

Infinite *Campus*

More than a Student Information System

Taking the Mystery Out of Early Warning

Campus Analytics Suite

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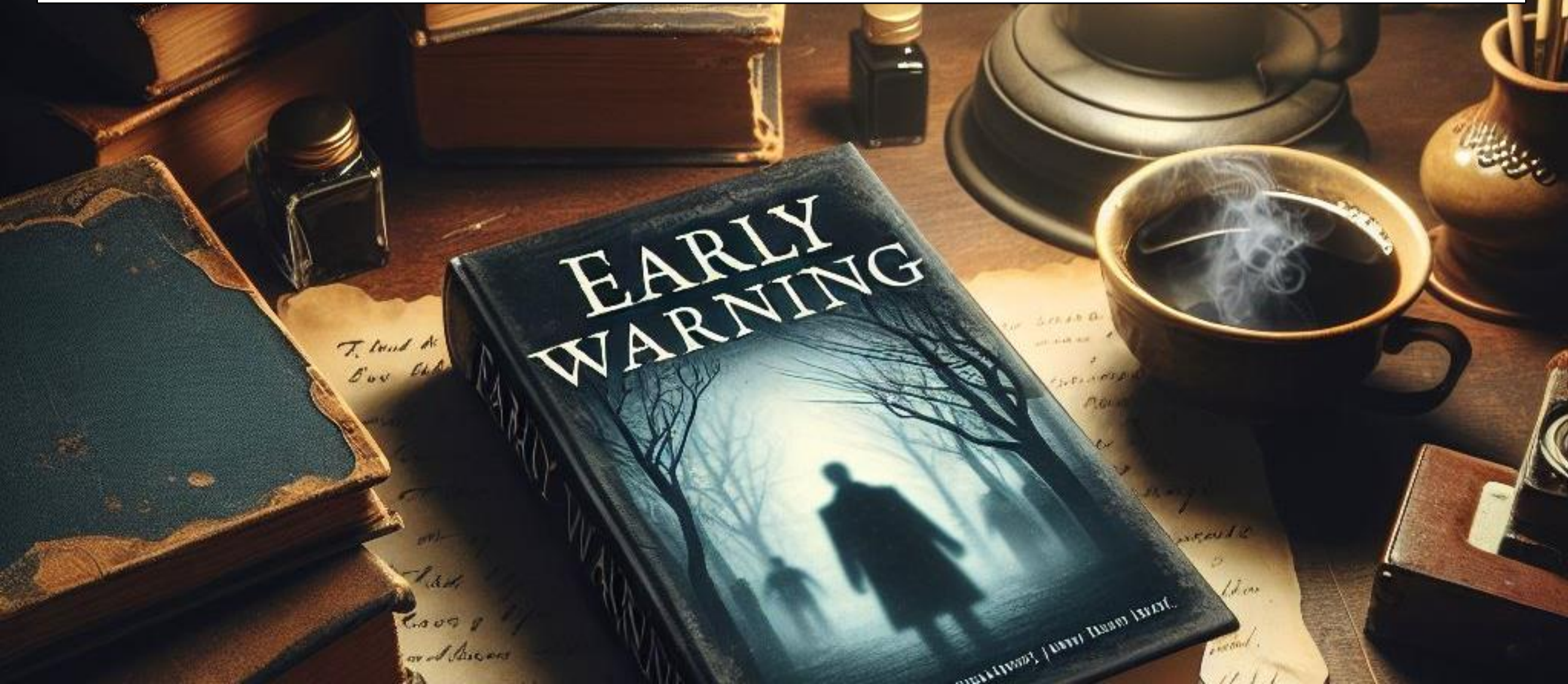


Taking the Mystery Out of Early Warning



- What is Early Warning?
- What is a GRAD Score?
- Is Early Warning AI?
- What is Machine Learning?
- How is a GRAD Score calculated for a student?
- What can I learn from the GRAD score?

What is Early Warning?



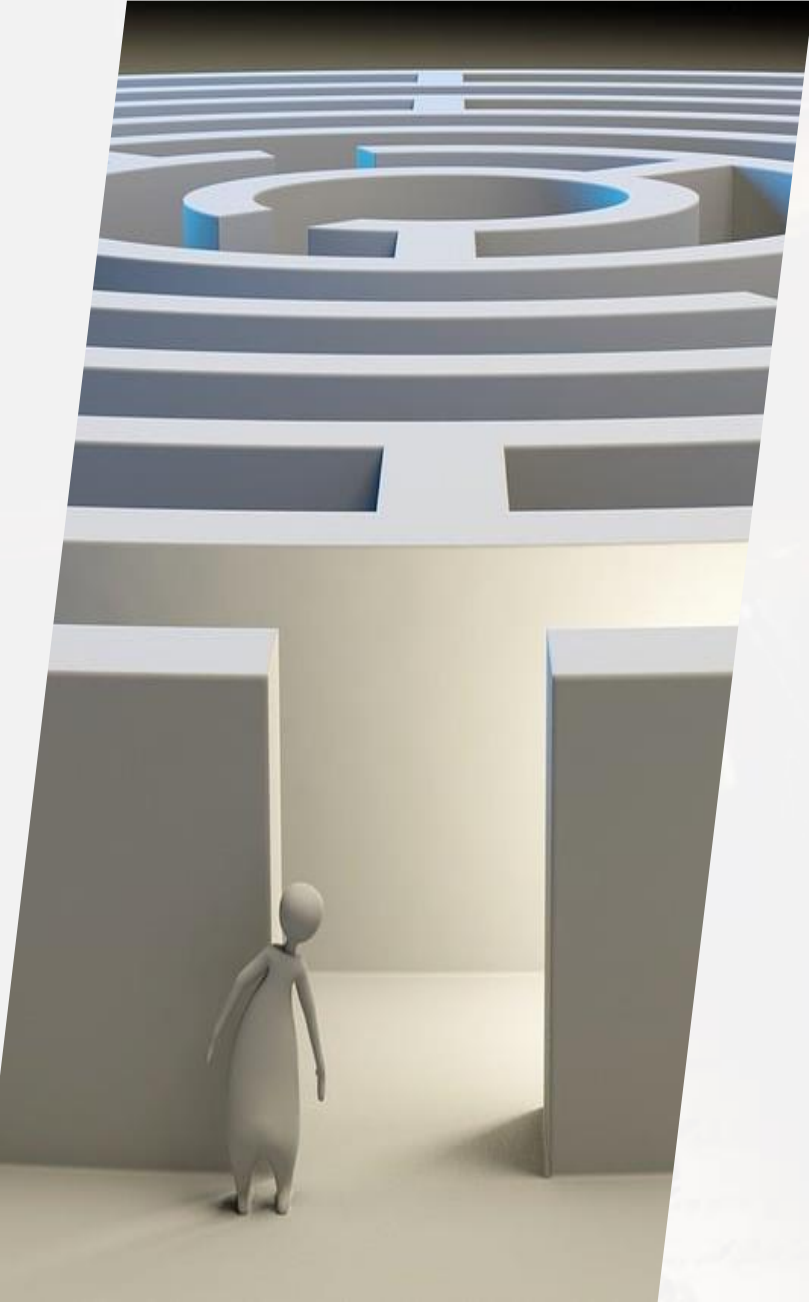
Early Warning

The Infinite Campus Early Warning tool is a data dashboard that helps inform stakeholders of each student's likelihood of completing high school and/or progressing to the next grade level.



Why did we build it?

The Early Warning system was built to address a problem with conventional Graduation Monitoring systems



Conventional Systems

Conventional systems utilize a set of thresholds to identify at risk students

- Typical examples of this would be:
 - Students with more than 5 concurrent absences
 - Students with a GPA below 2.0
 - Students with a specific FRL Status



Conventional Systems

The problem is that Threshold Systems lack the context to understand the data, which can result in over and under identifying students

“While it is true that many students who qualify for FRL face additional challenges that may impact their ability to graduate, it is not true that ALL students who qualify for FRL will struggle to graduate”

MIND THE GAP

Taking the Mystery out of Early Warning



Robert Balfanz, PhD

Distinguished Professor at the Johns Hopkins University School of Education, Center for Social Organization of Schools, and Director of the Everyone Graduates Center.

His work focuses on translating research findings into effective school improvement strategies and educational reforms. He conducts research on improving high school graduation and college readiness rates, student success systems, chronic absenteeism, and instructional improvements in schools.

