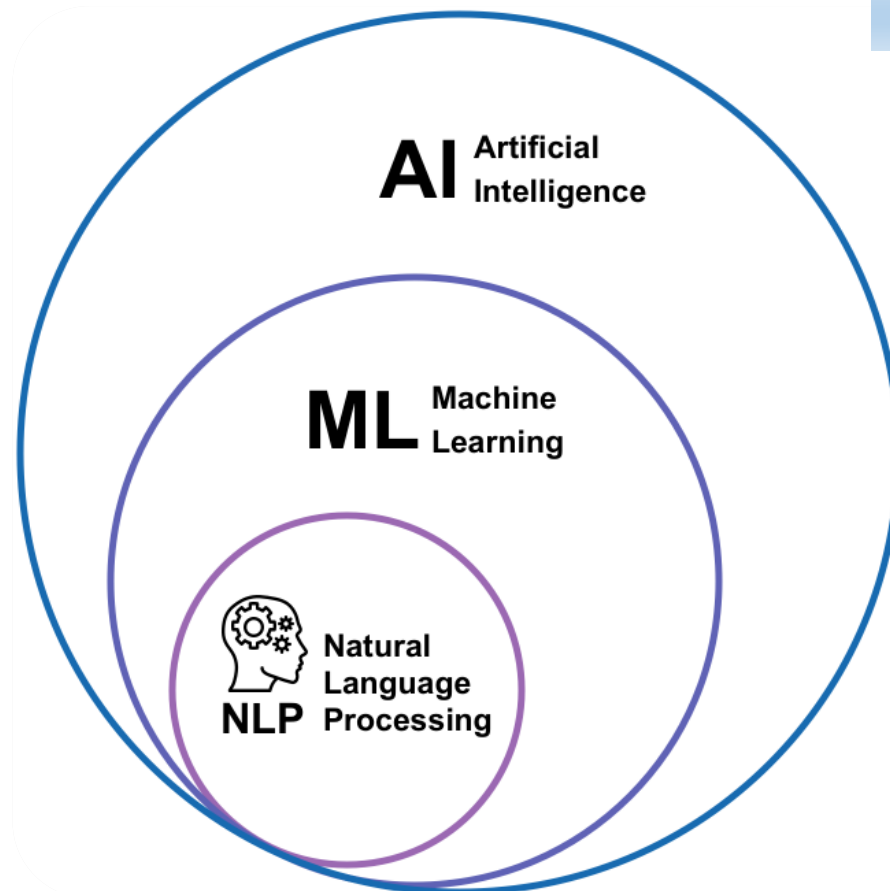
Natural Language Processing

How AI, ML and NLP work together

Machine Learning has the potential to detect patterns beyond hand written rules and validation.

We aim to improve upon handling of

- Misspellings
- Double and triple negations
- Indirect and passive phrases
- Regional differences in language



AI

Teaches system to do **intelligent** things

ML

Teaches systems to do **intelligent** things that can **learn** from experience

NLP

Teaches systems to do **intelligent** things, **learn** from experience and **understand** human language

