Promoting Research Collaboration with Open Data Driven Team Recommendation in Response to Call for Proposals



Siva Likitha Valluru, Biplav Srivastava, Sai Teja Paladi, Siwen Yan¹, Sriraam Natarajan¹ University of South Carolina, ¹University of Texas







Introduction

Background:

- We introduce ULTRA (<u>U</u>niversity <u>L</u>ead <u>T</u>eam Builder from **R**FPs and **A**nalysis), a novel Al-based system for assisting team formation when researchers respond to calls/requests for proposals (RFPs) from funding agencies.
- This is an instance of the general problem of building teams where demand opportunities come periodically and potential members that best fit the criteria and requirements may vary over time.

Broad Objective:

 Build novel methods and useful tools for group recommendation with fairness, and drive different use cases (e.g., meal recommendation).

Use Cases

- Collaborative team formation.
- Balanced meal plans for diabetics: recommend meal items (appetizer, main dish, dessert) based on, e.g., dieticians' and doctors' recommendations, calorie count, added sugars/fat grams count, portion sizes, and types of dishes (e.g., vegetables, dairy, poultry, and other meat).
- Recommending an balanced exercise routine/plan.
- Sending relief teams during natural disasters.
- Building surgical teams with factors that maximize a patient's chance of survival and recovery.

Prior and Current Work

- We built ULTRA, a novel Al-based prototype for assisting with *team formation* when researchers respond to calls for proposals from funding agencies.
- Using open data, we deployed ULTRA in two universities of varying geographies: **University of** South Carolina (USA) and Indian Institute of Technology - Roorkee (IIT-R) (India).
- Current goals:
- Scale ULTRA to university-level.
- Evaluate the system for fairness. (E.g., does the system give better team recommendations for those in a certain demographic category?)

Contact Us

- Siva Likitha Valluru svalluru@email.sc.edu
- Dr. Biplav Srivastava biplav.s@sc.edu

References

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- 2. Srivastava, B., Koppel, T., Paladi, S. T., Valluru, S. L., Sharma, R., & Bond, O. (2022). ULTRA: A Data-driven Approach for Recommending Team Formation in Response to Proposal Calls. IEEE ICDM Workshop on Al for Nudging and Personalization (WAIN).
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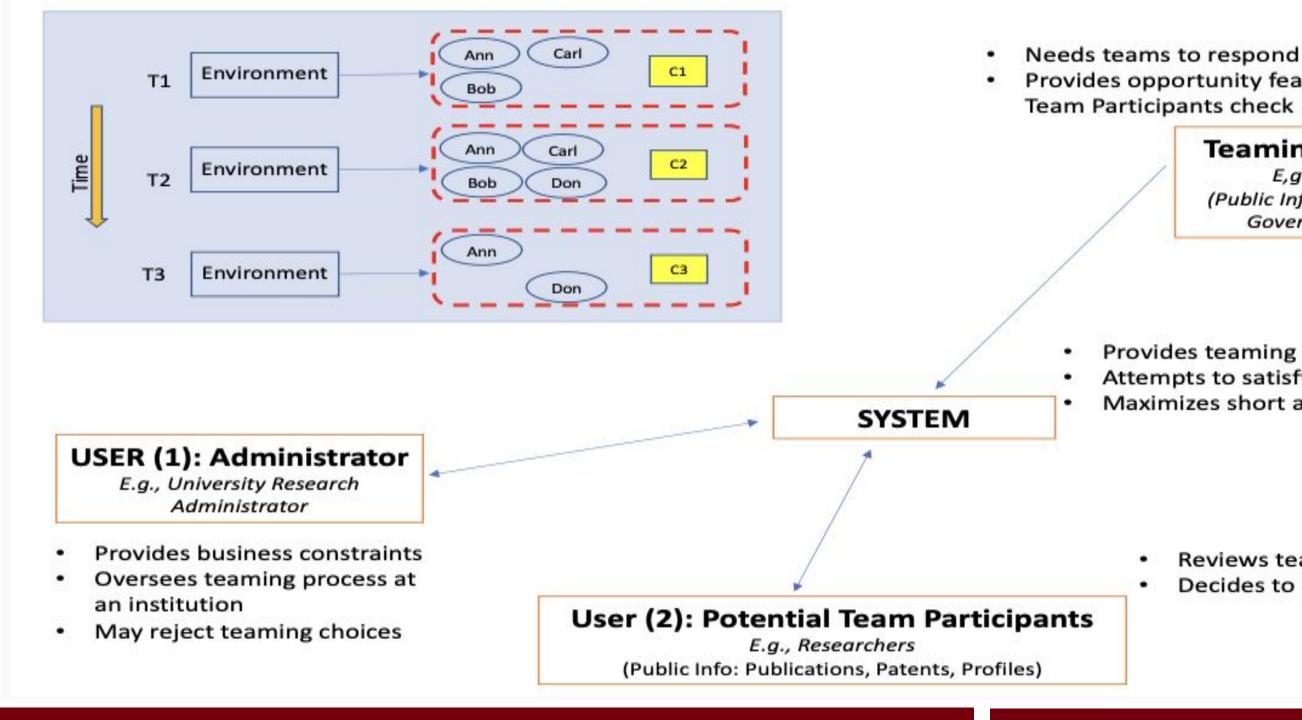
Sample Input Data from University of South Carolina

