



Adaptive Learning and the Power of Analytics

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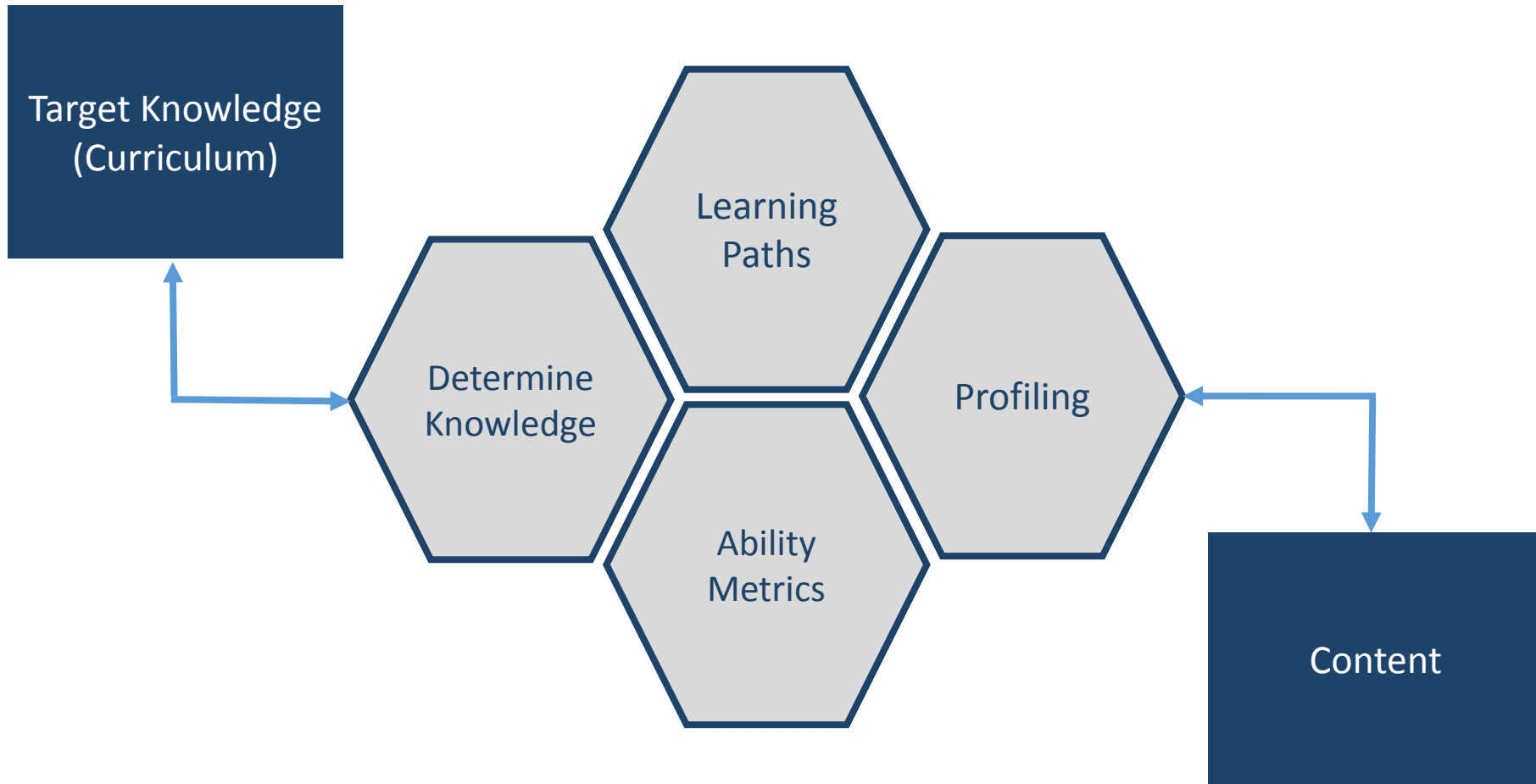
Outline

- Adaptive Learning Framework in Realizeit
 - How we achieve adaptivity
- Learning Analytics
 - Integrated into learning
 - Student and instructor views
 - Academic analytics examples
- Predictive Analytics
 - Example of behavior based early warning system
- Questions

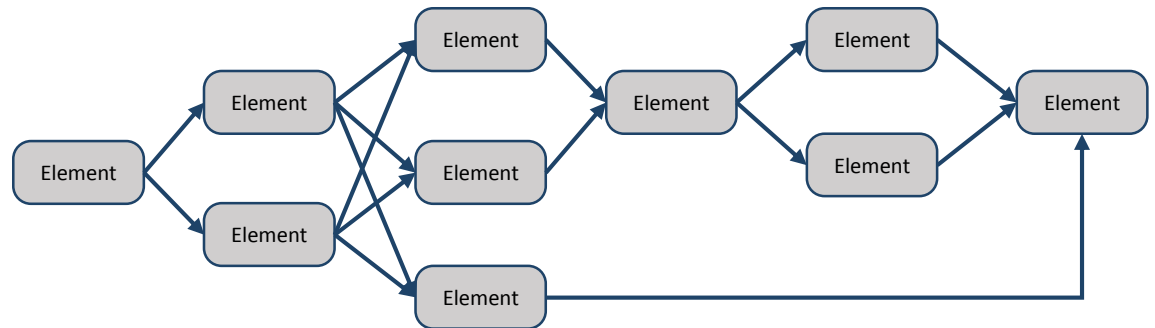
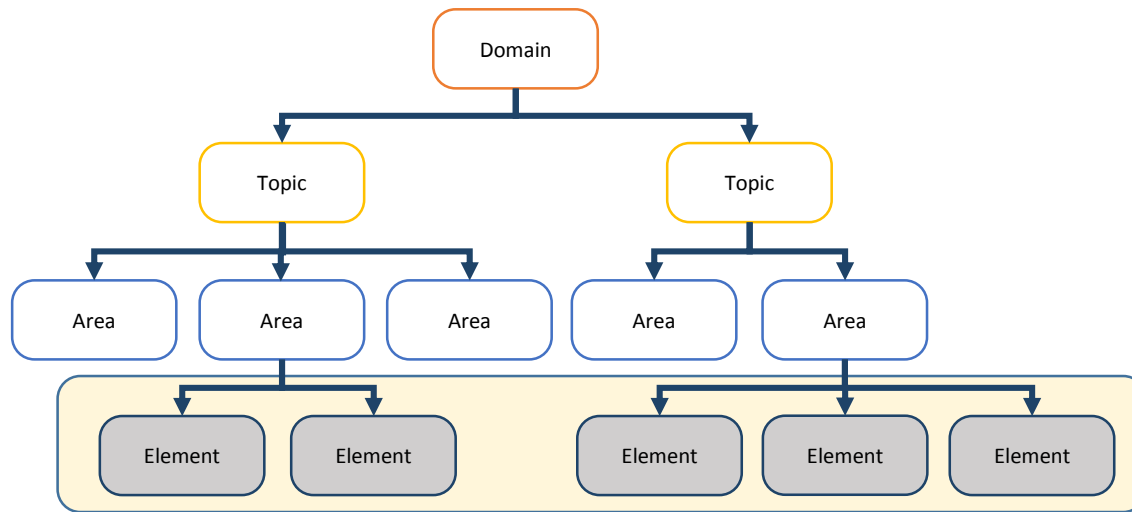
Product Vision

- Emulate a good teacher in one-on-one learning situation
- Provide an individualized learning experience
 - Deliver learning at an appropriate time
 - Deliver appropriate learning material
 - Learn about the learner
 - Manage and adapt to change: abilities, metrics, behavior etc.
 - Identify weaknesses and try to remedy
- Remain subject and content independent

