



# From online education to gig science and augmented intelligence

Oct, 2019, AISIS

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# Co-research process for applied data analysis challenges.

Failures, values, instrumentation and open calls.

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### Trends and Goals

- > Online teaching, blended learning
- Interdisciplinarity (Al challenges for Particle Physics: nature.com/articles/s41586-018-0361-2)
- Al hype attracts bright students from other science domains

Goal: How can we teach students practical data science AND advance domain science at the same time at no cost?

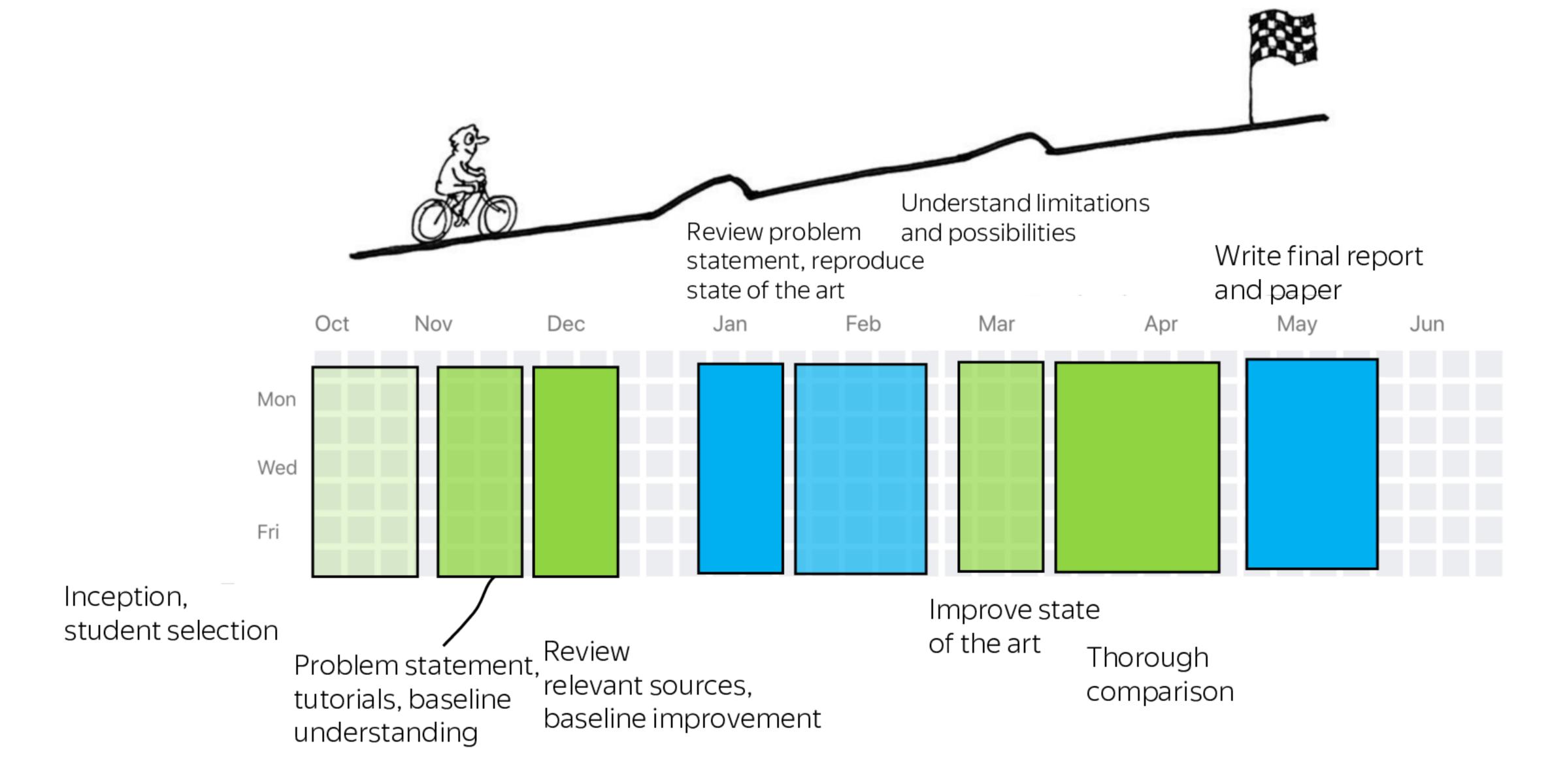
improve junior student involvement improve efficiency of research (predictability of results) reduce cost per student



# Lab Project Highlights 2018/19

- 1. OPERA shower reconstruction and generation (ACAT'19)
- 2. Generation of LHCb calorimetry images (ACAT'19)
- 3. FERMI telescope image denoising (JINST paper draft)
- 4. JUNO event reconstruction (JINST paper draft)
- 5. Transfer learning for fast Monte-Carlo event generation (ACAT'19)
- 6.Cityscape autonomous vehicle camera images reconstruction

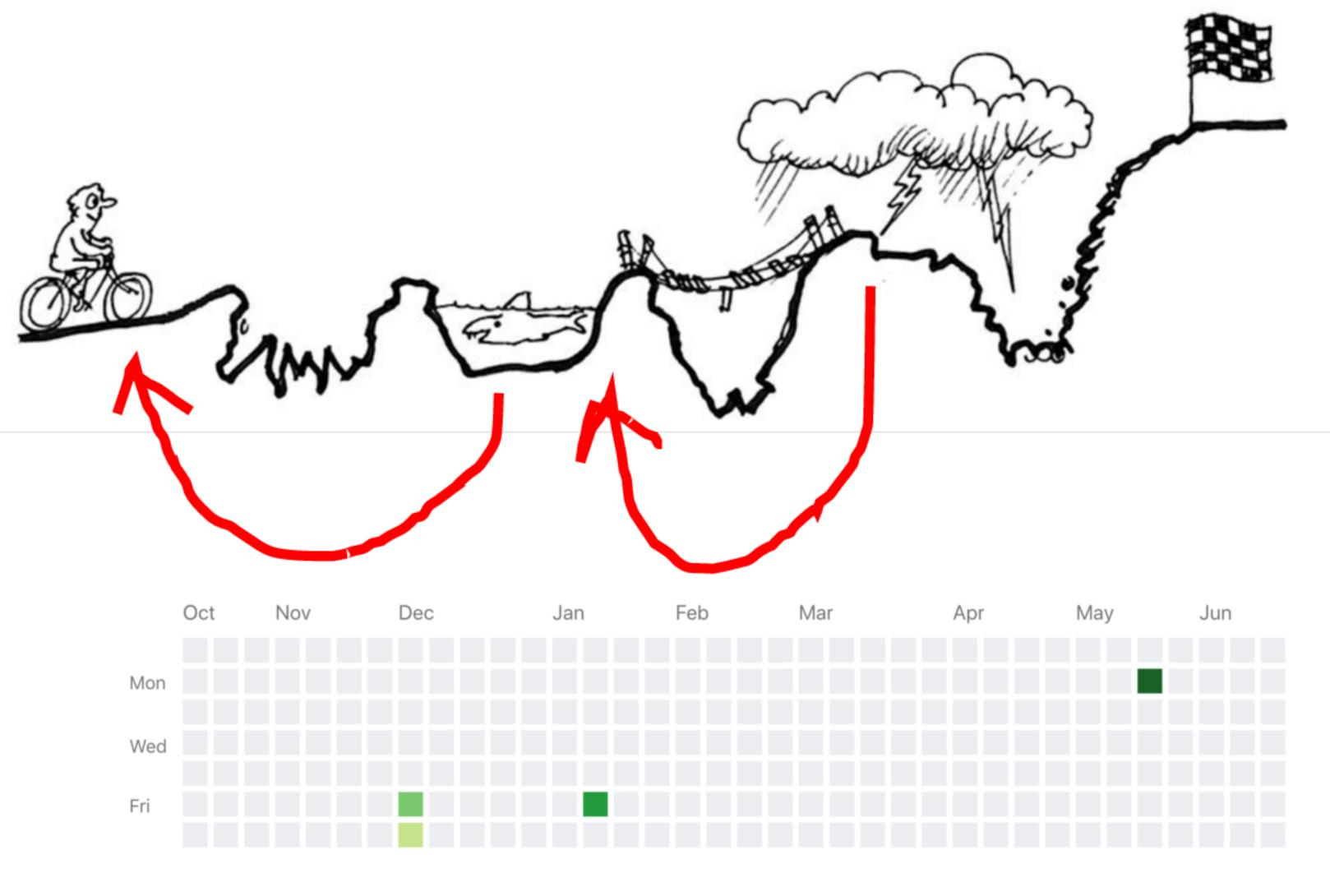
### Research timeline - ideal



### Research timeline - Real

### Main issues

- Unknown unknowns (data, design, others' work, computations)
- Underestimated risks
  Lack of proper
  communication
- Lack of supervisor time/attention
- Bad luck