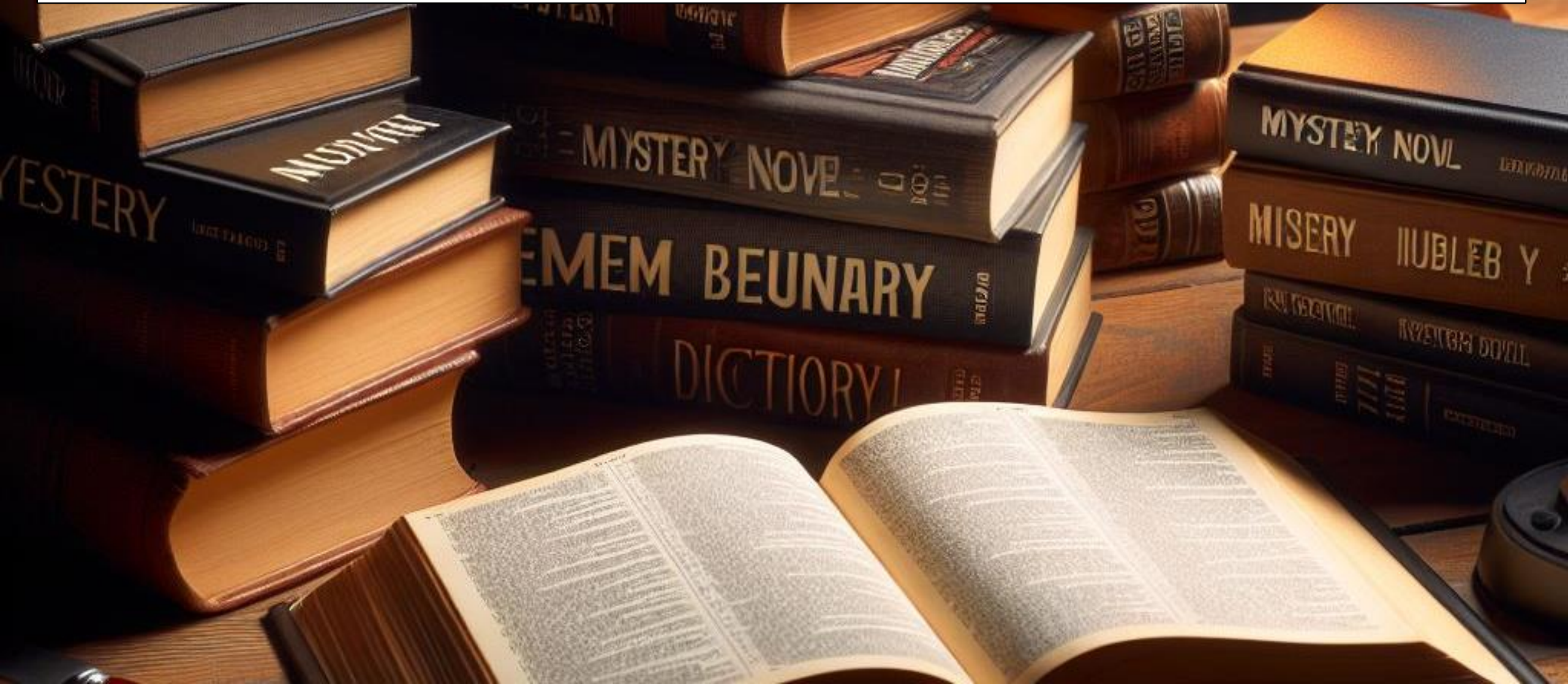
MIND THE GAP

An Innovative Approach

Early Warning was built upon the initial indicators identified in the work of Dr Robert Balfanz and further enhanced through Machine Learning to create a pattern-matching algorithm capable of predicting student risk

The result of this prediction is called a
"GRAD Score"

What is a “GRAD” Score?



GRAD Score Defined

Graduation Related Analytic Data

Summarizes a student's educational record with a single number indicating a student's likelihood of graduating or being promoted to next grade level



GRAD Score Defined

Scores range from 50 – 150

The lower the score, the higher the risk

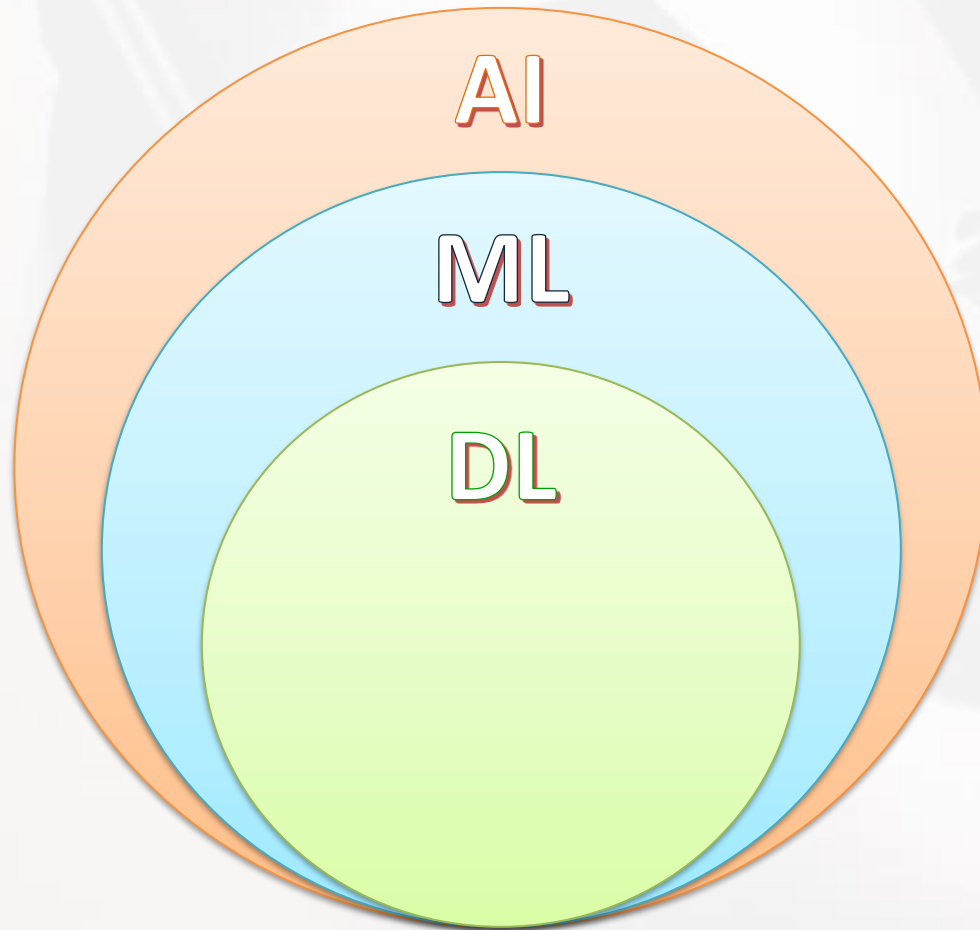


50.....60.....70.....80.....90....100...110...120...130...140...150

A man in a dark suit and hat is sitting on a large, dark, rectangular sign that has the letters 'AI?' written on it in a bold, white font. The scene is dimly lit, with a strong light source from the left creating a bright glow around the sign and casting a shadow of the man onto the surface. The background is a wooden floor.

Is Early Warning AI?

Is Early Warning AI? No



Artificial Intelligence

Any software that uses complex methods and algorithms to imitate human intelligence and decision making

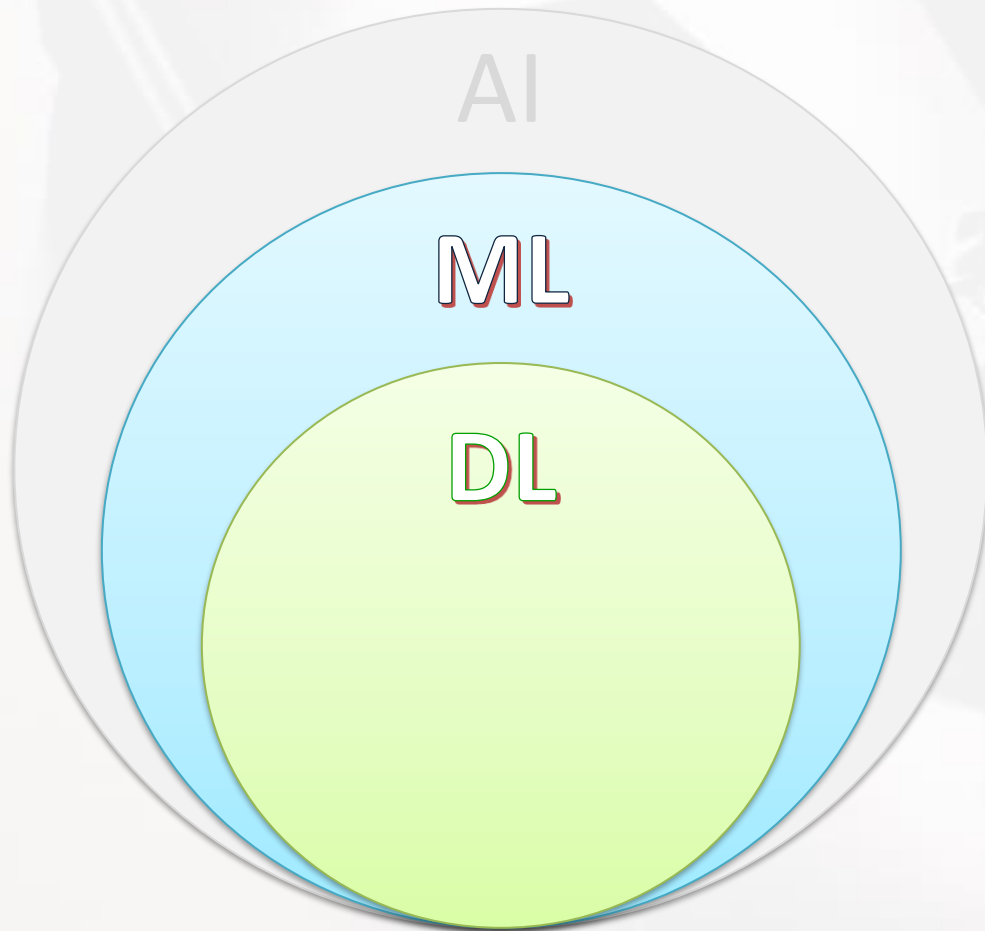
Machine Learning

The use of statistical methods to allow a piece of code or model to improve over time with experience