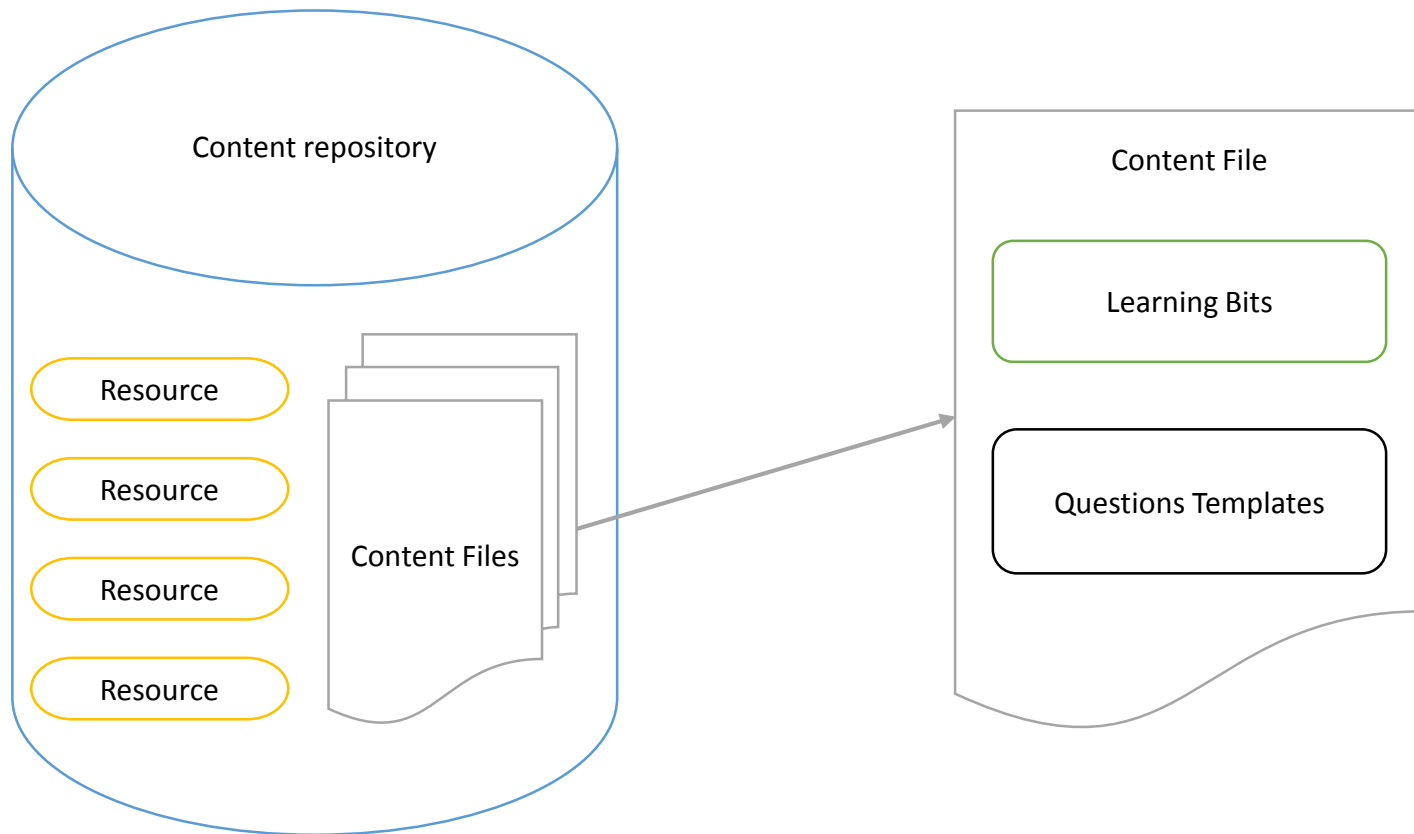
The Realizeit Learning Model



The Realizeit Learning Model



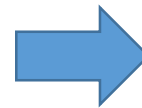
The Realizeit Learning Model



Content

- Learning Bits

- Introduction
- Learning Material
- Example
- Worked Example
- Interactive Example
- Questions
- Summary
- Review

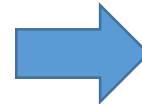


ILEWNQSR
LRQ
LSQNILQR
LSIWQR

- Questions

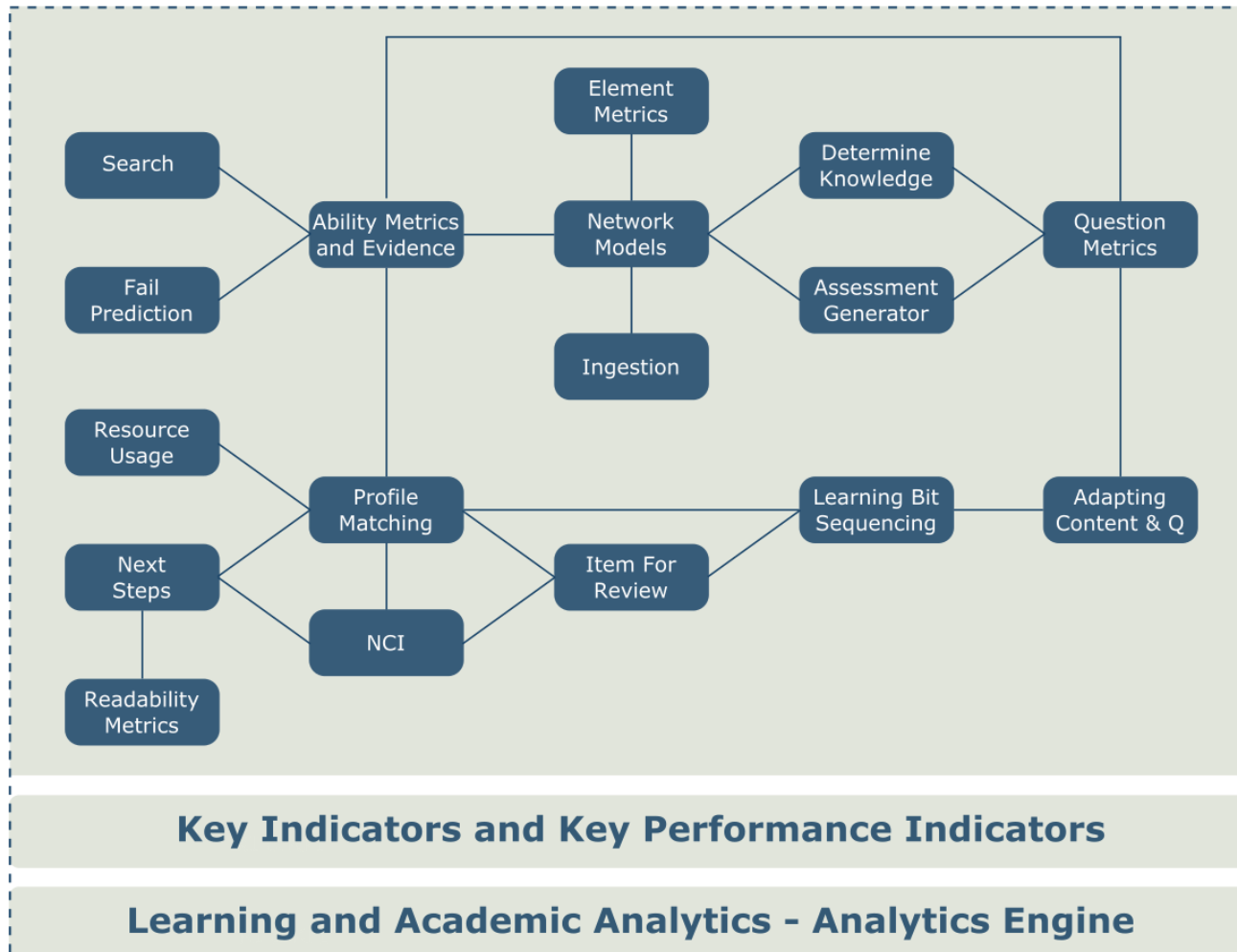
Template:
 $ax + bx = ?$

Conditions:
 $1 \leq a \leq 10$
 $1 \leq b \leq 7$



$2x + 7x = ?$
 $x + 3x = ?$
 $10x + 2x = ?$

The Realizeit Learning Model



Adaptivity in Realizeit

- Tailor a student's start position
- Alter a learner's pathway in real-time
- Select the most suitable content
- Select the most suitable pedagogical elements
- Adapt the content in real-time
- Main source of adaptivity comes from the adaptive intelligence engine
- Adaptivity and personalization also come from the instructor or the student themselves