# Can it be improved?

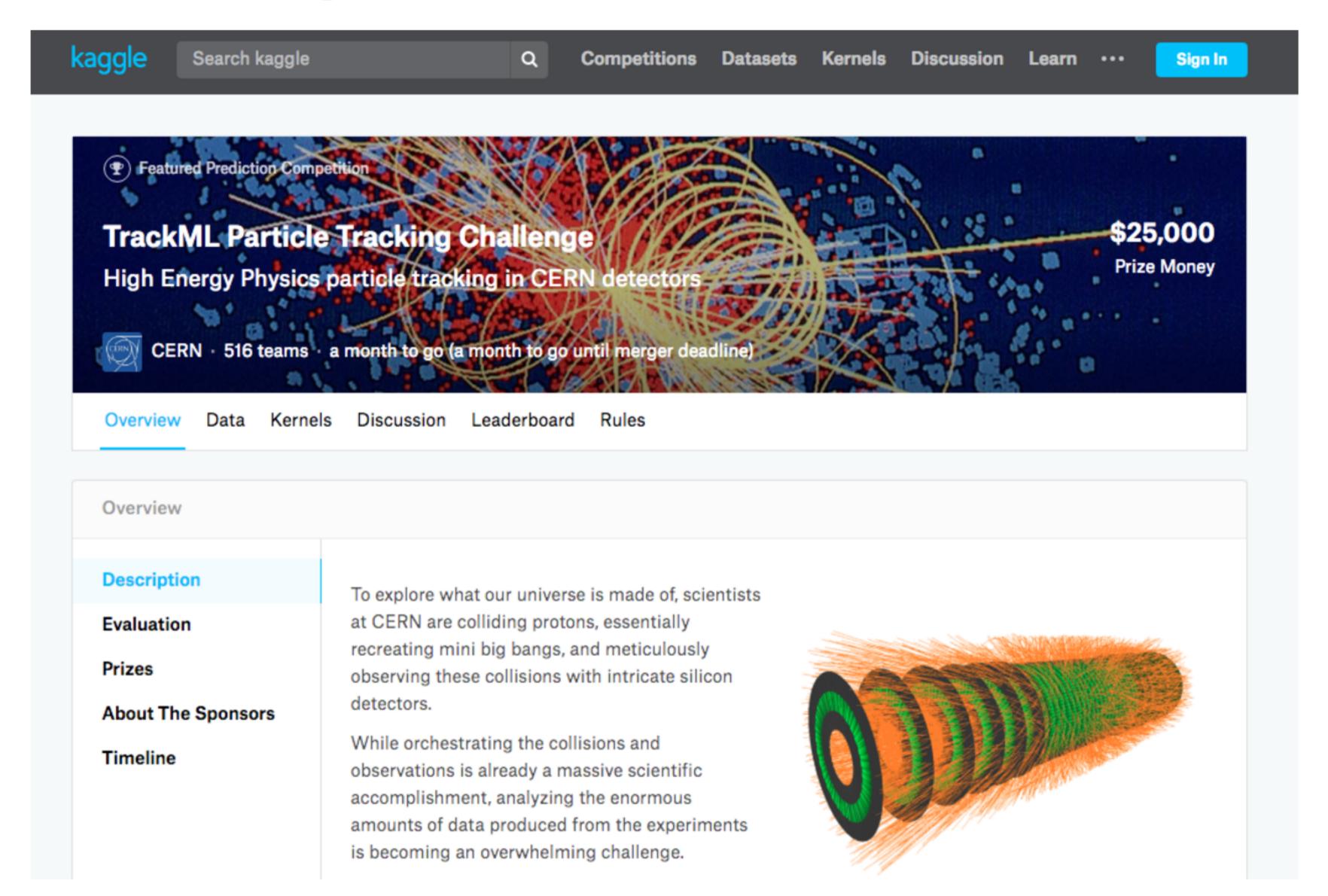
Traditional approach: 1-on-1 supervision

proven by ages, but doesn't scale well

Challenge-based approach

scales better, but a bit away from science difficult to explain, students consider it as homework they skip until deadline

# Machine Learning + Science examples



# "Kaggle" style shortcomings

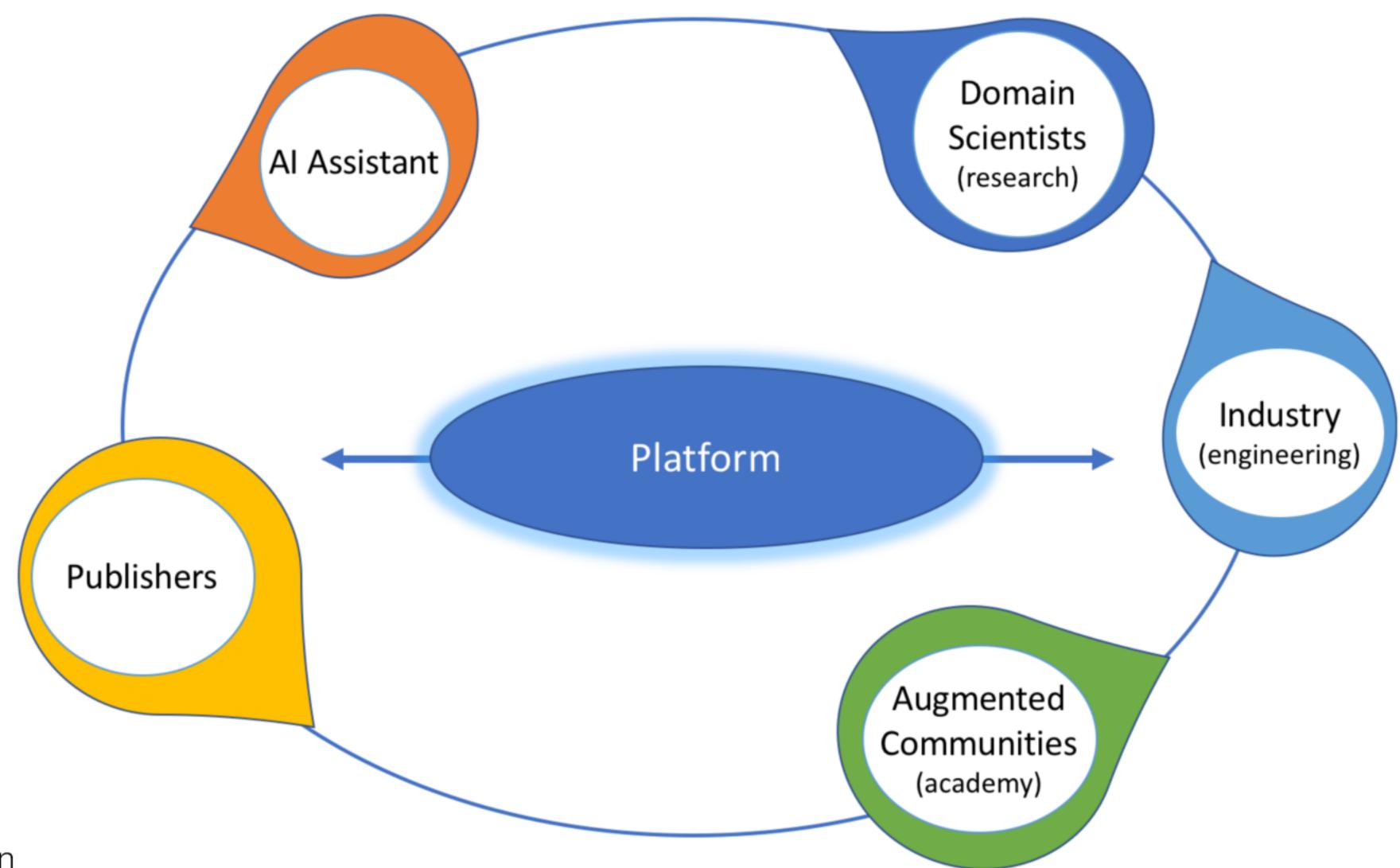
- Domain specificity are difficult to check automatically:
  - > Non-trivial metrics
  - Constraints
  - Participation Incentive
  - ML-practitioners sometimes consider it as real science
  - Beyond iid assumption (the test set differs from the training set)

Maybe we need to go deeper? Explain and support values, explain and run the process, have infrastructure in place to support it.





### Co-research holistic view



### Co-research values

Transdisciplinary approach #transdiscip #data Data quality #reproducible Reproducible results Scientific ethics #ethics Innovative efficient research #innovation #coolresearch Amazing results Social good #socialgood Strong Al-scientific community #community (International) collaboration #teamwork Science is not done until communicated #PR

# Project big lifecycle (co-research)

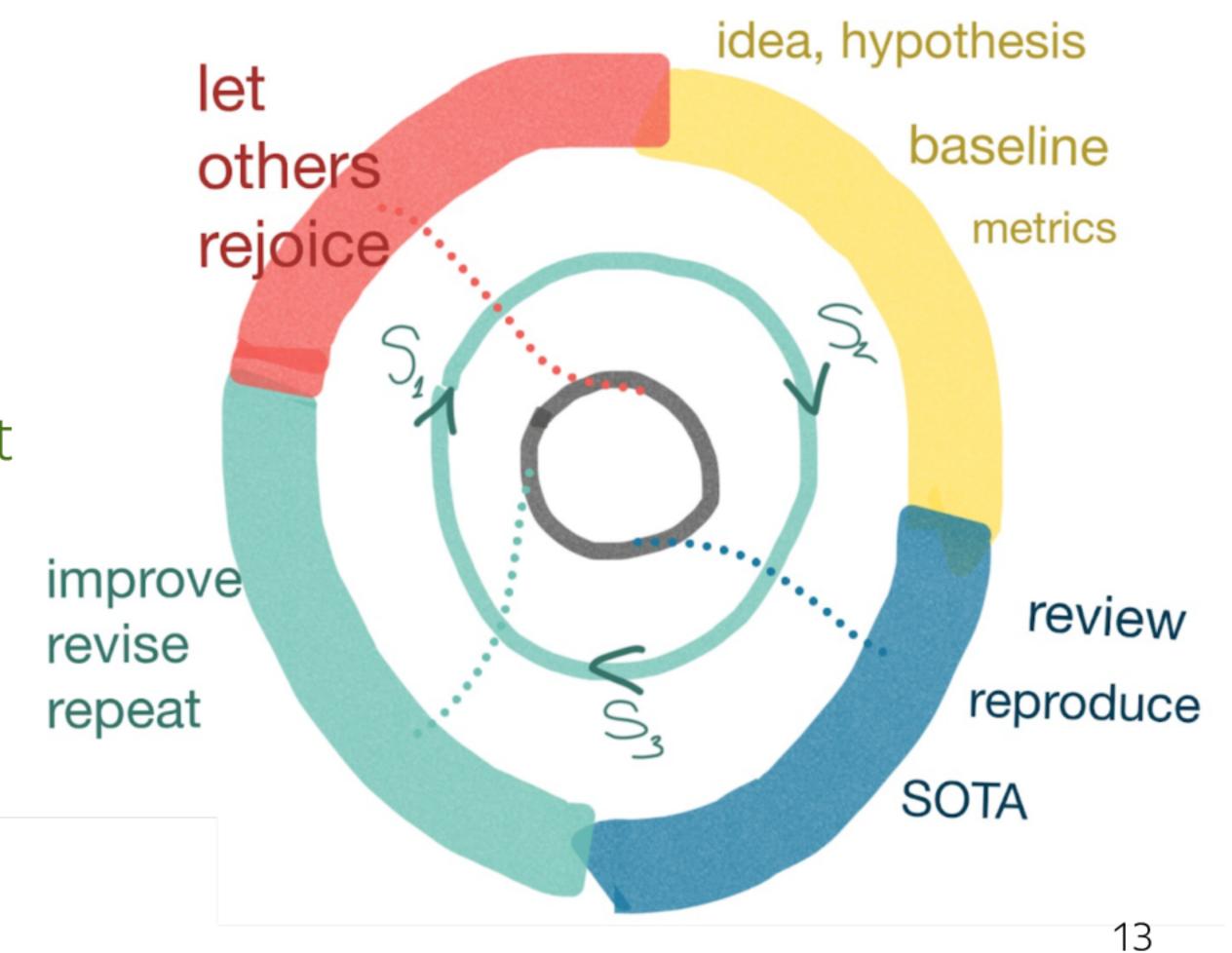
### Inception stage

Learning: doing papers review, study of «state-of-the-art» methods

individual solution: iterative search, model understanding and improvement

Writing final report, slides and paper

NB: iterations may be nested on every stage.