Deep Learning

The use of vast amounts of data, complex algorithms and deep neural nets to train a model

Is Early Warning AI? **No**



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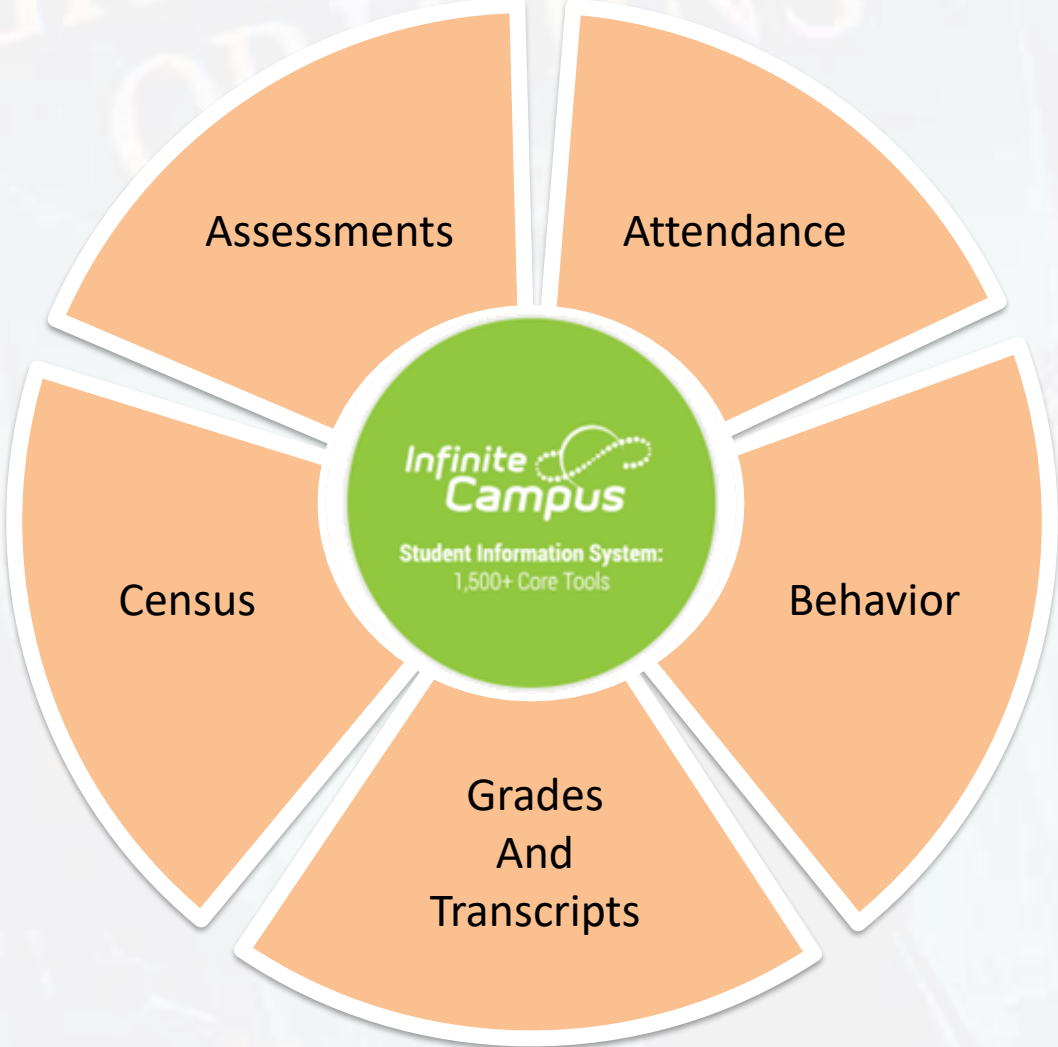
Is Early Warning AI? **No**

The Early Warning system utilizes a Machine Learned model to make its predictions. Machine Learning is a building block of AI, but the model is not artificially intelligent



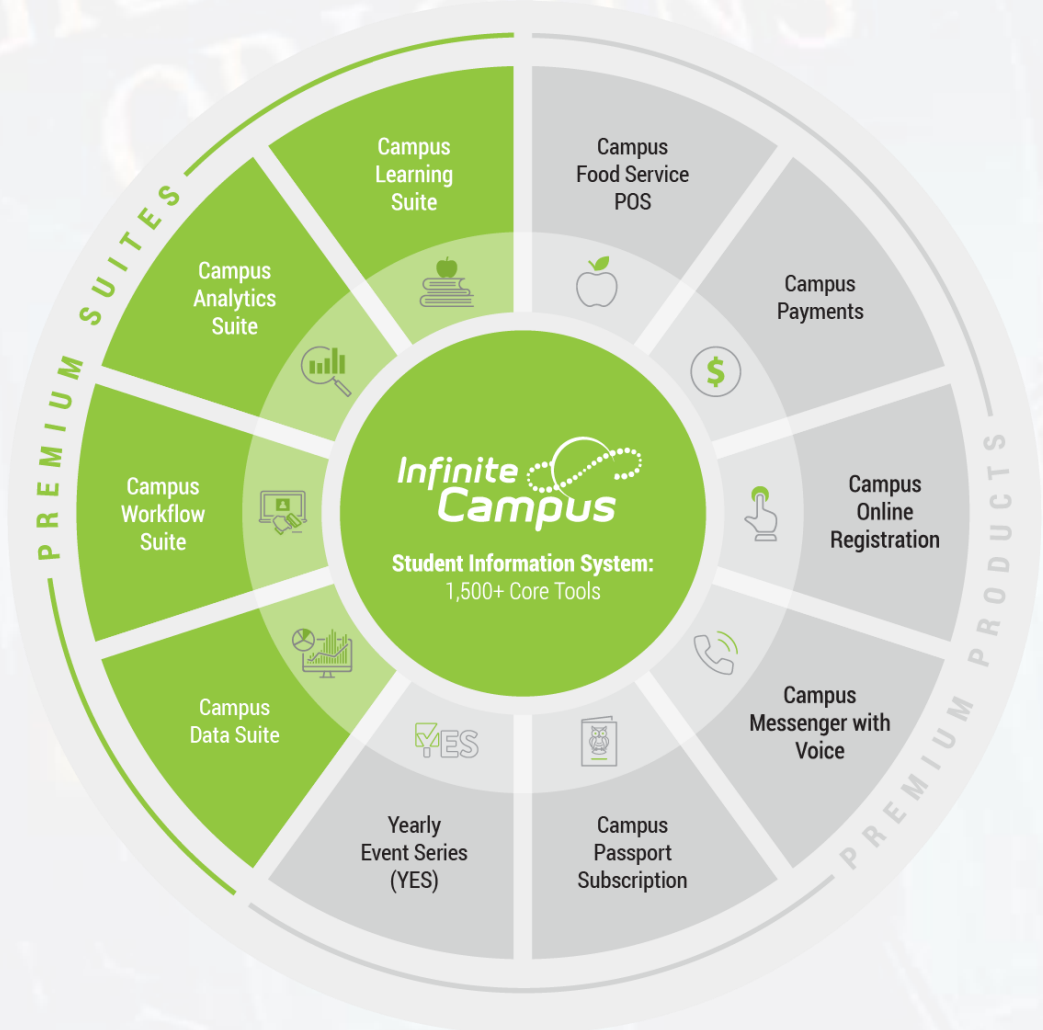
What is Machine Learning?