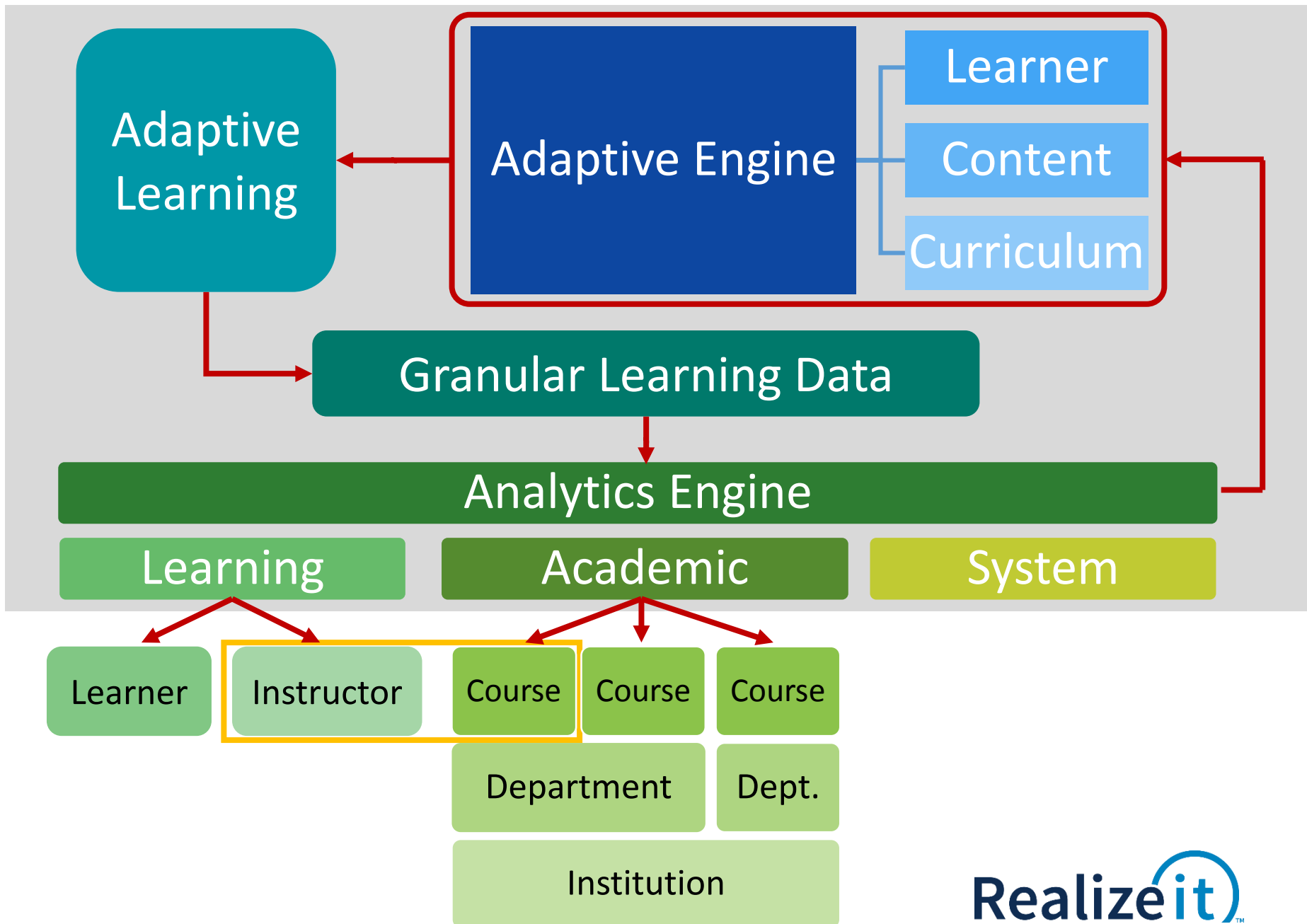
Learning and Academic Analytics

TYPE OF ANALYTICS	LEVEL OR OBJECT OF ANALYSIS	WHO BENEFITS?
Learning Analytics	Course-level: social networks, conceptual development, discourse analysis, “intelligent curriculum”	Learners, faculty
	Departmental: predictive modeling, patterns of success/failure	Learners, faculty
Academic Analytics	Institutional: learner profiles, performance of academics, knowledge flow	Administrators, funders, marketing
	Regional (state/provincial): comparisons between systems	Funders, administrators
	National and International	National governments, education authorities

Penetrating the Fog

Phil Long and George Siemens

Educause Review, September/October 2011



Learner and Instructor Analytics

- Go beyond the traditional summative metrics – adaptive analytics
- Based on both attainment and behavioral metrics
- Easily understandable
- Actionable
- Integrate analytics into the functionality and use of the system
- Makes use of data as it is gathered and generates *real-time* analytics

Learner



Wilber

[My Courses](#) / [Fundamentals of Management](#) / Introduction to Management



Objective

Introduction to Management

40 mins
Estimated Time Left

Competent

13/15
Progress

Notifications

Contact

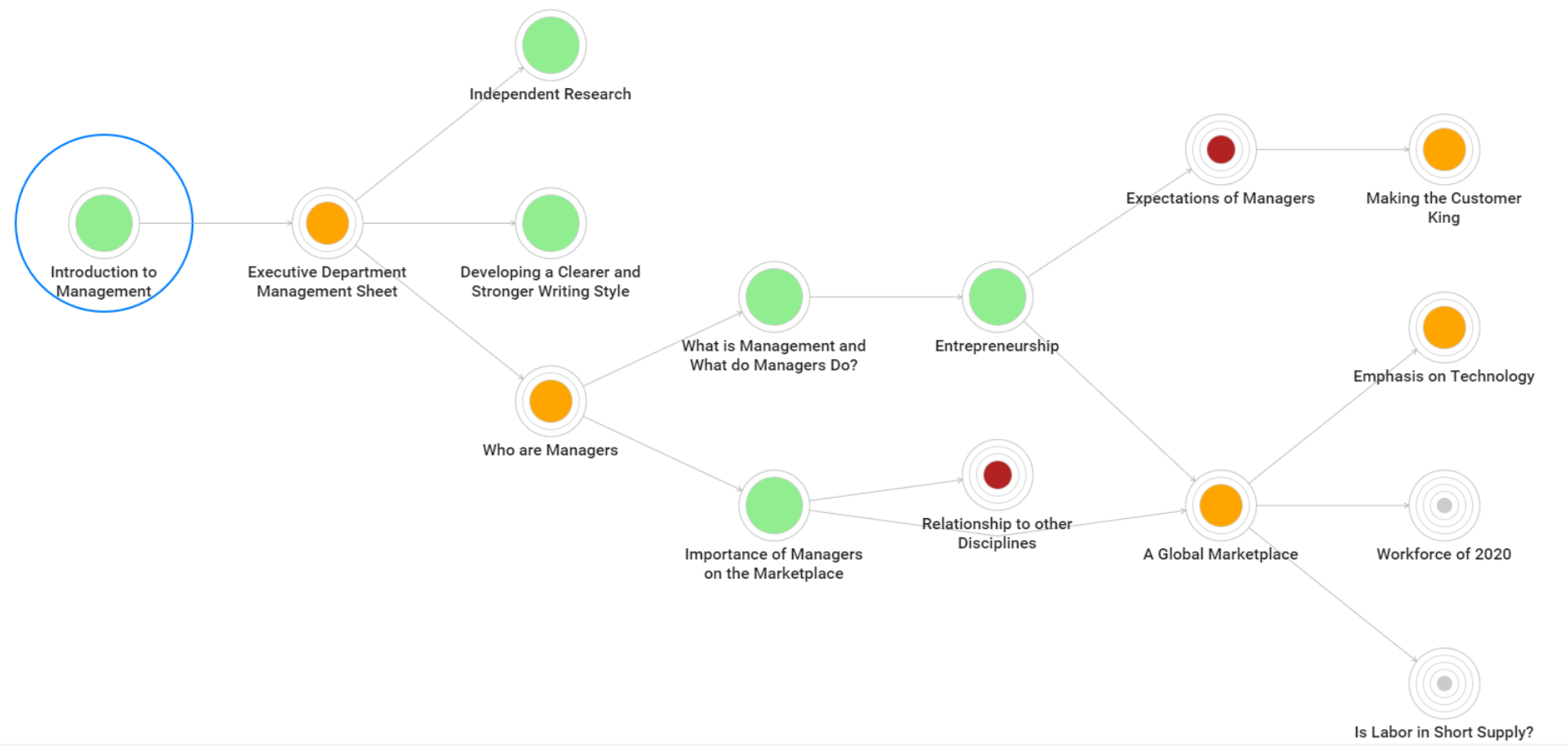
Menu

Next step

Revise

Introduction to Management

Alternative steps



LIST KEY

Locked Available Beginner Improving Competent Expert Master

Instructor



/ Roles / Courses / Common Core Standards: Grades 6 and 7 (Start at 6) / Spring 2016 / Algebraic Notation and Expressions