# Lifecycles

- Weekly lifecycle
  - > Iterate through current tasks,
  - > report results at weekly meeting / trello
  - > Choose tasks for the next week
  - > Artifacts: git commits, trello cards, models
- Bi-monthly lifecycle (stages)
  - > Team report to wider audience (lab meeting)
  - Artifacts: slides, trained models, re-evaluated datasets and metric scripts

### Process Roles

### Domain experts

- > Problem statement
- Dataset, explanation of dataset, initial metric, state of the art explanation

### ML experts

- > Supervise mentors, students
- Refine metric in ML terms, methods landscape

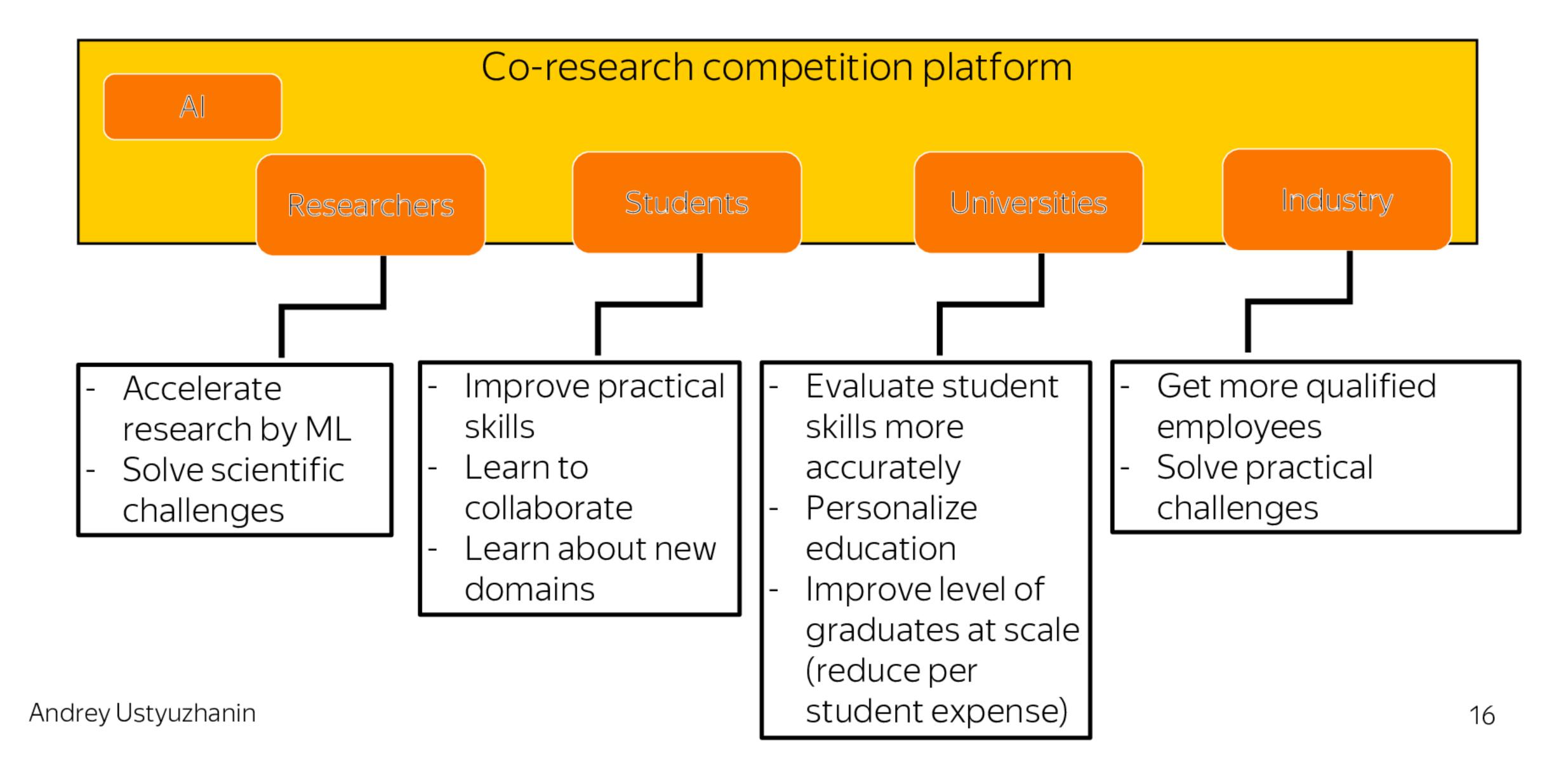
### Mentors

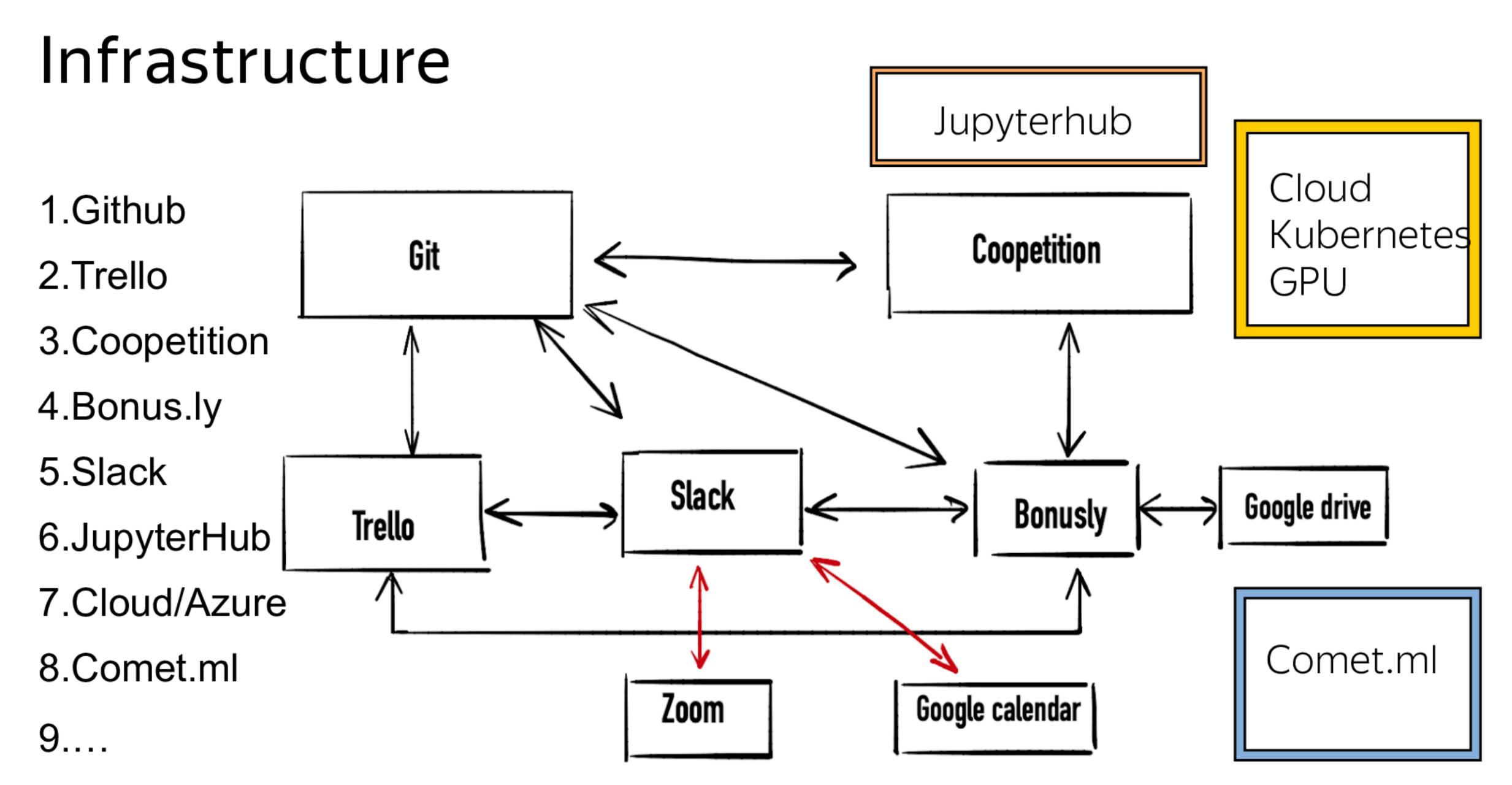
- Implement baseline, setup infrastructure
- > Facilitate weekly/monthly lifecycles with students

### Students

- > Understand the problem
- Understand and follow the process (implement, write, present)
- > Understand the final solution
- > Help and learn from others