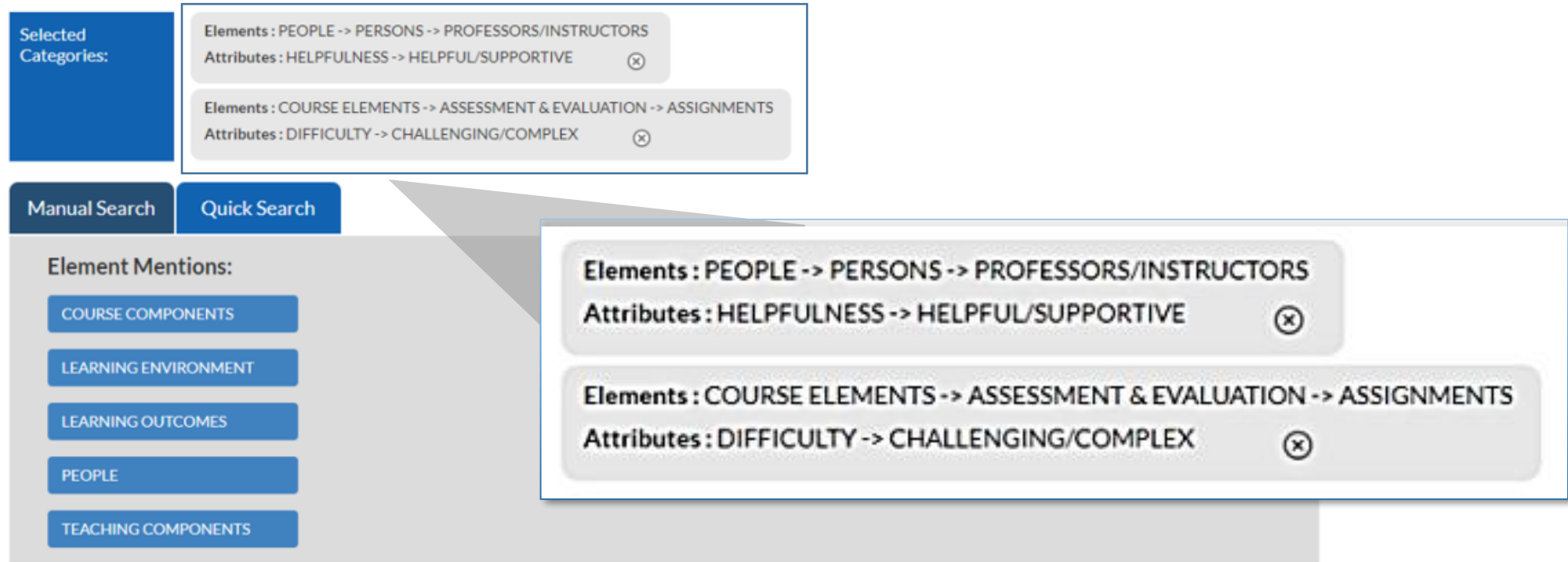
understand human language

Detect The Themes

- Goal: We want the model to detect the sentiment and the themes found in comments related to Teaching & Learning experiences.
- Approach: Supervised
- Data collection:
 - Coming from the relevant fields, i.e. T&L
 - Containing regional variations
- Data processing:
 - Manual tagging of the data with sentiment and themes
 - Divided into training and validation data sets
- Building of the model and the API

Tagging

Professor ~PERSON~ is always willing to help and even stayed after class a few times. He is super nice and clearly understands the material and presents in a way that helps me understand it as well. I was struggling with my assignment and he helped me work thought it.



The screenshot displays a tagging interface with the following components:

- Selected Categories:** A blue box on the left containing two category tags:
 - Elements : PEOPLE -> PERSONS -> PROFESSORS/INSTRUCTORS
 - Attributes : HELPFULNESS -> HELPFUL/SUPPORTIVE (with a close icon)
- Manual Search / Quick Search:** Two tabs at the top of the main content area.
- Element Mentions:** A list of five blue buttons: COURSE COMPONENTS, LEARNING ENVIRONMENT, LEARNING OUTCOMES, PEOPLE, and TEACHING COMPONENTS.
- Tagging Callout:** A large grey callout box on the right, connected to the 'PEOPLE' button, showing the same two category tags as the 'Selected Categories' box.

1st Year Students

2nd Year Students

All Students



- Now let's uncover:
- Top poor experiences
 - Top good experiences

Let's uncover: What students mentioned by year of study

Discovering themes automatically

- Goal: We want the model to find potential new themes automatically, and a committee will decide whether to include the newly discover themes.
- Approach: Semi- Supervised
- Unsupervised Learning is more difficult to implement since the machine doesn't have common sense. The patterns it will find will not necessarily make sense to a human. Hence the big challenges behind an unsupervised learning algorithm.



We need your help to train the model!

What we need:

- Open text responses from different regions and across different academic domains
 - Comments should be anonymized
 - Comment titles should be related to Teaching & Learning
 - Any names found in comments will be masked as much as possible
 - New data is needed yearly to update the model
- Your help with tagging the comments
 - We aim to have a minimum of 1000 tagged comments per theme
 - You will only access your own data if you help us tag the comments
- How this data will be handled, stored and processed for the project?
 - We will not share this data outside of Explorance
 - How long we will keep the data?

Thank you!

- Questions?
- *You can contact directly with yjang@explorance.com in case you are interested to contribute*