Milestone Algebraic Notation and Expressions

Expert

2/10
Students

0/13
Items

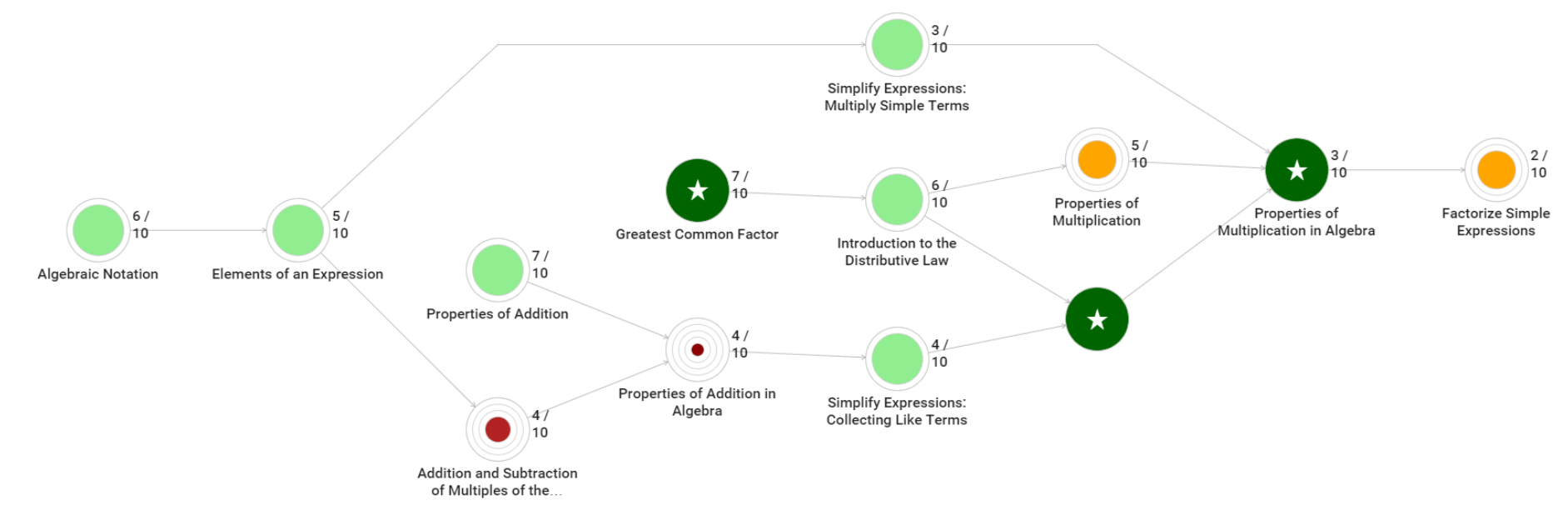
Grading

Send Message

Watch

Menu

Basic data New Messages Analysis Map



Grouping Map Show student map

LIST KEY

Beginner

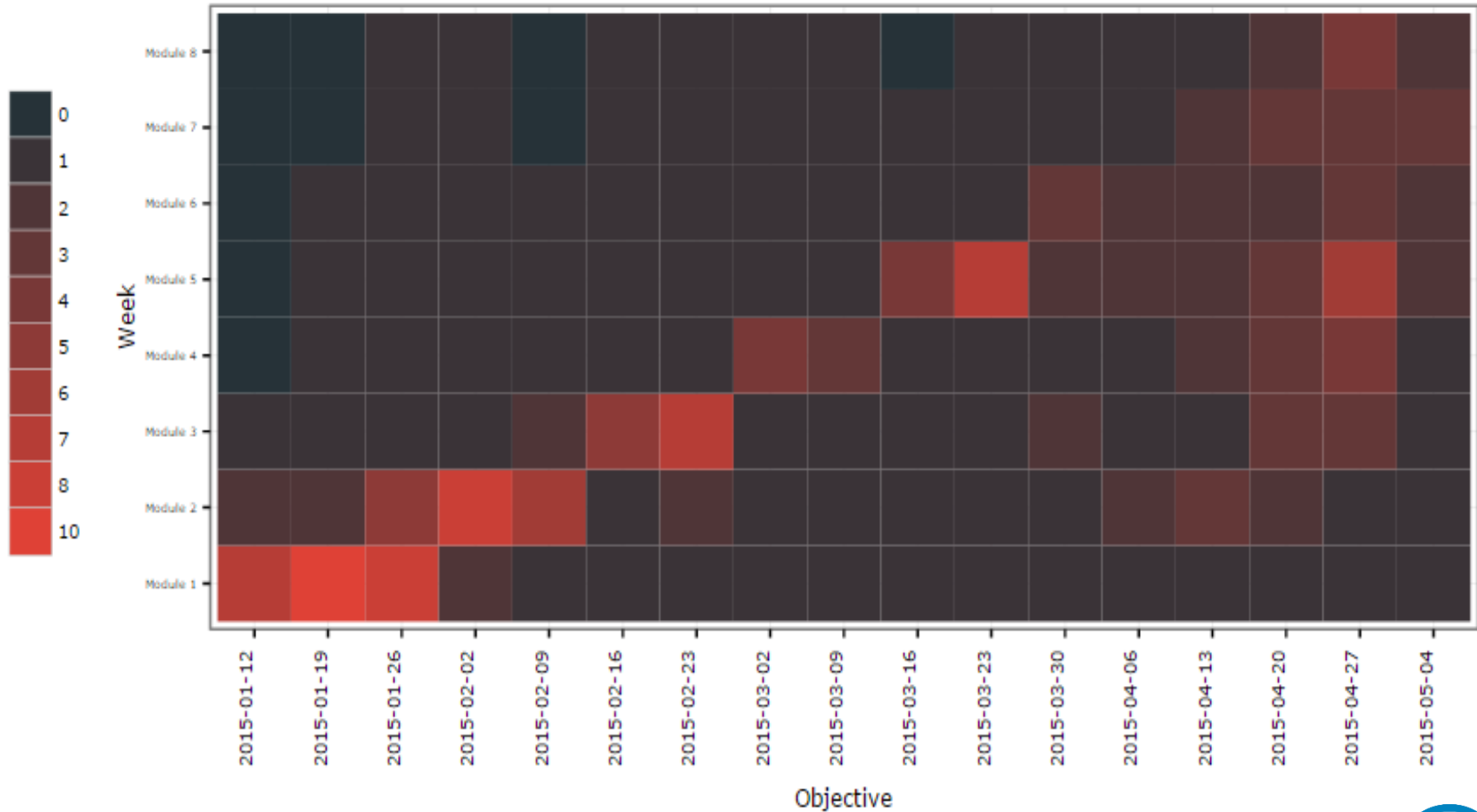
Improving

Competent

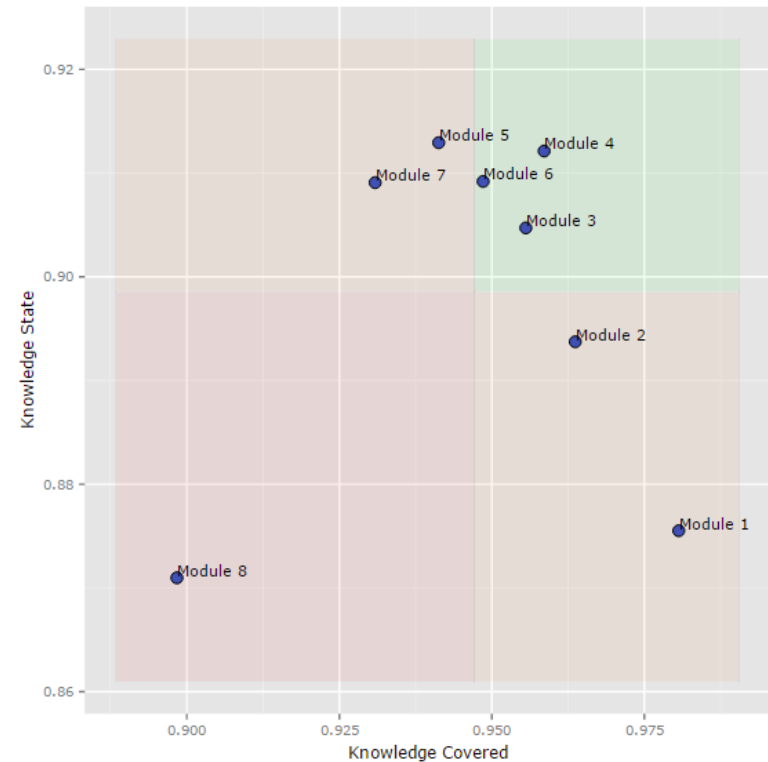
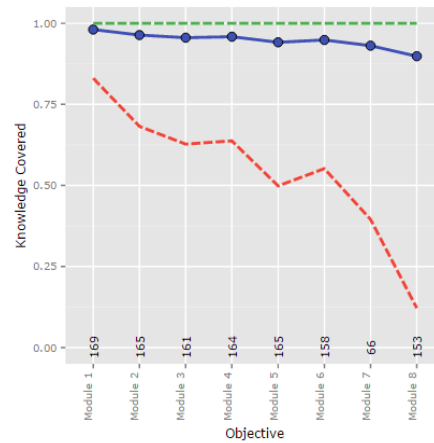
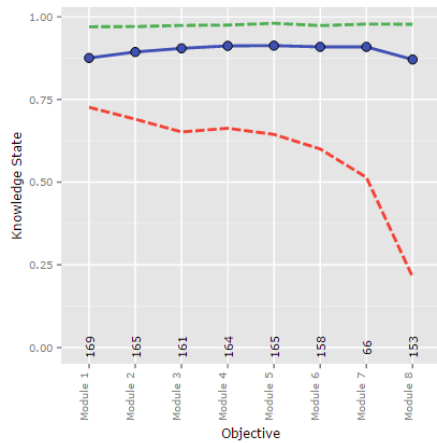
Expert