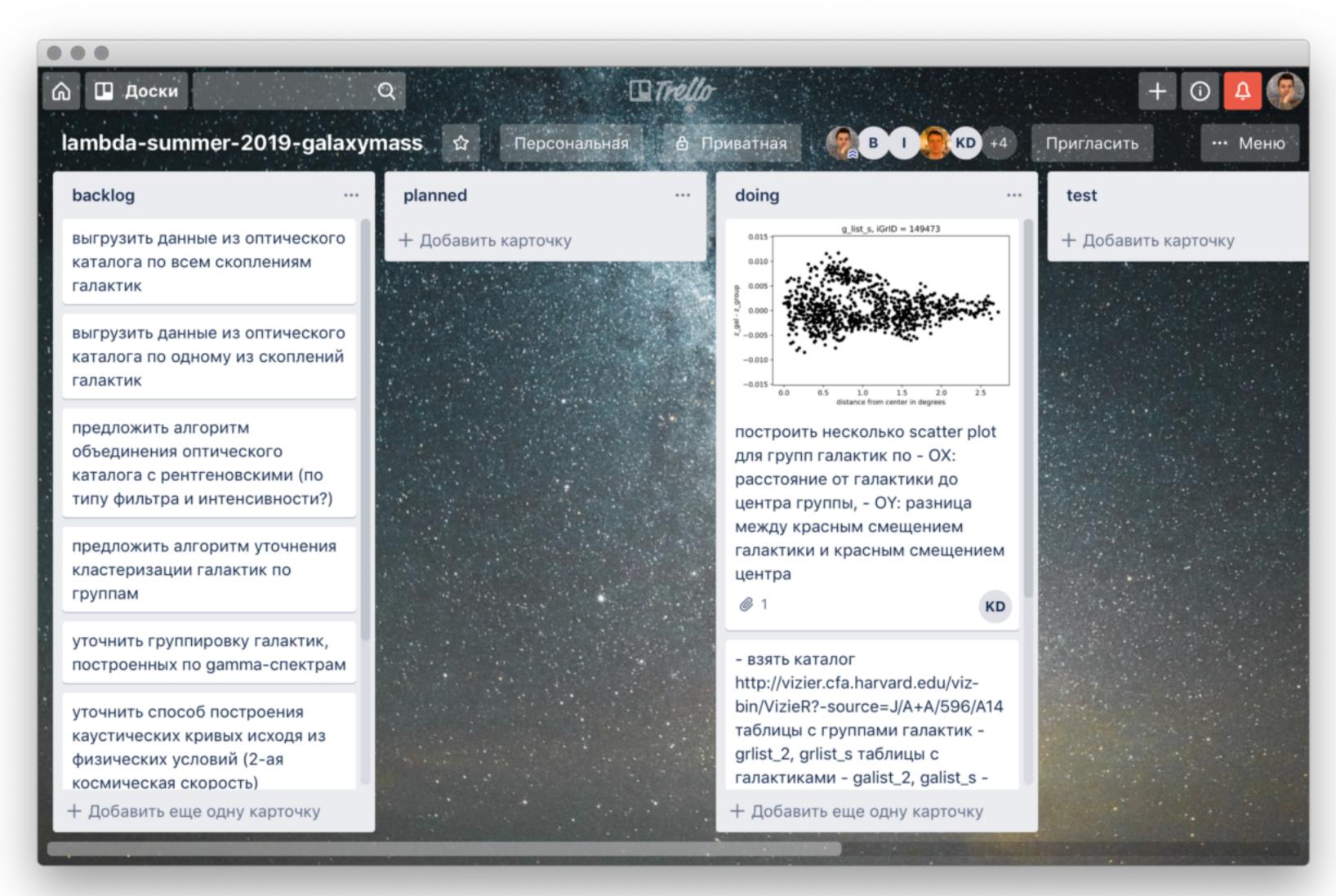
# Co-Research Platform - Coopetition





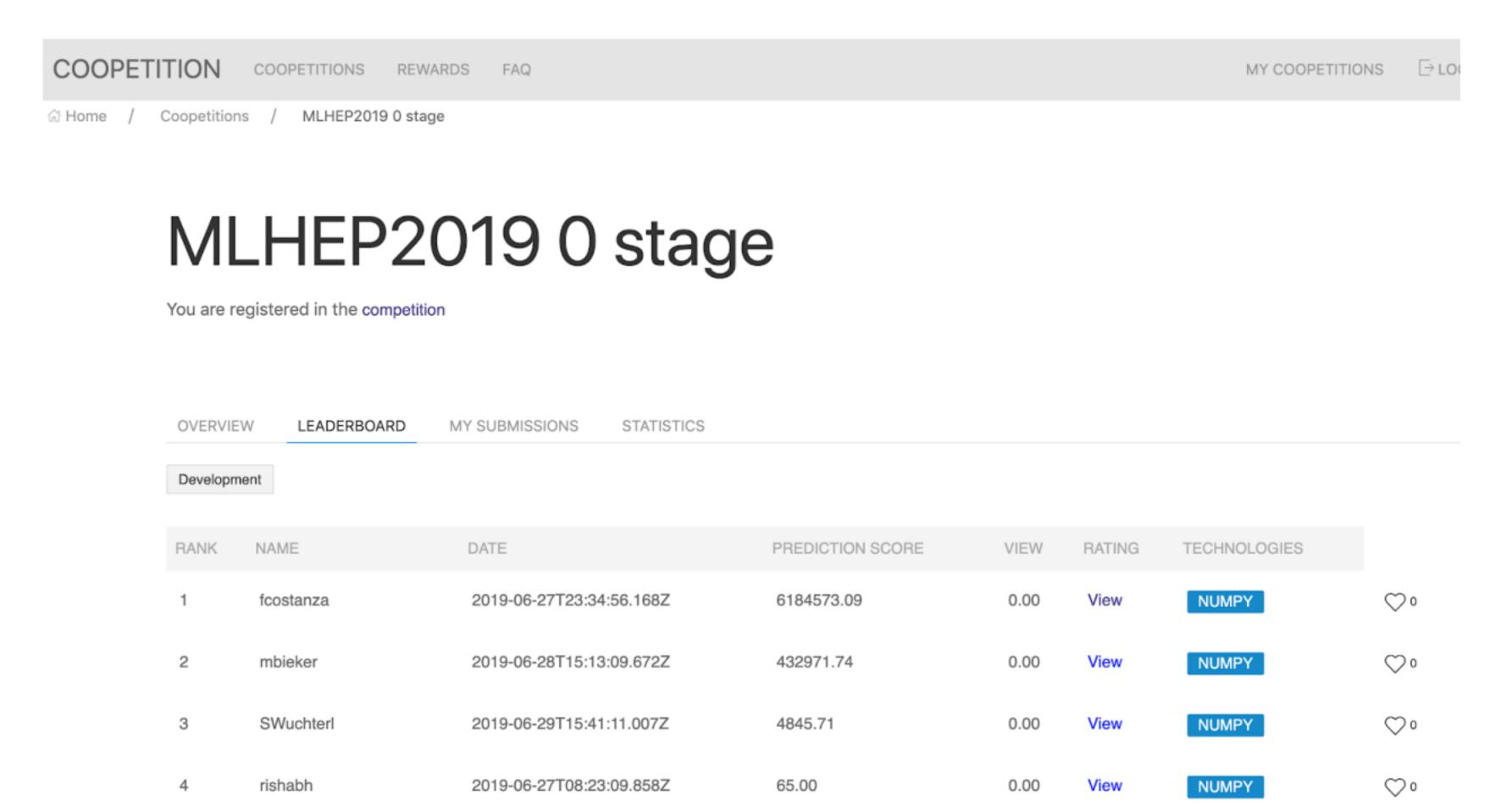
# Trello – project roadmap

- Project roadmap (links to info, who's doing what, what's going on now):
- A board with the current tasks or «cards»
- Card may contain checklists, comments, deadlines, etc.
- Cards are grouped into columns

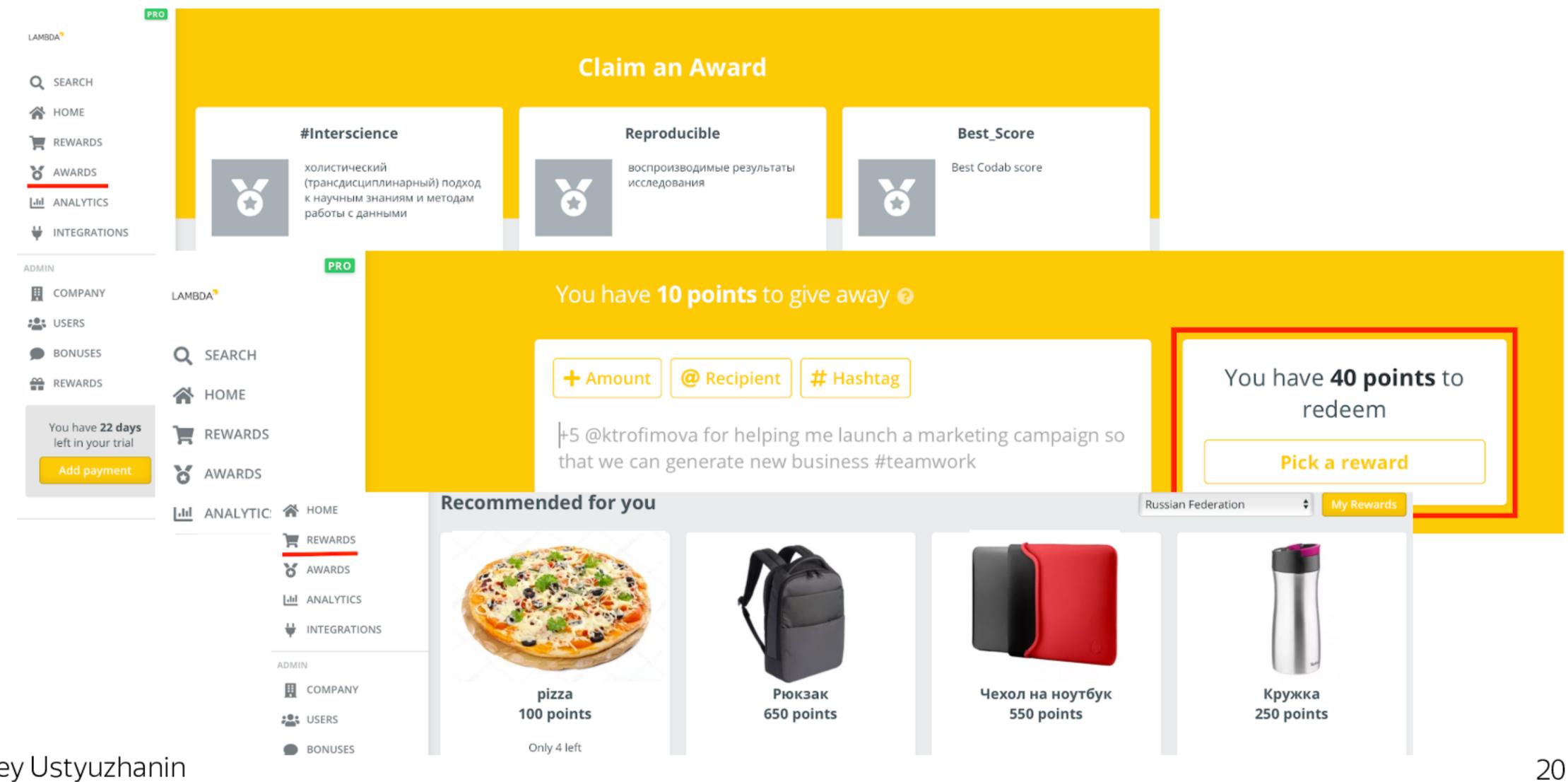


# Coopetition – submission checker

Tracks user submissions from github
Verifies the submission code
Give recommendations \*)