Starts with the SIS



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Starts with the SIS



Data Points

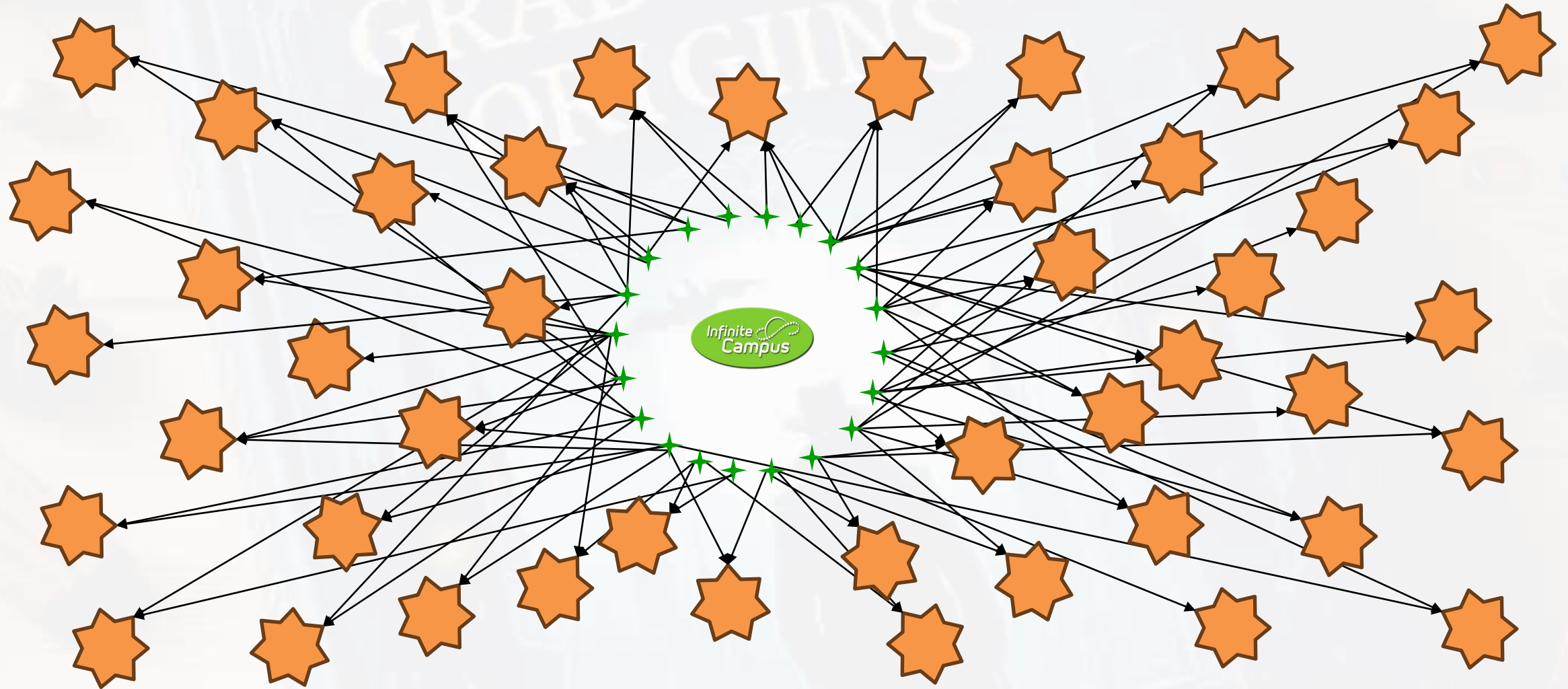


Data Combines to Features



One Data Point may be, and likely is, used in Multiple Features

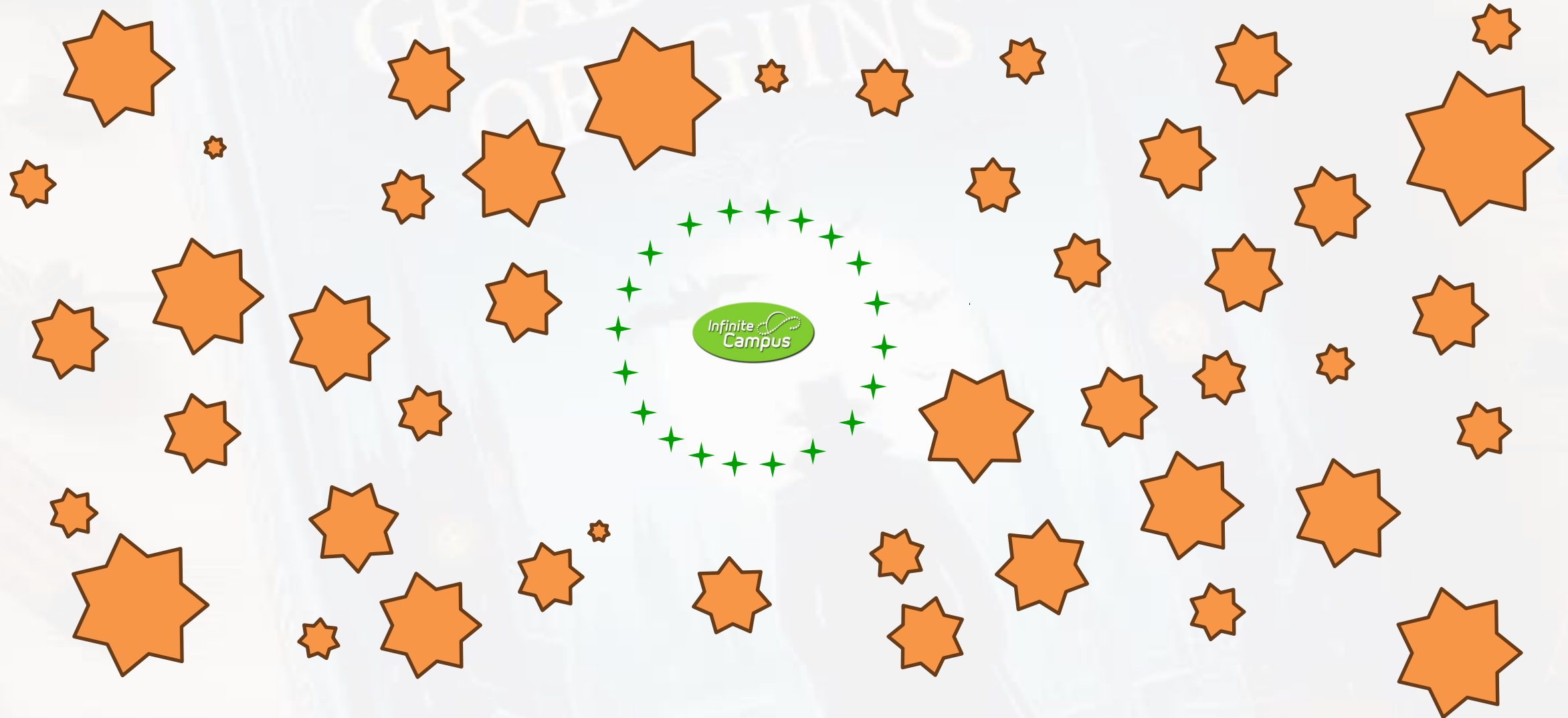
Data Combines to Features



Data Combines to Features



Features are evaluated for predictive quality



Features are evaluated for predictive quality

Examples of features with a high predictive value

- Guardian Portal Logins / Unexcused Absences
- Number of times students changed schools in the same District
- GPA / Sequential Unexcused Absences

Features are evaluated for predictive quality

Examples of features with a low predictive value

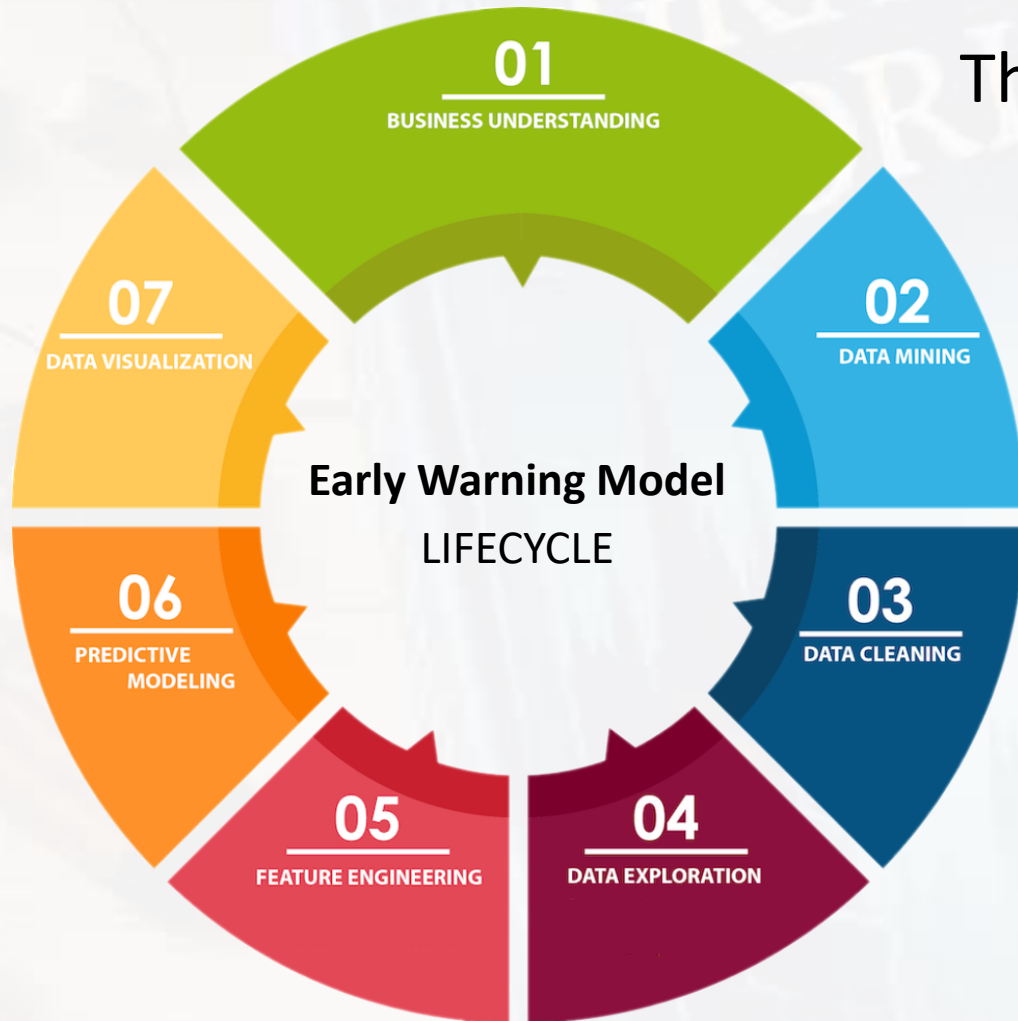
- Last Year's Enrollment Start Status
- FRAM Status
- SPED Status