Master

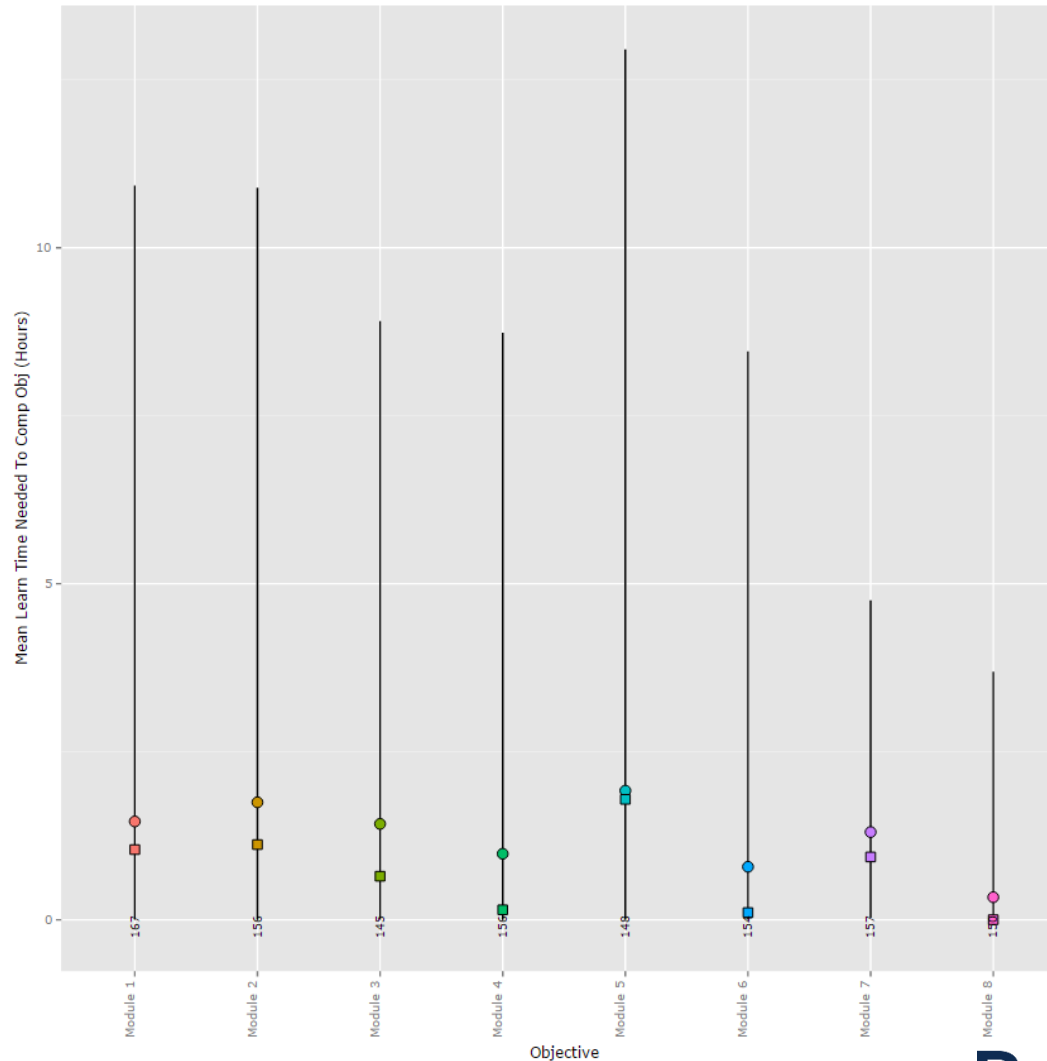
Instructor – Analytics Examples



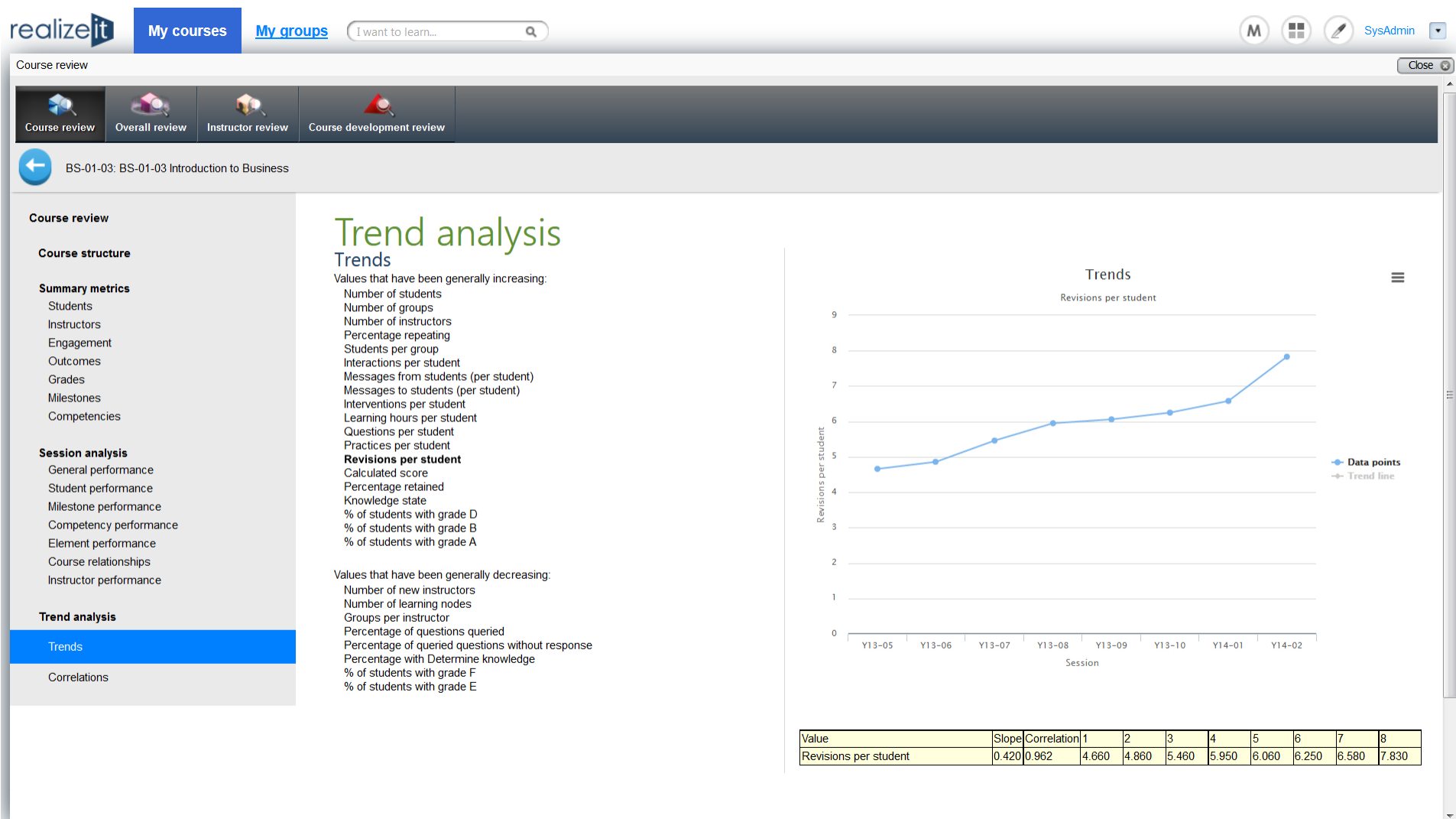
Instructor – Analytics Examples



Instructor – Analytics Examples



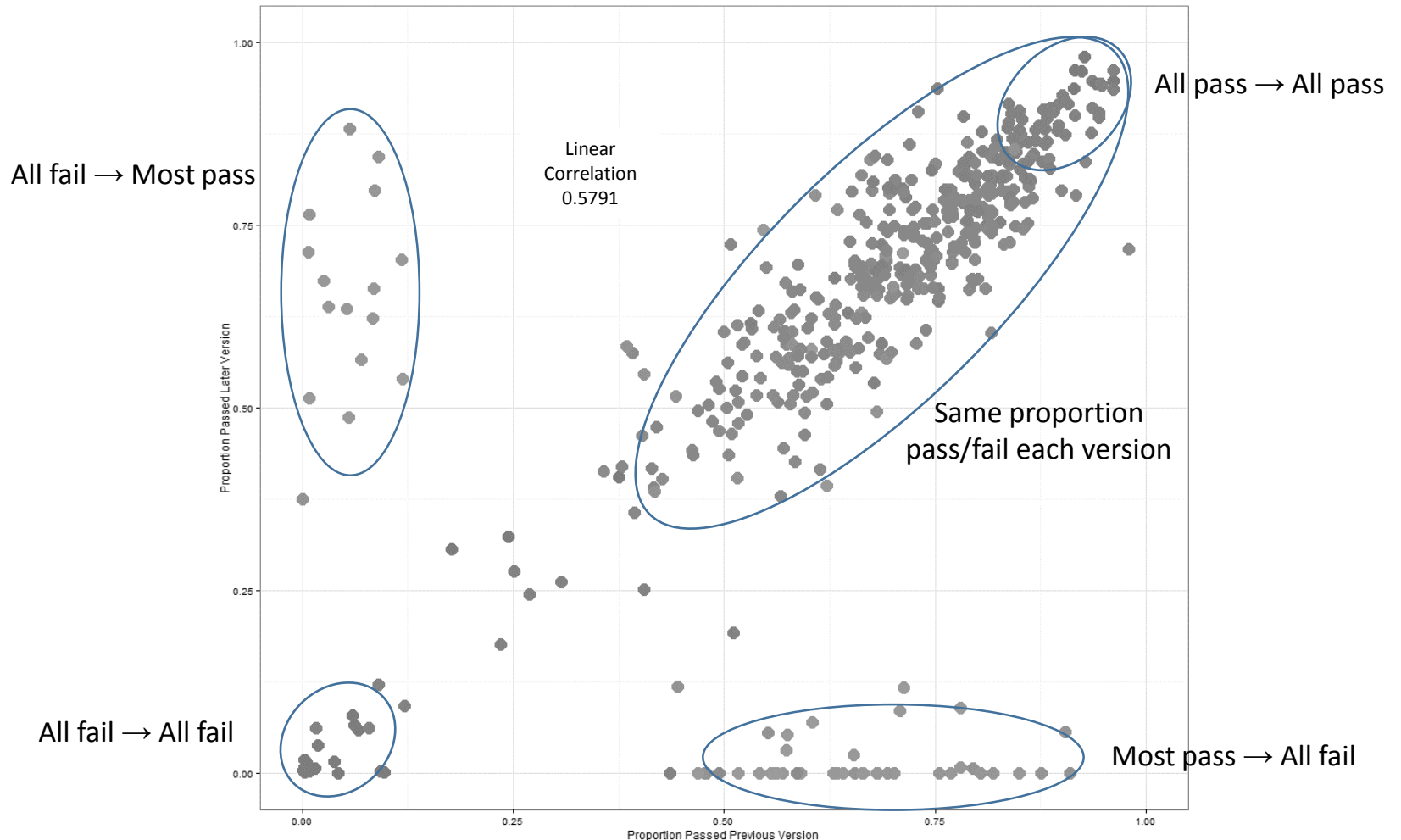
Course Review – Analytics Examples



Prediction of Student Success

- To build an early warning system to detect at-risk students
- Traditionally built on demographic - cannot change
- Focus on learner behaviour - one week of learning
- From the predictions and associated probabilities determine a list of at-risk students.
- Take action - Inform the appropriate stakeholders and provide appropriate remediation
- Monitor the impact of any intervention

Relationship between versions



Dependant Variable

- **FinalGrade** - The average student grade across the course objectives at the final due date.
- **PassFail** - A binary version of FinalGrade

$$PassFail = \begin{cases} 1 & \text{if } FinalGrade \geq 60\% \\ 0 & \text{otherwise} \end{cases}$$

Independent – Course Level

Behavior

- **Activity1** – First activity on course
- **Activity2** – Second activity on course
- **ActivitiesEachDay** – Total number of learning activities
- **AvgActivitiesEachDay** – Average number of activities each day
- **DayStarted** – Number of days before finish date that student started learning
- **Days** – number of days on which learning activities were delivered
- **NumPrevCourses** – The number of previous courses in which the student was enrolled.

Independent – Course Level

Attainment

- **Grade** – Final grade if calculated at this point in time
- **GradePF** – Binary version of Grade.
- **QuestionsAsked** – Num of questions
- **QuestionsCorrect** - Num correct
- **QuestionsIncorrect** – Num incorrect
- **PerQuestionsCorrect** – Percent correct
- **PerQuestionsIncorrect** – Percent incorrectly

Independent – Objective Level

Attainment

- **Obj_*i*_Grade** – The grade the student.
- **Obj_*i*_KC** – The knowledge covered.
- **Obj_*i*_KS** – The knowledge state.
- **Obj_*i*_Pass/Fail** – A binary version of the objective grade.