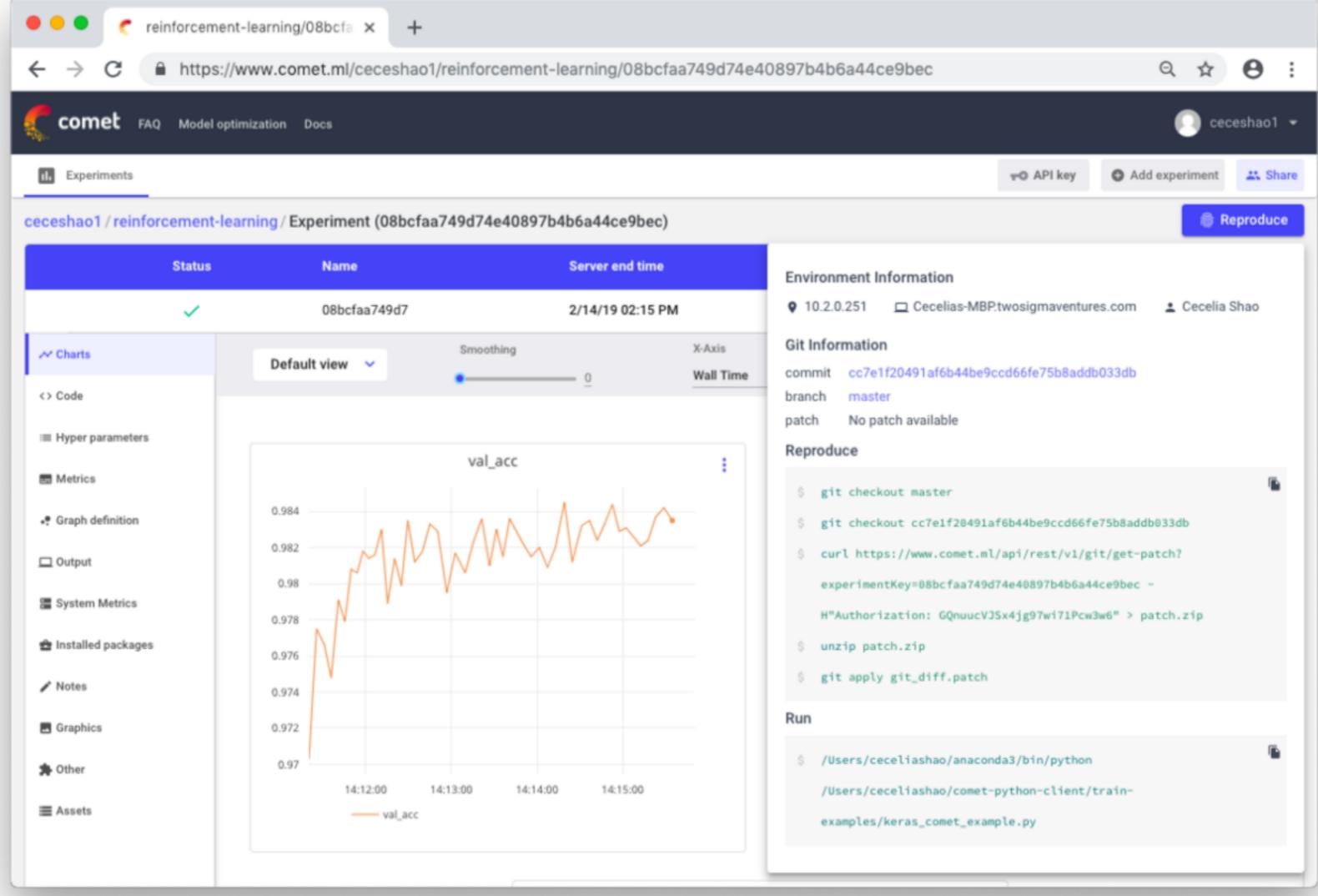


# Bonus.ly – employee engagement



# Comet.ml – model quality monitor

Tracks of model
training progress
Keeps track of
dependence between
params, data and results
Presents results in
nice plots



# Process - infrastructure matching

### Preparation

- All students have access to small Jupyter server with data and limited GPU resources
- > Student forks baseline git repository

### Iteration

- > Student clones his/her repository and plays with it on Jupyter server or locally on data subsample until he/she wants to test it on big dataset
- > Student commits/pushes to git and it automatically runs training on Cloud
- > Comet.ml evaluates solution metrics while training (updates constantly)
- Coopetition evaluates model automatically, updates leaderboard and bonuses
- Coopetition suggests recommendations (materials, code of others)

# Open challenges towards gig science

Problem selection adaptive to student level
Low-code workflow language
Addressing core ML (more abstract) problems
Addressing out-of-iid settings (e.g. data has been generated under the same causal assumptions but with different initial settings)
Student code analysis and improving recommendations
Al Research assistant (meta-learning agent)

# Our Lab Project Highlights 2019/20

- 1. X-ray tomography
- 2. DUNE event reconstruction speed-up by deep neural architectures
- 3. X-ray Laser simulation and tuning
- 4. Estimation of population size by genomes analysis
- Study of community polarization by search/recommendation output system analysis
- 6. Leadership style identification by social network profile
- 7. Al Researcher
- 8. Natural Language to Machine Learning corpus assembly
- 9. Improving Earth satellites position prediction by simulation tuning
- 10. Galaxy cluster mass estimation and registry update
- 11. Warm and cold dark matter hypothesis comparison

http://bit.ly/2VNGoll

### Conclusion

- Huge demand for collaboration between ML & Domain scientists
  - That also can help students to improve practical skills
- Co-research process defines 2-level iterations to help participants to stay tuned:
  - relies on explicit collaborative values and infrastructure
  - implies reproducibility and motivation that leads to better results and PR
  - more scalable than peer-to-peer supervision (less cost per student)
  - flexible to adjust for your team needs
  - way towards cross-domain multi-level Al community and gig science
- We are open for collaboration with your group and university 🥮

