

Adaptive Academic Advising for 100 Astronomy Students

Author: Basmala Samir

1. 📚 Curriculum Graph

The astronomy program was modeled as a directed graph, where:

- **Nodes** represent courses (e.g., A101: Intro to Astronomy)
- Edges represent prerequisite relationships (e.g., A101 → A201)

The full curriculum includes 11 core astronomy courses:

• Introductory level: A101 – A103

• Intermediate level: A201 – A203

• Advanced/specialized: A301 – A402

We used:

- NetworkX to represent and analyze course dependencies
- Matplotlib to visualize the graph

This structure enables our system to understand which courses unlock access to others, and how students can progress effectively.

2. Mark Student Simulation

We simulated a realistic cohort of 100 astronomy students, each with:

- 3–8 completed courses chosen randomly
- **Grades** from A to F converted to GPA (0.0–4.0 scale)
- 1–2 interests, such as:
 - o Stellar Astrophysics
 - Cosmology
 - Planetary Science
 - Astrobiology
 - Observational Techniques

We used:

- random to vary interests and outcomes
- numpy to compute GPA
- pandas to store and manage the simulation

Each student ends up with a profile that mimics a real-world academic history — ideal for testing recommendation strategies.

3. Personalized Course Recommendation Strategy

Rather than train a complex reinforcement learning agent, we implemented a **rule-based** (heuristic) system that behaves like a real academic advisor.

Our logic works as follows:

Eligibility first: Only recommend courses for which prerequisites are fully completed

Interests next: Prioritize courses that match a student's declared interests

Max Load: Limit to 3–5 courses per term

GPA balance: Avoid overly difficult courses for low-GPA students unless required

4. I Sample Recommendation Output

```
{
    "student_id": 45,
    "interests": ["Planetary Science", "Cosmology"],
    "completed": ["A101", "A102", "A201", "A202", "A203"],
    "GPA": 3.1,
    "recommended_courses": ["A303", "A401", "A402"]
}
```

This student completed all the prerequisites for A303, unlocking access to advanced astronomy courses that match their interests.

5. Key Design Choices

- Surriculum based on astronomy, not Al to show transferability
- * Heuristic-based Al: simpler, explainable, and effective
- Smart logic mimics real academic advising

This shows the power of even simple rule-based AI to drive real value in personalized education.

6. A Final Notes

This project successfully demonstrates a full pipeline:

- See Curriculum graph modeling
- Realistic student simulation
- Smart personalized recommendations