Behaviour

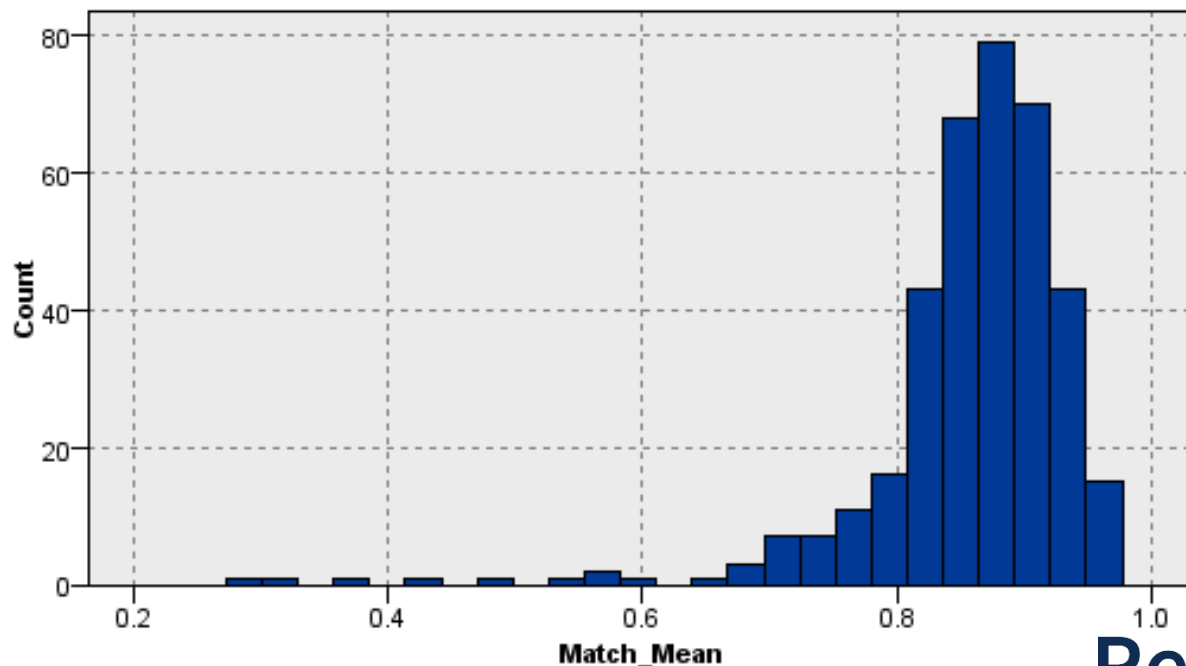
- **Obj_*i*_Learning Prop** – The proportion of activities that are learning activities (before mastery).
- **Obj_*i*_Revision Prop** – The proportion of activities that are revision activities (after mastery).
- **Obj_*i*_TimePerNode** – The average amount of time per node.

Model Accuracy – Unseen Data

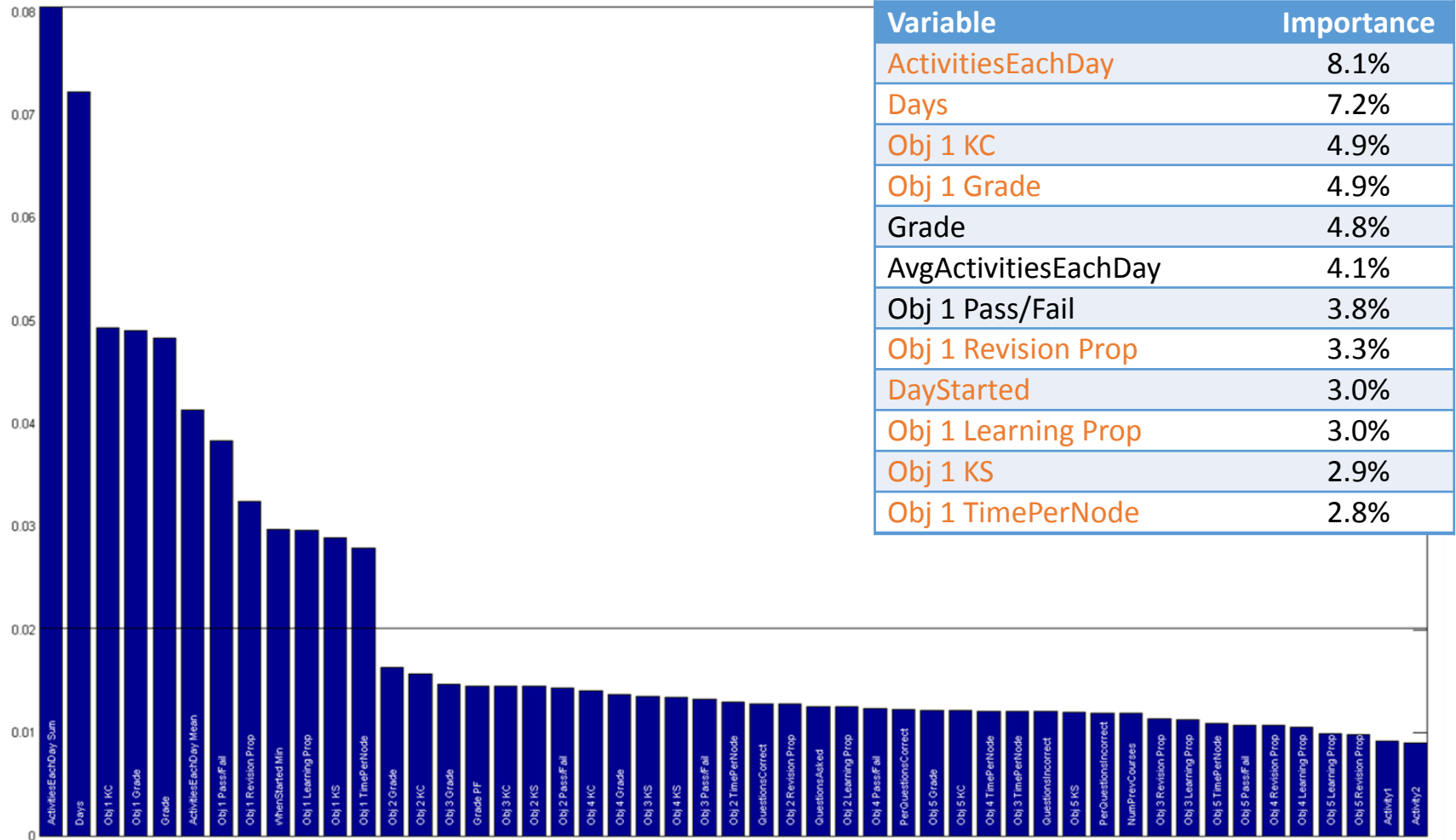
		Predicted	
		Fail	Pass
Observed	Fail	22.35%	7.11%
	Pass	7.98%	62.57%

Fit - 94.23%

Accuracy - 84.92%



Predictor Importance



Simplified Model

Behavioural Metrics

- Activities Each Day
- Days
- Day Started
- Obj 1 Revision Prop
- Obj 1 Learning Prop
- Obj 1 Time Per Node