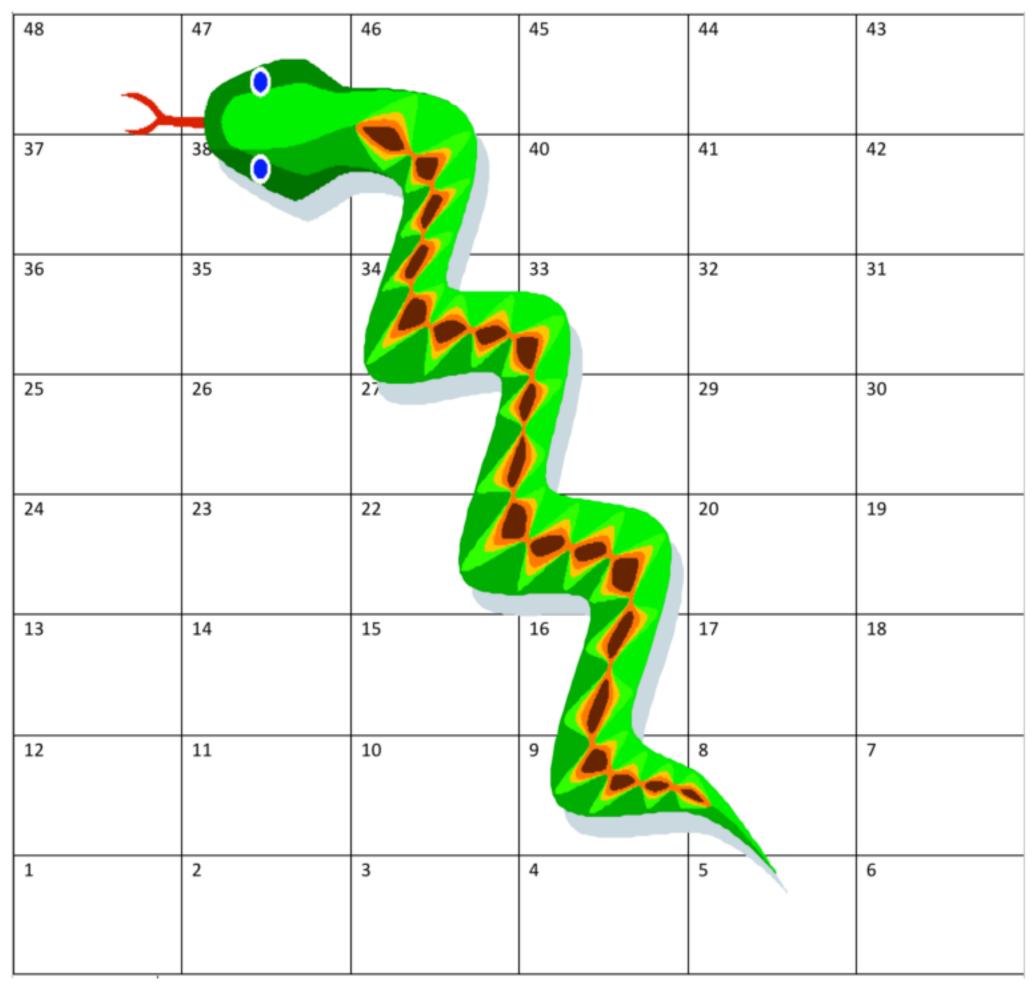


# Backup

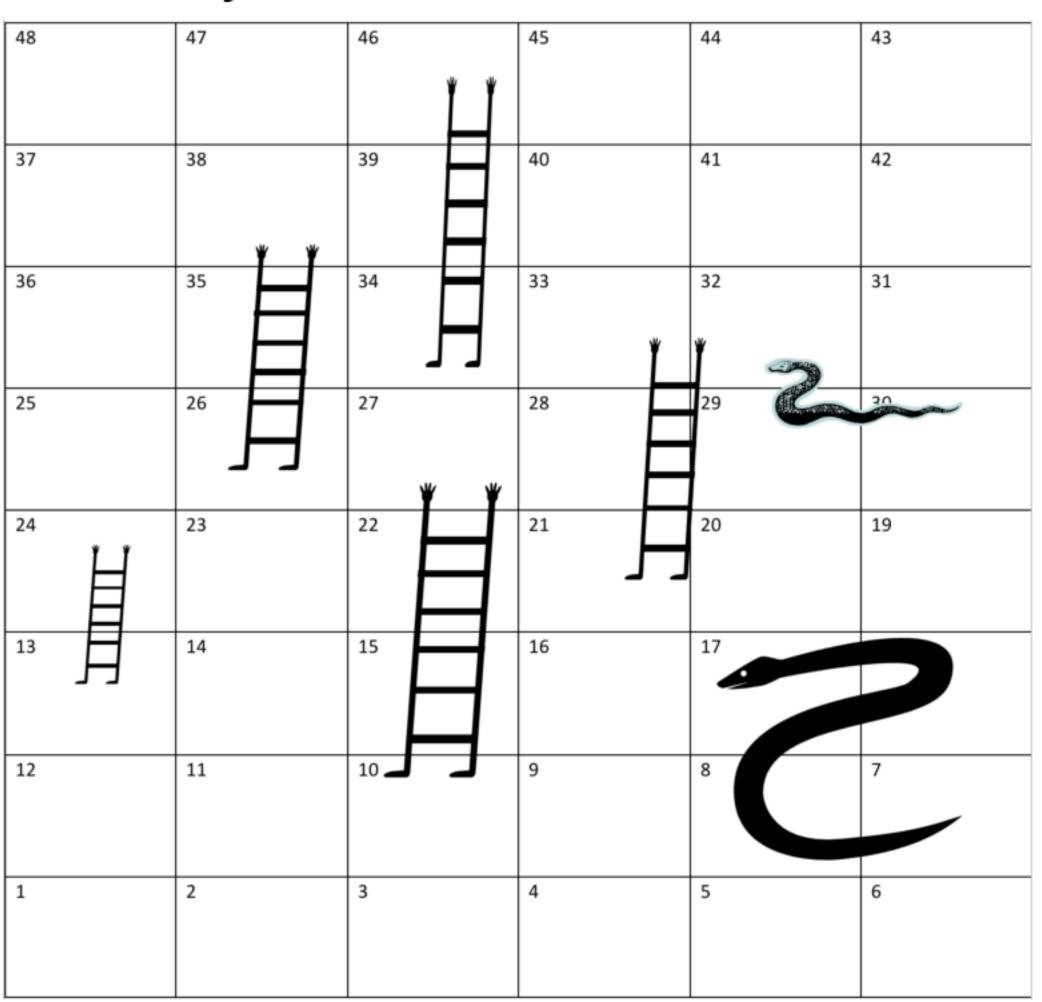


# Research process metaphor

### Optimistic view:



In reality:



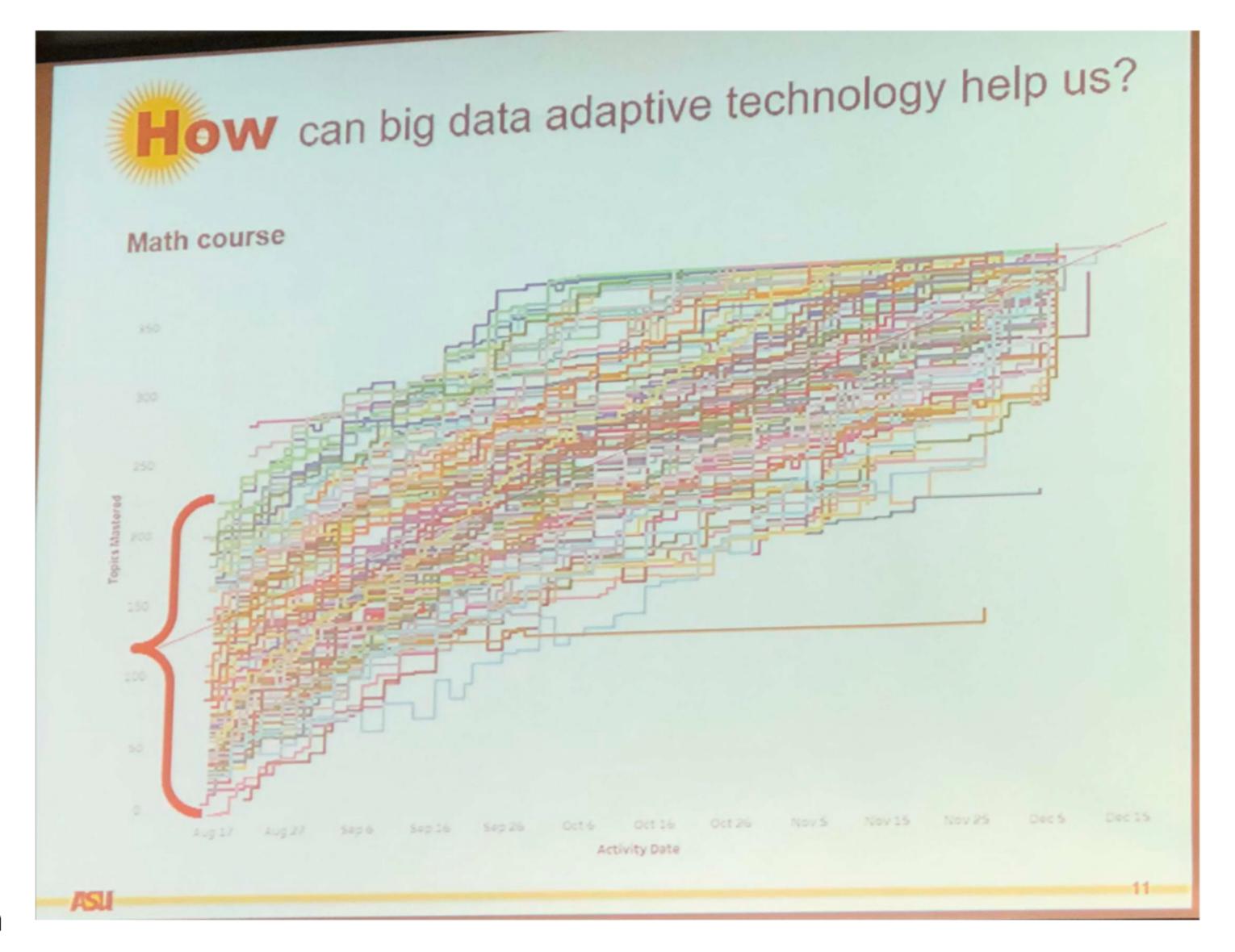
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# Machine Learning + Science examples



# Arizona State University teaching math



# Quick self-intro





Head of LHCb Yandex School of Data Analysis (YSDA) team
Head of Laboratory (link) of methods for Big Data Analysis at
Higher School of Economics (HSE),

> Applications of Machine Learning to natural science challenges

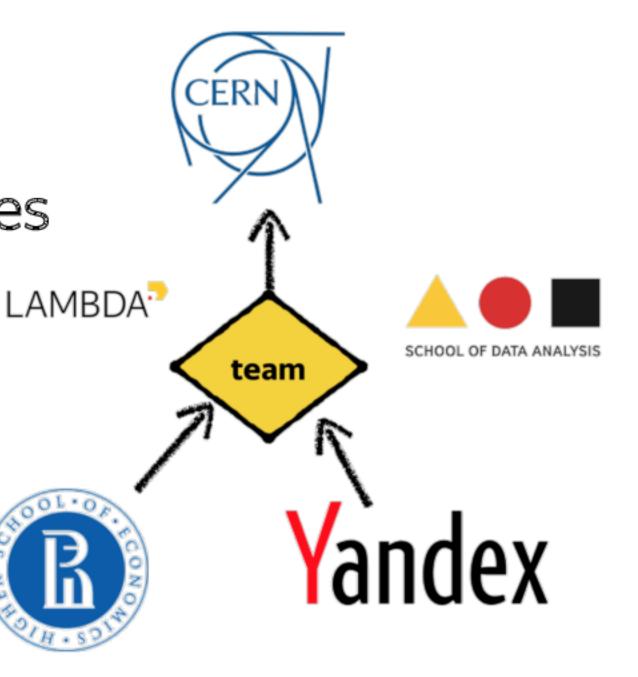
> HSE has joined LHCb in 2018!