Continuous Improvements

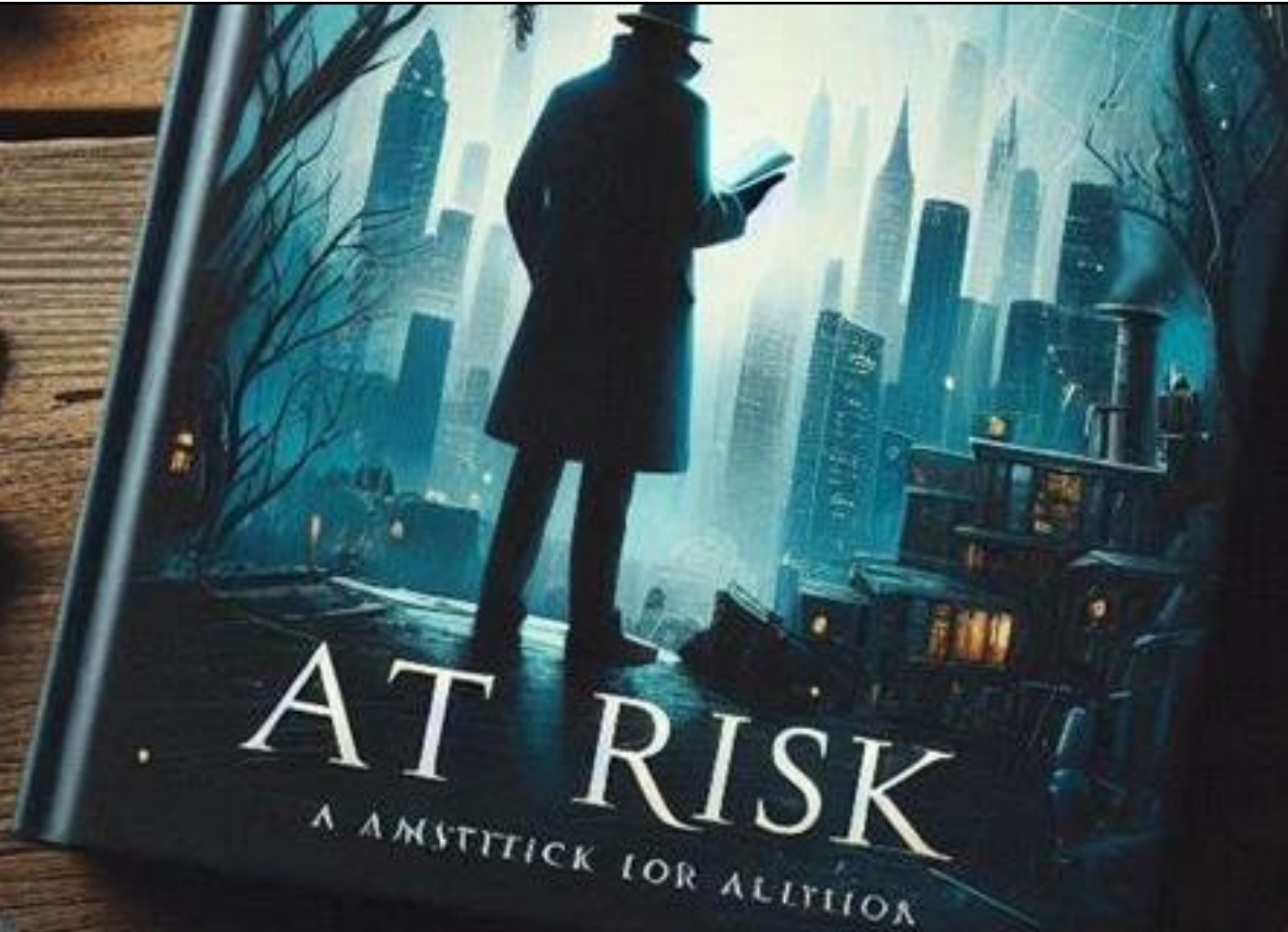
The Early Warning Model is updated Annually

This may include:

- Inclusion of additional years of Student Data
- Changes to specific data points being evaluated
- Addition or removal of Features
- Accommodations for Feature volatility



How is the GRAD Score Calculated for a Student?



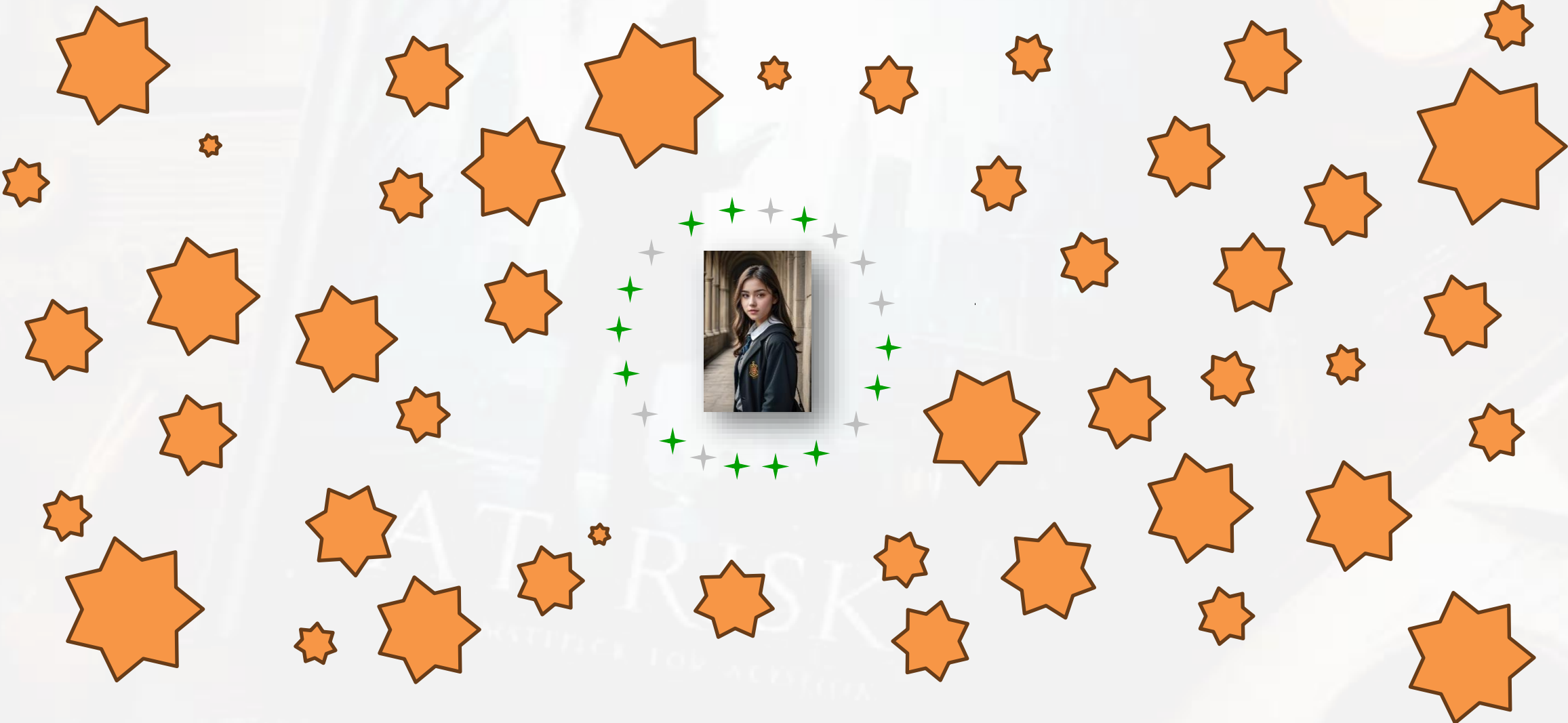
Data Points



Data Points



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Only the features that are linked to data that exists for the student are considered