Faculty Researcher Data

Requests/Calls for Proposals (RFPs) Data

aculty Name	Research Interests	RFP ID	RFP Title	Description
Agostinelli, Forest	['Artificial Intelligence, Deep Learning, Reinforcement Learning, Search, Bioinformatics']	nsf20540	NSF-Simons Research Collaborations on the Mathematical and Scientific Foundations of Deep Learning	The National Science Foundation Directorates for Mathematical and Physical Sciences (MPS), Computer and Information Science and Engineering (CISE), Engineering (ENG), and the Simons Foundation Division of Mathematics and Physical Sciences will jointly sponsor up to two new research collaborations consisting of mathematicians, statisticians, electrical engineers, and theoretical computer scientists. Research activities will be focused on explicit
Bakos, Jason D.	['computer architecture, reconfigurable computing, heterogeneous computing, high performance computing, embedded systems']			
Banerjee, Sourav	['Wave Propagation, Ultrasonics, Acoustics, Metamaterials, Biomedical']			
Bayat, Mahmoud	['Structural Health Monitoring, Probabilistic Analysis, Nonlinear Vibrations, Machine Learning, Earthquake Eng']			topics involving some of the most challenging questions in the general area of Mathematical and Scientific Foundations of Deep Learning.

ULTRA - System Architecture



Provides opportunity features that System and Team Participants check **Teaming Opportunity**

Provides teaming recommendations

E,g., Funding Call (Public Info: Funding agencies, Government websites)

Attempts to satisfy teaming business constraints Maximizes short and long term objectives

Reviews teaming recommendation(s) Decides to participate or reject

Group Recommendation With Teaming as a Use Case: ULTRA UI Prototype for University of South Carolina

<u>Website Demo – 3 Use Cases (UCs) With 4 Teaming Methods (M#) Each</u>

http://casy.cse.sc.edu/ultra/DataExplorer/ UC2: Proposal/Method → Teams **Ultra Family of Tools Ultra Demonstration and Survey** Given a proposal and a matching method, show a list of highest ranked Select a Use Case: • Explore Ultra Data - See Data See API Select proposal Click on See Data to see the latest extracted proposals, users and awards. Select method: M3: Boosted Bandit Matching V Click on See API to redirect to Swagger pages and retrieve the content. UC1: Names/Method → Proposal/Teams Number of teams per proposal: 3 V • Explore Text Content - Unsupervised Text Visualization with KITE Given a researcher's name and a matching method, show a list of highest An Unsupervised, Effective and Inclusive Approach for Textual Content proposals and candidate teams. UC3: Research Interests/Method → Proposal/Teams Select researcher's name: Agostinelli, Forest • Text to Classification Mapper - Mapper Given a desired research interest and a matching method, show a list of highest ranked proposals and candidate teams. Select method: M3: Boosted Bandit Matching V Tool that display the code and description from the ACM and JEL tree based on the text we enter Number of Results: 5 V, Number of teams per proposal: 3 V Select method: M3: Boosted Bandit Matching V • See Recommendation for Teaming - Teaming Under http://casy.cse.sc.edu/ultra/teaming/ Enter **Development**