Co-organizer of Flavours of Physics Kaggle competitions (2015)

Co-organizer of TrackML challenge (2018)

Education activities (MLHEP, ML at ICL, Clermont Ferrand,

URL Barcelona, Coursera)



### Outline

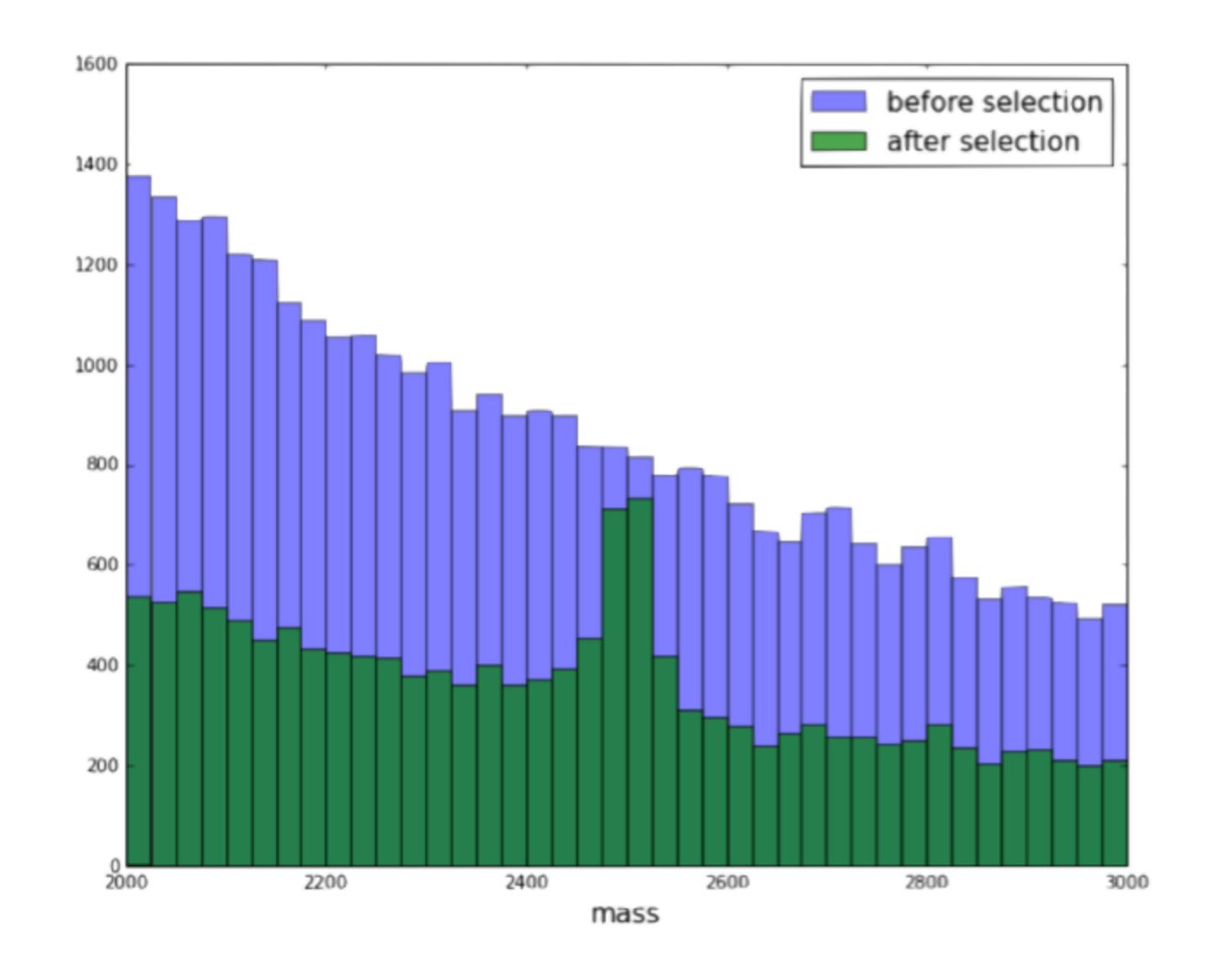
Applied ML research trends

Research process

Competition + Cooperation

- Strengths
- Issues: reproducibility, metrics, beyond iid, motivation
- And beyond
  - Yalues, Lifecycle, Process
  - > Roles, Infrastructure
- Open issues and call for cooperation
  - > Project highlights 2019/20

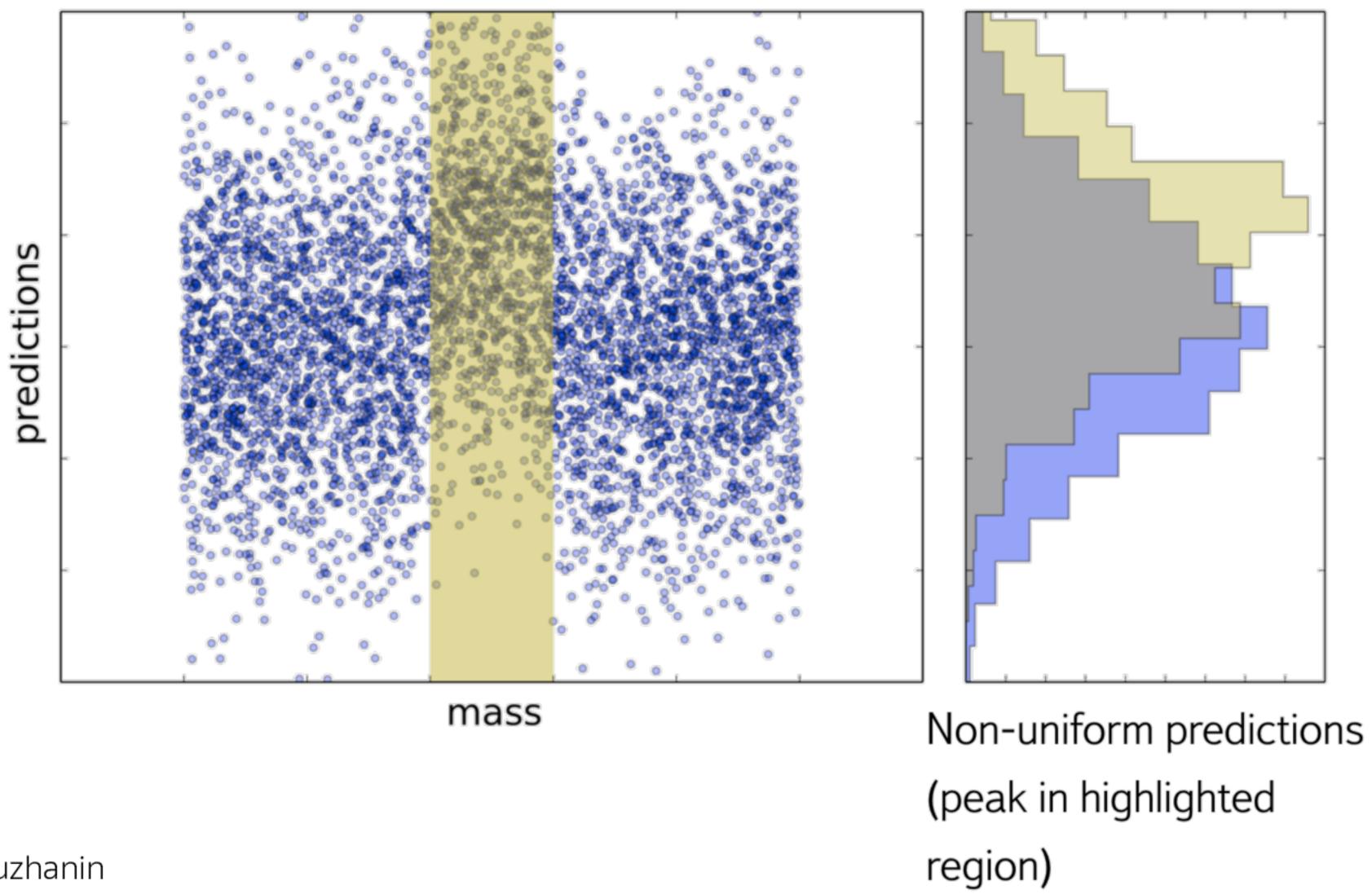
# Domain specific example



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# Non-uniformity check



# Kaggle incentive: winner takes it all



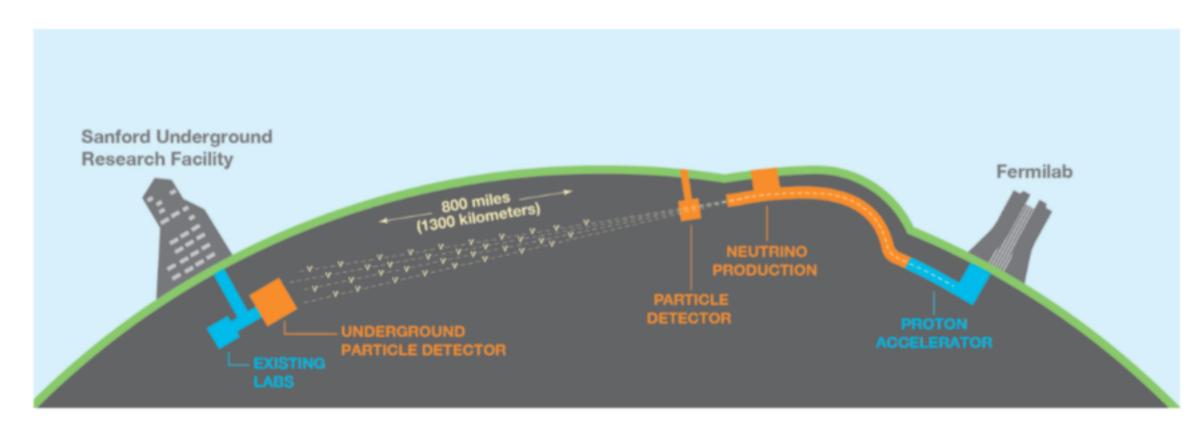
# Can it be more fair?