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Features are evaluated in order based on how predictive they are in something called a "Decision Tree"



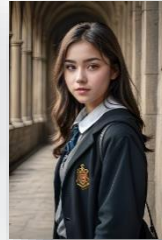
Category Scoring

In addition to the Composite Grad Score, individual GRAD scores are calculated for the following categories:

- Attendance
- Behavior
- Curriculum
- Stability

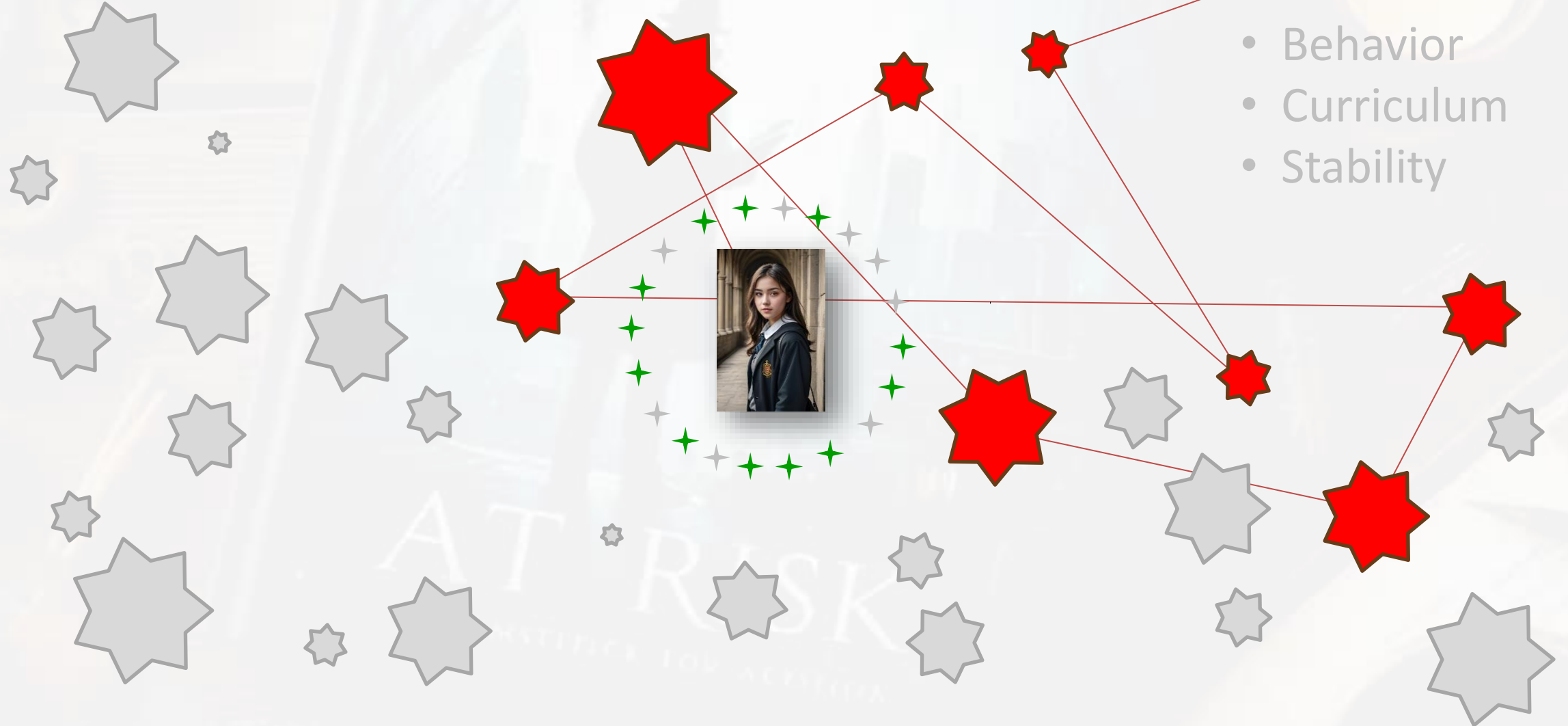
Category Scoring

- Attendance
- Behavior
- Curriculum
- Stability

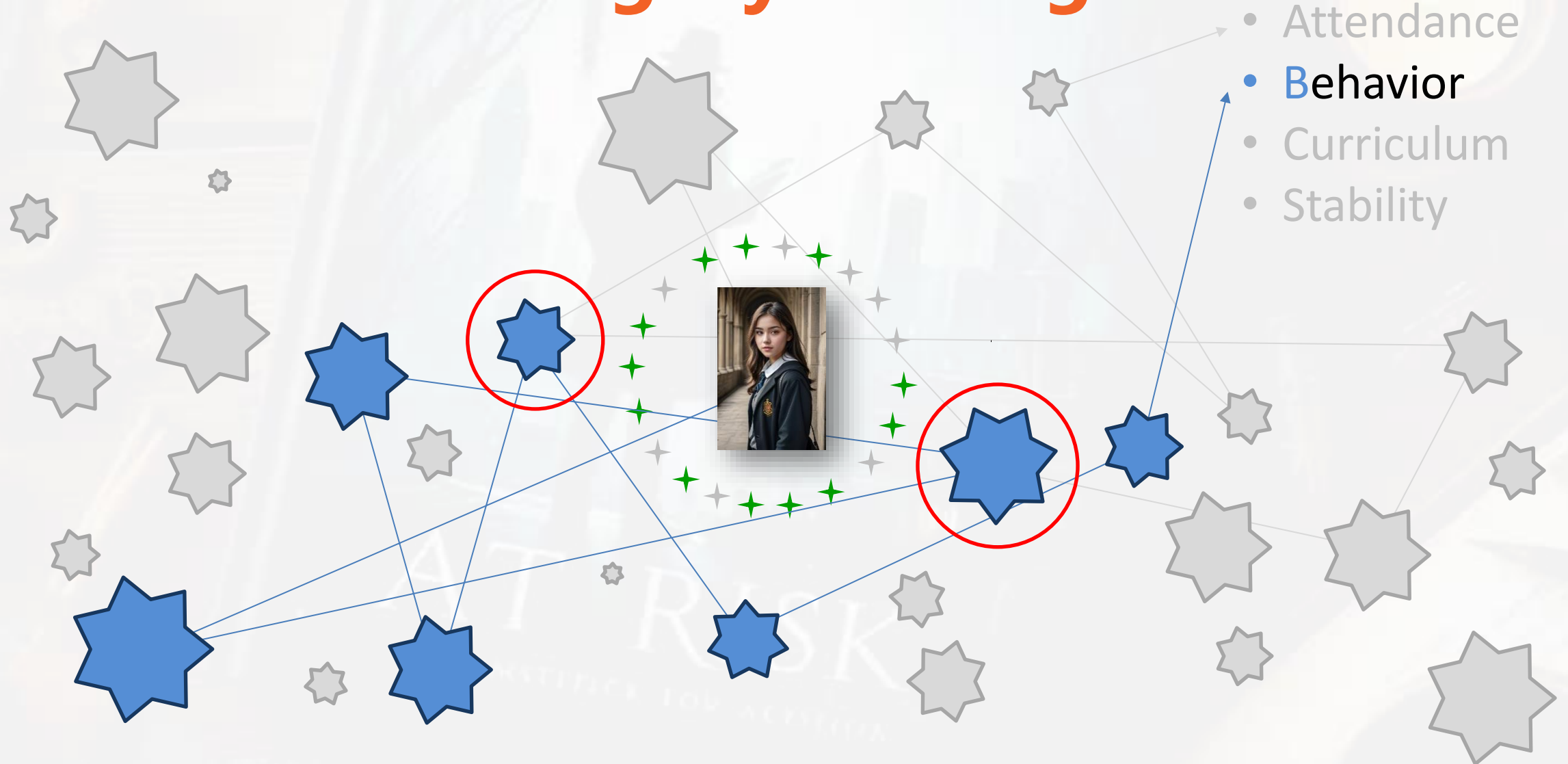


Category Scoring

- Attendance
- Behavior
- Curriculum
- Stability



Category Scoring



Student-Specific Evaluation

Because students have different data points, as well as different occurrences of these data points (count of behavior, attendance, etc) – and because the features are built from the combinations of multiple data points, the decision tree and feature importance will also be different from student to student.