Attainment Metrics

- Obj 1 KS
- Obj 1 KC
- Obj 1 Grade

		Accuracy	
		Predicted	
		Fail	Pass
Observed	Fail	20.92%	8.53%
	Pass	6.73%	63.82%

Unseen Data

- Full Model: 84.92%
- Simplified Model: 84.74%

Sample model

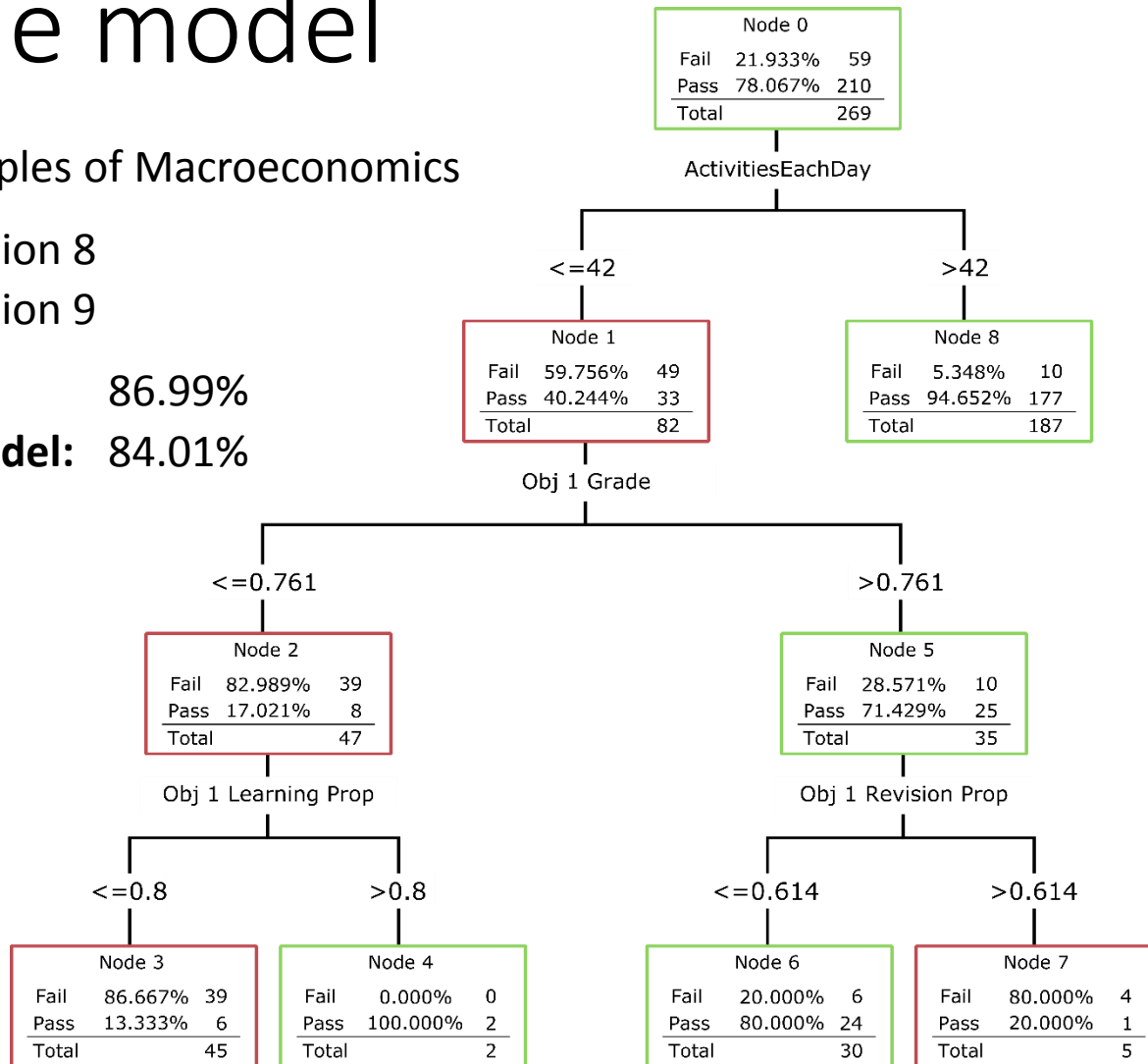
Course: Principles of Macroeconomics

Training: Version 8

Test: Version 9

Full model: 86.99%

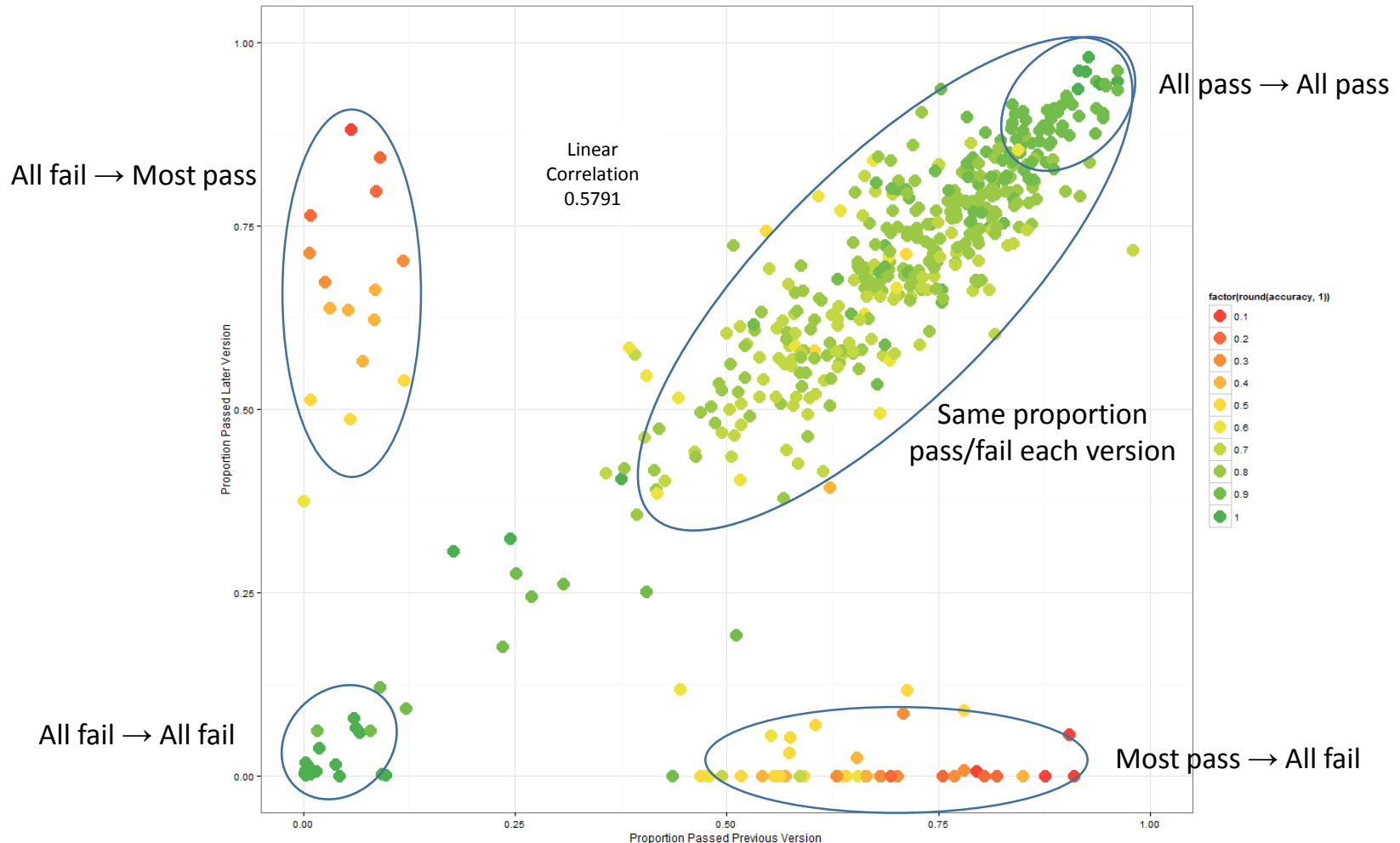
Simplified model: 84.01%



Insights

- Model based on behavioural and attainment metrics.
- Models allows insights into what behaviours make a student successful and what puts them at risk.
- Behavioural Metrics
 - Activities Each Day
 - Days
 - Day Started
 - Obj 1 Revision Prop
 - Obj 1 Learning Prop
 - Obj 1 Time Per Node
- Attainment Metrics
 - Obj 1 KS
 - Obj 1 KC
 - Obj 1 Grade

Overall Prediction Accuracy



Summary – three main takeaways

- Adaptive Learning can provide unique fine grained data which can power new insights and analytics
- Analytics in real-time, actionable and integrated into the use of the system
- Adaptive learning can power predictive analytics based on learner behavior



Learn More

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