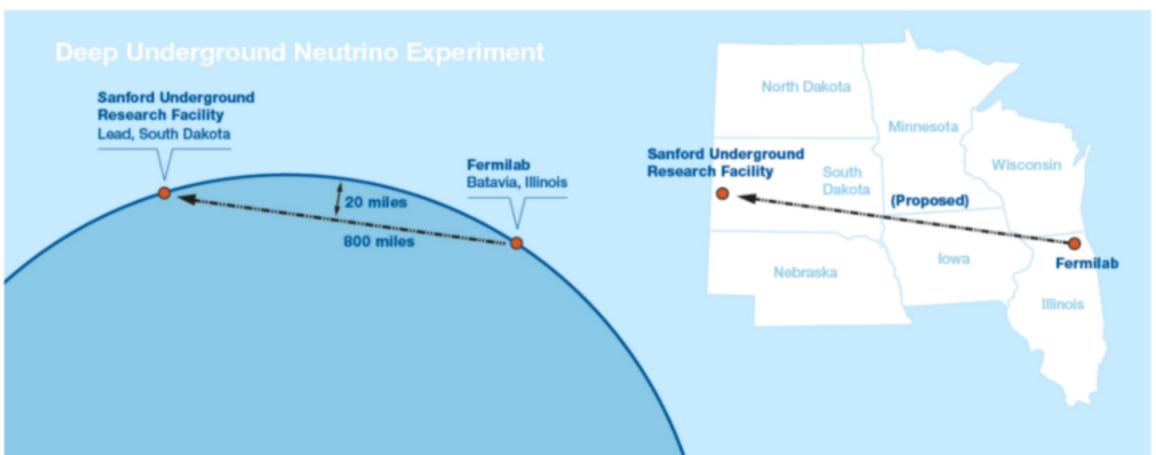


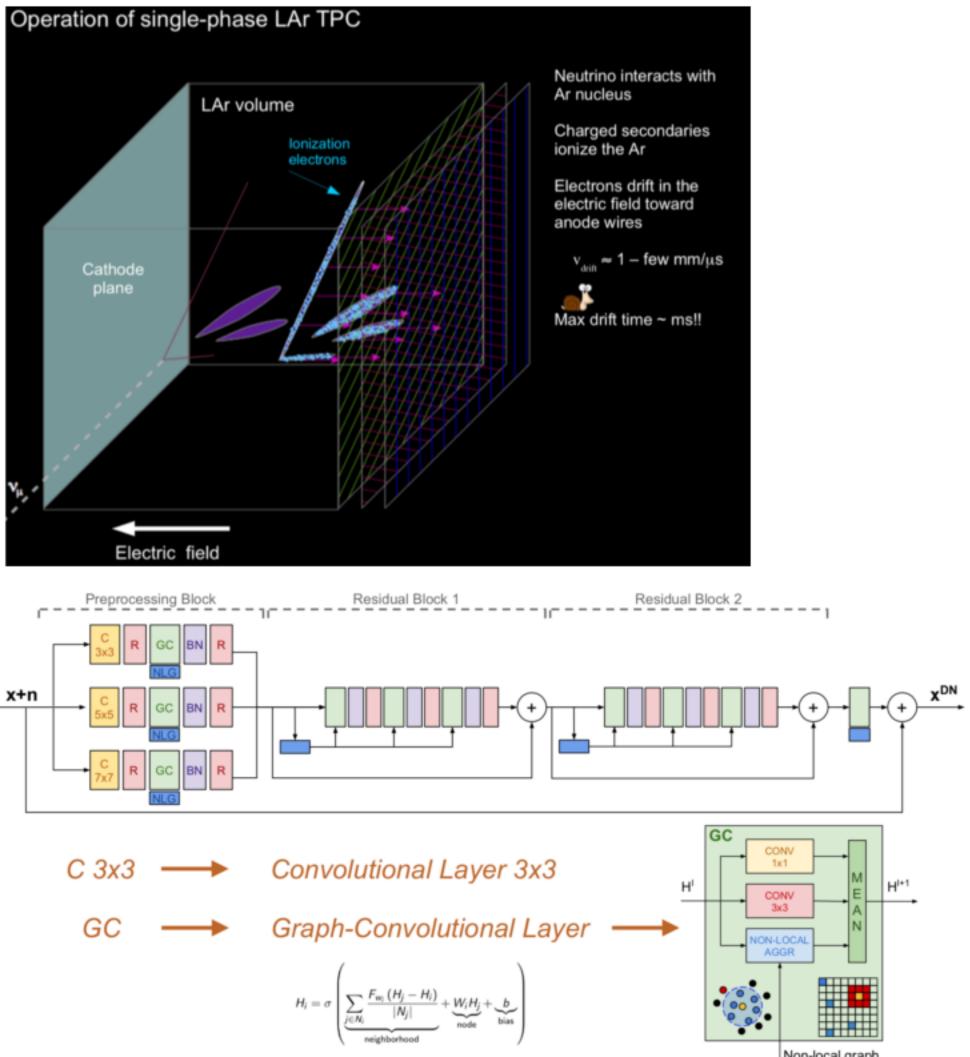
# Micro-rewards examples

- Metric-based:
  - > top X of the leaderboard,
  - time spent at the top,
  - > persistence
- Source code-based:
  - > Distinctive solution
  - > Source of other's inspirations
  - Your code is re-used by others

## Dune event reconstruction

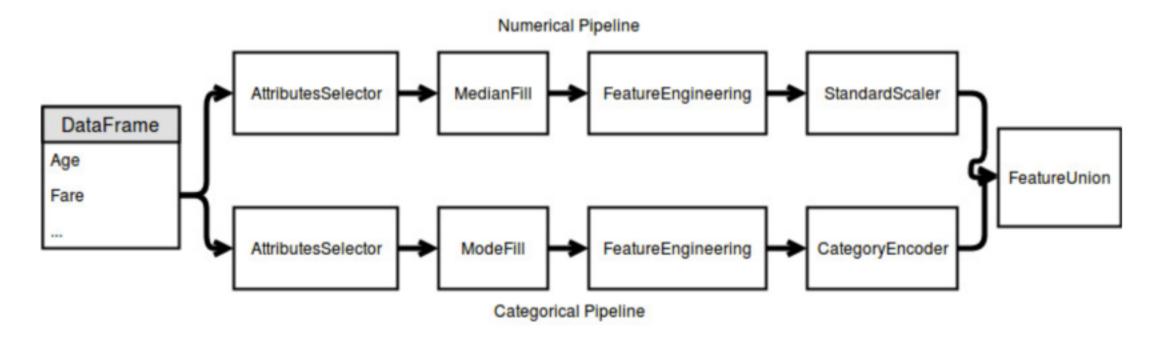






# Natural language to machine learning corpus

- Collect code snippets of ML examples (e.g. spacy.io, uber.github.io/ludwig/, scikitlearn.org, stackexchange.com)
- 2. Annotate
- 3. Build knowledge graph of possible cases and map pairs into it
- 4. Build proof-of-concept classifier for certain kinds of cases



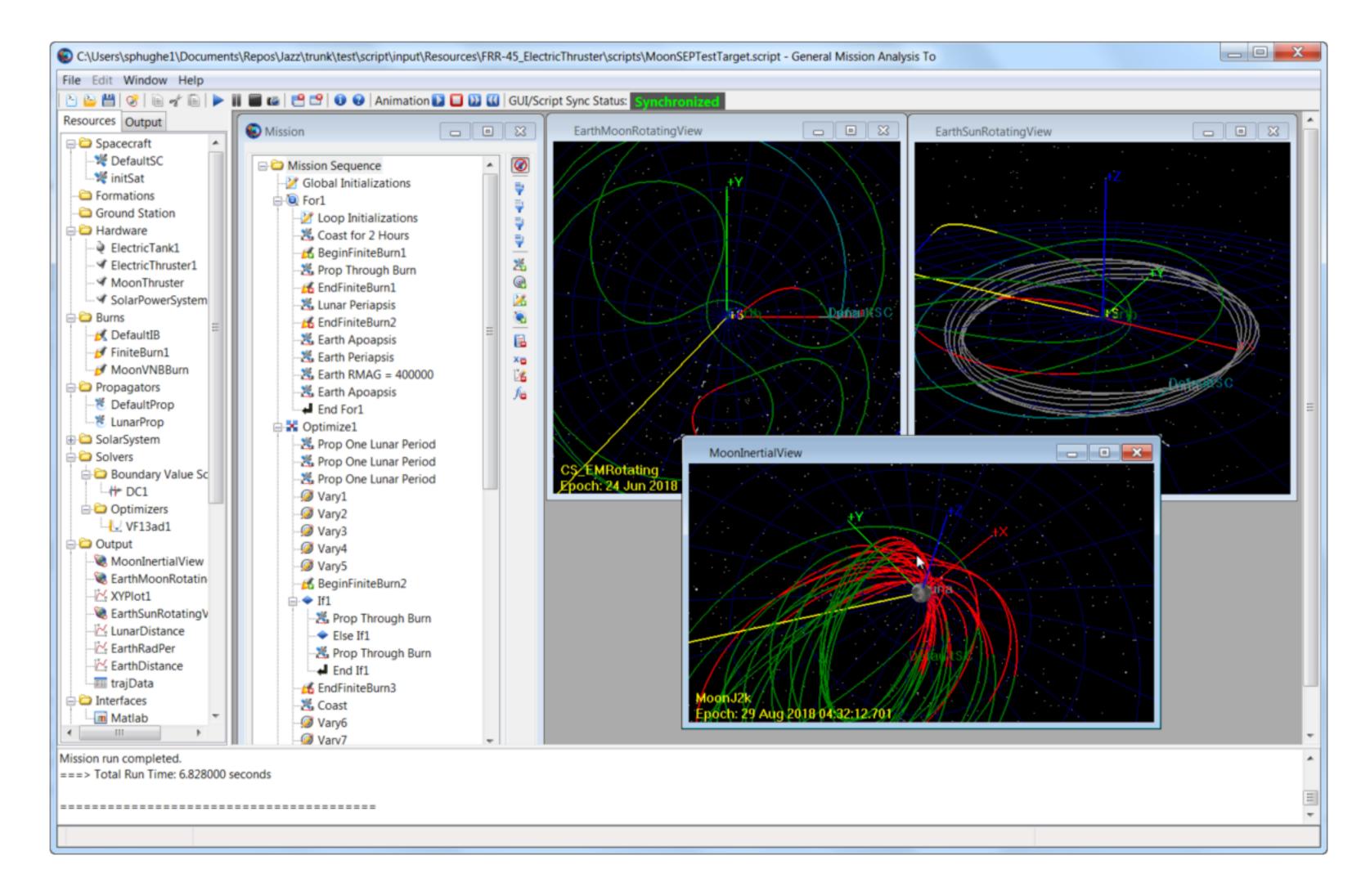
```
{
    "question_id": 36875258,
    "intent": "copying one file's contents to another in python",
    "rewritten_intent": "copy the content of file 'file.txt' to file 'file2.txt'",
    "snippet": "shutil.copy('file.txt', 'file2.txt')",
}

{
    "intent": "How do I check if all elements in a list are the same?",
    "rewritten_intent": "check if all elements in list `mylist` are the same",
    "snippet": "len(set(mylist)) == 1",
    "question_id": 22240602
}

{
    "intent": "Iterate through words of a file in Python",
    "rewritten_intent": "get a list of words `words` of a file 'myfile'",
    "snippet": "words = open('myfile').read().split()",
    "question_id": 7745260
}
```

# Satellite position prediction

- Explore observation uncertainty propagation in time
- 2. Find satellite feature representation
- 3. Tune simulation model parameters to improve precision of satellite position predictions



# Al Physicist++

### Unsupervised explorer algorithm