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Missing 5 days of school in a row may be identified as an attendance risk factor for a student who also has a low GPA and a pattern of certain behaviors, but NOT a risk factor for a student with a High GPA and a different pattern of behavior



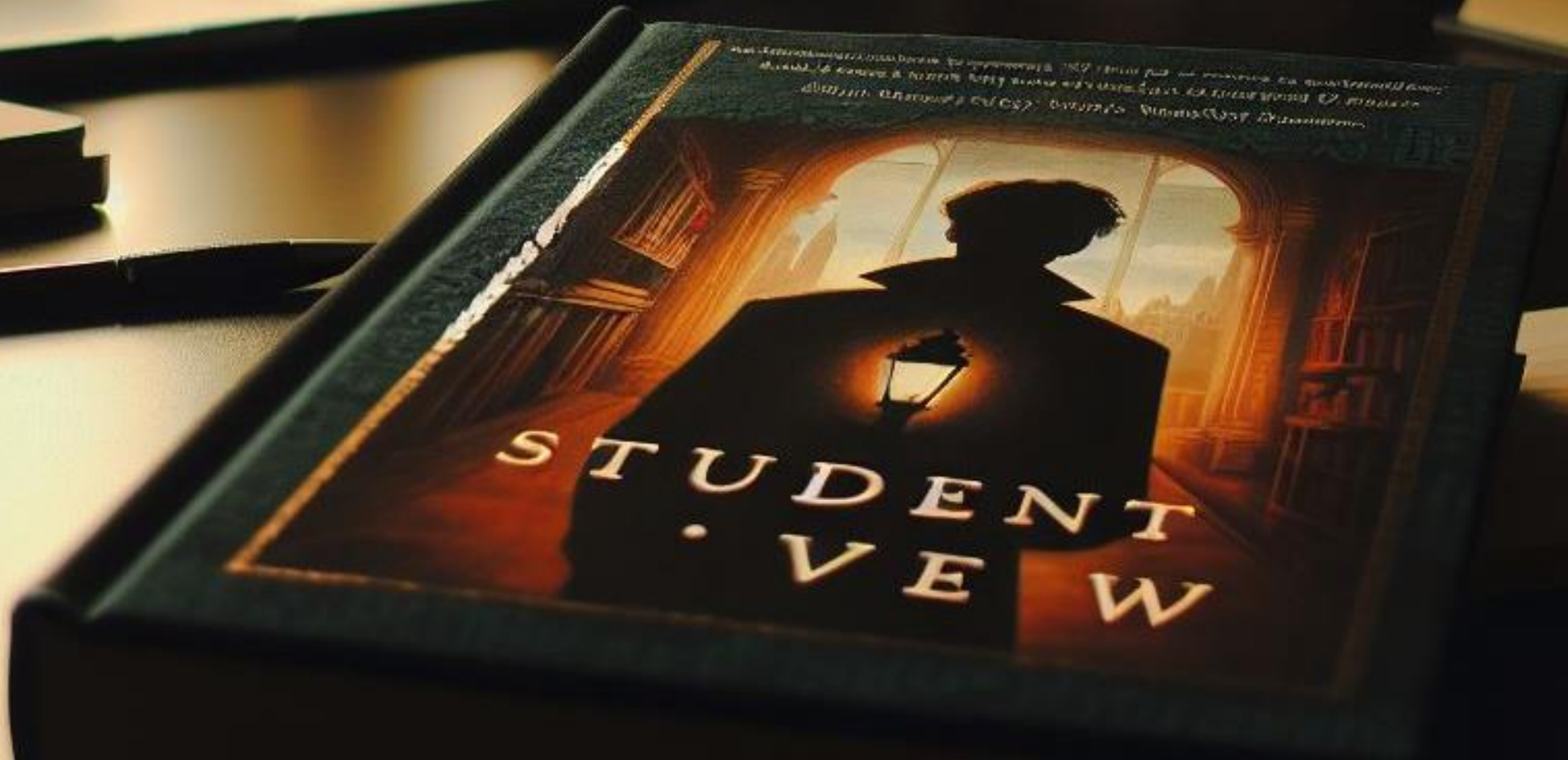
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Most Important Thing to Remember

Early Warning GRAD Score IS NOT a Calculator/Spreadsheet

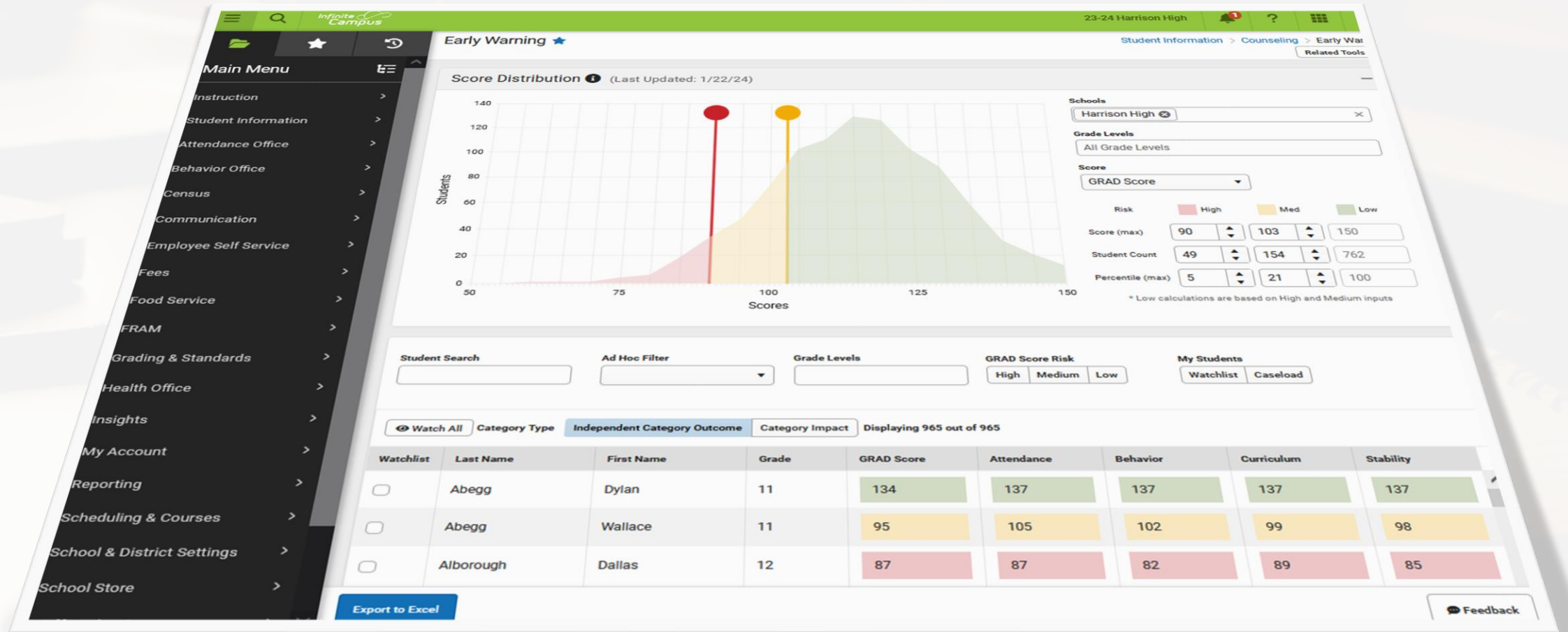
| Attribute | Score | Weight | Adjusted Value |
|-------------------|-------|--------|----------------|
| Days Absent | 3 | 1 | 3 |
| Assignments | 4 | 0.8 | 76 |
| GPA Score | 4 | 1 | 2.4 |
| Race/Ethnicity | 4 | 0.7 | 4.2 |
| Gender | 2 | 0.5 | 1 |
| Transcript | 3 | 1 | 30 |
| Grade Level | 8 | 0.8 | 6.4 |
| GRAD Score | | | 123 |

What can I learn from the GRAD Score?