Use Case 3 - Sample Results

Selected Proposal: quantum, Selected Method: M3: Boosted Bandit Matching Proposal ID Index Skills **Recommended Teams** Proposal Name Score ['Chao, Yuh J. ', 'Farouk, Tanvir I.', 'Huang, Xinyu ' 'Rizos, Dimitris', 'Zhang, ['Farouk, Tanvir I.', 'Gatzke, Edward P.'. 'Huang, Xinyu ', 'Rizos, Dimitris', 'Zhang, Bin'] ['Chao, Yuh J. ', 'Farouk, Tanvir I.', 'Huang, Xinyu Rizos, Dimitris', 'Zhang, Fully-Connected quantum nsf17548 Quantum Computer ['Farouk, Tanvir I.', Challenge (PFCQC Gatzke, Edward P.', 'Huang, Xinyu ', 'Rizos, Dimitris', 'Zhang, Bin'] ['Banerjee, Sourav' Farouk, Tanvir I.', 'Huang Xinvu ', 'Rizos, Dimitris', 'Williams, Christopher'l ['Banerjee, Sourav', 'Chao, Yuh J. ', 'Huang Xinyu ', 'Rizos, Dimitris' 'Williams, Christopher'l

Goodness Metric

 We consider four methods for group **recommendation** and the following metrics to evaluate teams. We penalize the first two metrics, and reward the former.

Metric	Definition
Redundancy	Number of skills that a team of researchers commonly have.
Set size	Team size.
Coverage	Number of skills satisfied by a team in response to an RFP.
<i>k-</i> Robustness	Effectiveness of a team if <i>k</i> members are unavailable.

ULTRA - Human Study Evaluation and Computational/Qualitative Feedback

Human Study Responses for USC (28 Days, 212 Responses)

M0: 52 (24.53%) UC1: 79 (37.26%) Very Relevant: 157 (74.06%) Very Useful: 172 (81.13%) M1: 43 (20.28%) UC2: 68 (32.08%) M2: 49 (23.11%) Somewhat Relevant: 34 (16.04%) Somewhat Useful: 35 (16.51%) M3: 68 (32.08%) UC3: 65 (30.66%) Neutral: 8 (3.77%) Somewhat Irrelevant: 8 (3.77%) Neutral: 4 (1.89%) Irrelevant: 5 (2.36%) Not Very Useful: 1 (0.47%) —

Qualitative Feedback

- An IRB-approved user study was conducted in May 2023, where asked participants were interact with ULTRA and provide feedback:
- "This is incredible and has a lot of potential. Can't wait for this to be in real time!"
- "Very well thought out! Great resource to the university"

Average Goodness Scores and Quality of Teaming Results for 2 Universities

Method	Average Quality	Average Volume
МО	0.0879 ± 0.0290	10
MI	0.3673 ± 0.1569	10
M2	0.4097 ± 0.1313	9
M3	0.5295 ± 0.0816	6

Method	Average Quality	Average Volume
MO	0.0896 ± 0.0006	10
M1	0.4218 ± 0.0011	8
M2	0.4292 ± 0.0017	7
M3	0.5835 ± 0.0203	1

We created and implemented Al methods using string, taxonomy, and more advanced contextual boosted bandits in ULTRA. (Please refer to our paper for more details.)

Across two different settings in the US (top) and India (bottom), as recommendation methods become more data-informed, they improve teaming quality while reducing team size.