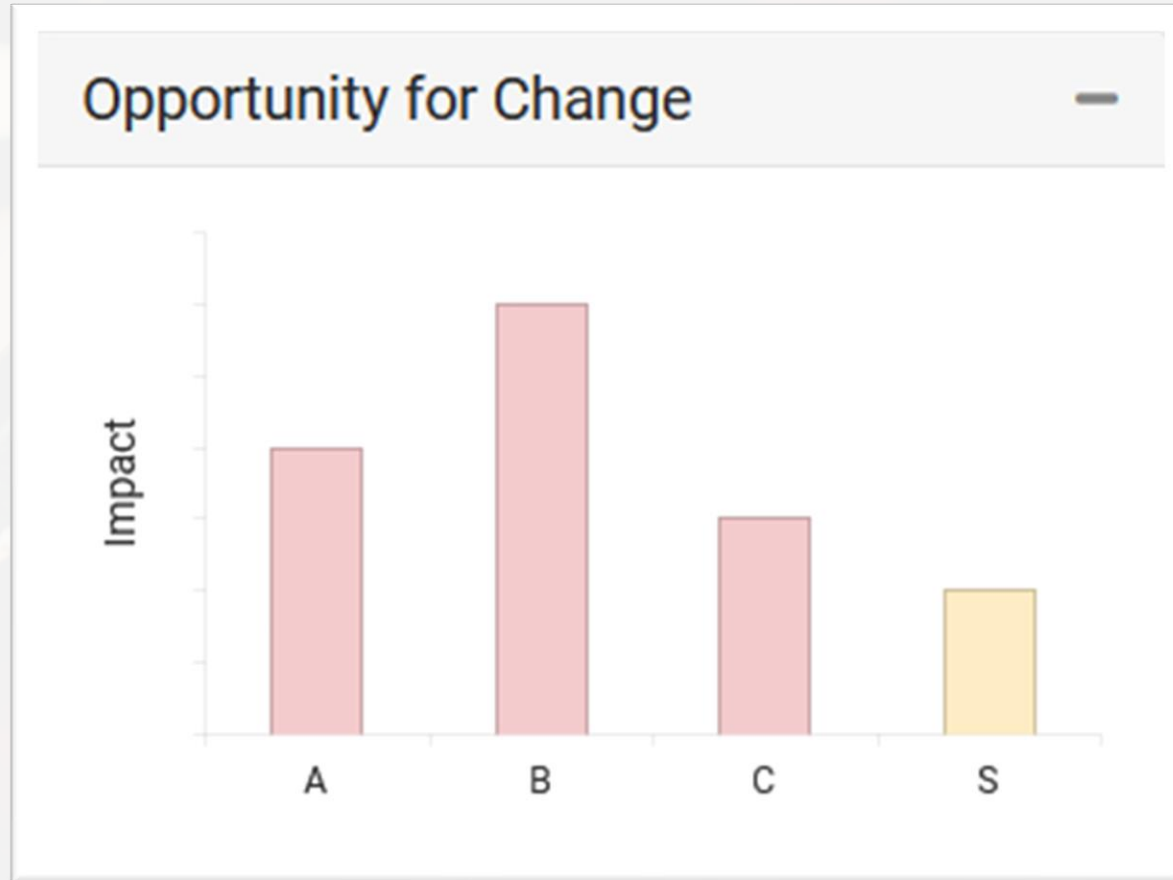


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Near-Realtime Student Risk Data

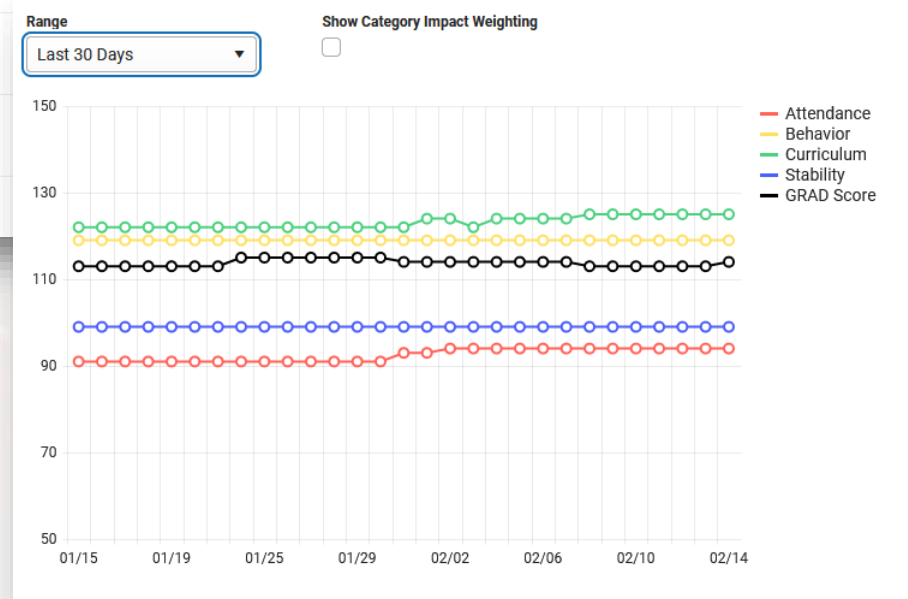
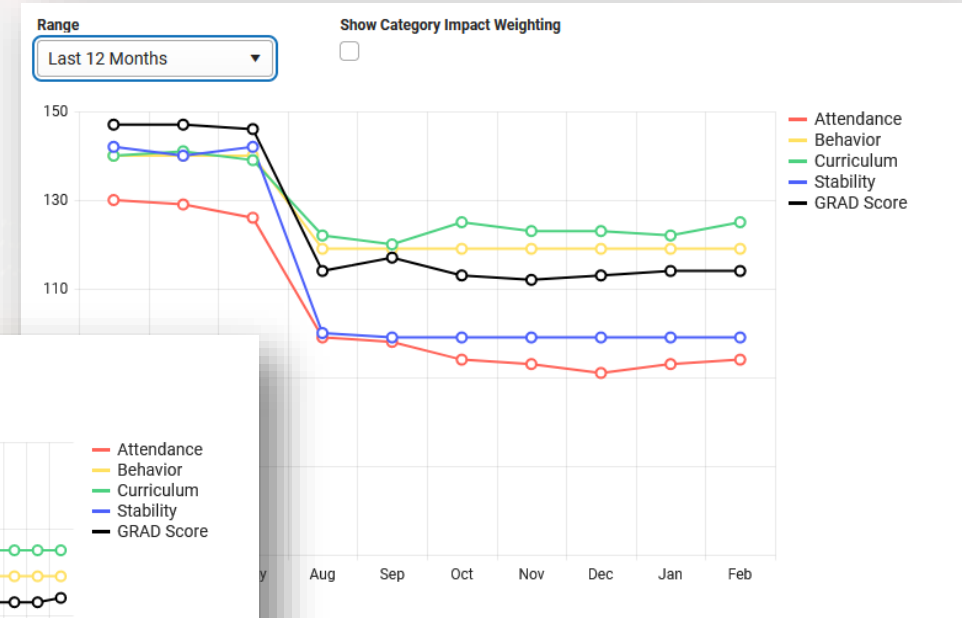
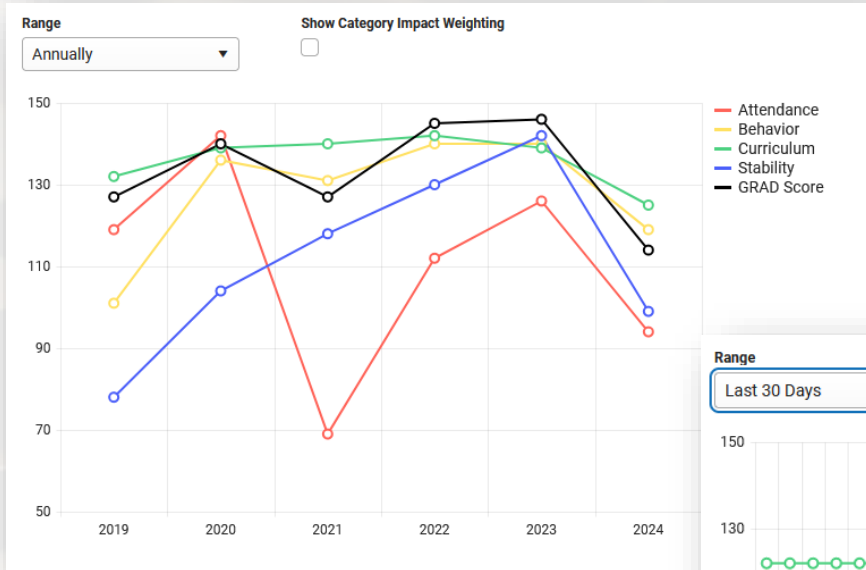


Areas for Impact



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Student Risk Changes Over Time



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Cohorts, Watchlists and Caseloads

Student Search Ad Hoc Filter Grade Levels 11 X GRAD Score Risk High Medium Low My Students Watchlist Caseload

Watch All Category Type Independent Category Outcome Category Impact Displaying 29 out of 408

| Watchlist | Last Name | First Name | Grade | GRAD Score | Attendance | Behavior | Curriculum | Stability |
|--------------------------|-----------|------------|-------|------------|------------|----------|------------|-----------|
| <input type="checkbox"/> | STUDENT | SAVANNAH | 11 | 56 | Low (-1) | Low (-0) | Low (-2) | Low (-3) |
| <input type="checkbox"/> | STUDENT | JACOB | 11 | 56 | Low (-0) | Low (-0) | Low (-0) | Med (-13) |
| <input type="checkbox"/> | STUDENT | CHARLES | 11 | 58 | Low (-0) | Low (-0) | Low (-2) | Low (-9) |
| <input type="checkbox"/> | STUDENT | KAITLYN | 11 | 58 | Low (-1) | Low (-0) | Low (-1) | Low (-7) |
| <input type="checkbox"/> | STUDENT | AUSTIN | 11 | 56 | Low (-0) | Low (-0) | Low (-0) | Low (-9) |
| <input type="checkbox"/> | STUDENT | SELINA | 11 | 57 | Med (-3) | Med (-1) | Low (-0) | Med (-12) |

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Insights

