

**Towards integration of Learning Analytics
and intelligent online assessment service:
Nordic passion, WISEflow case**



OMAP

Online Massive Assessment Platform

For stress-less school



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What is the OMAP we propose?

Definition



OMAP covers all the tests.

What does stress-less school mean?

Problem & Value

Money stress



Security stress

stress-less school

OMAP makes all easier

stress

stress



Teachers



Students

Who are our customers?

Target



Take away all the stress in schools.



Teachers



Students

How much?

Business model



school



A License fee per year/semester



Prices may vary according to schools size (no of students)

How do I go to market?

GO-TO MARKET PLAN



Recruit Pilot schools

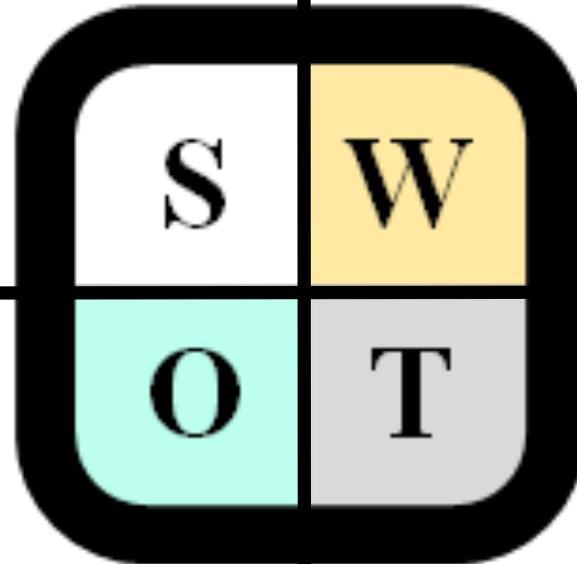


**Search Promotion & Sales
partner**

What superiorities does OMAP have?

SWOT Analysis

OMAP



Versatile

Unfriendly attitude toward Foreign services

Security

Competing with SI

Cost-saving

Teachers want to make exams easy

Red ocean

Schools want secure online exams

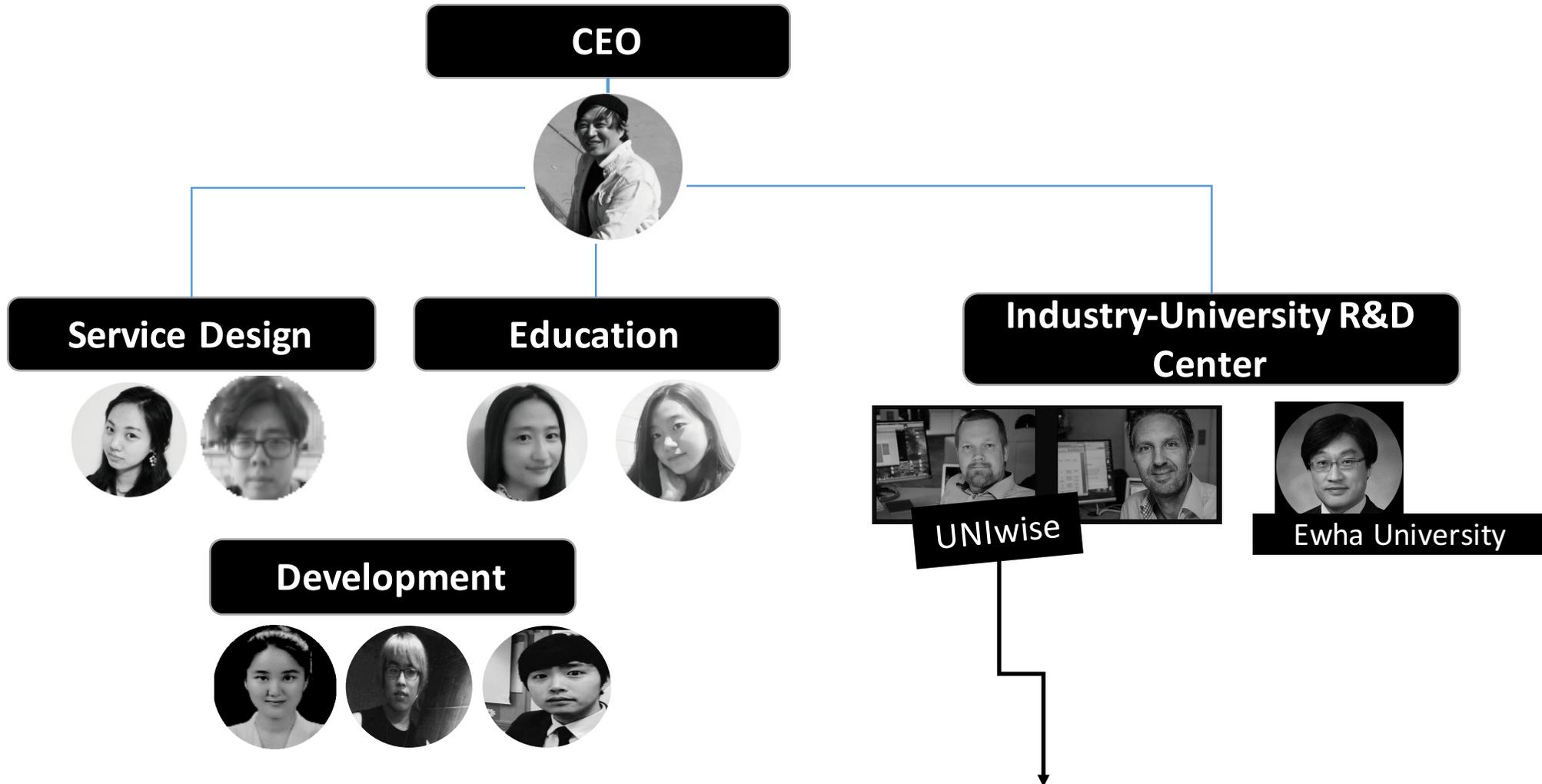
Teachers hate technologies

Schools want cost-saving

independence

Who are we, DUCOgen?

WHO WE ARE



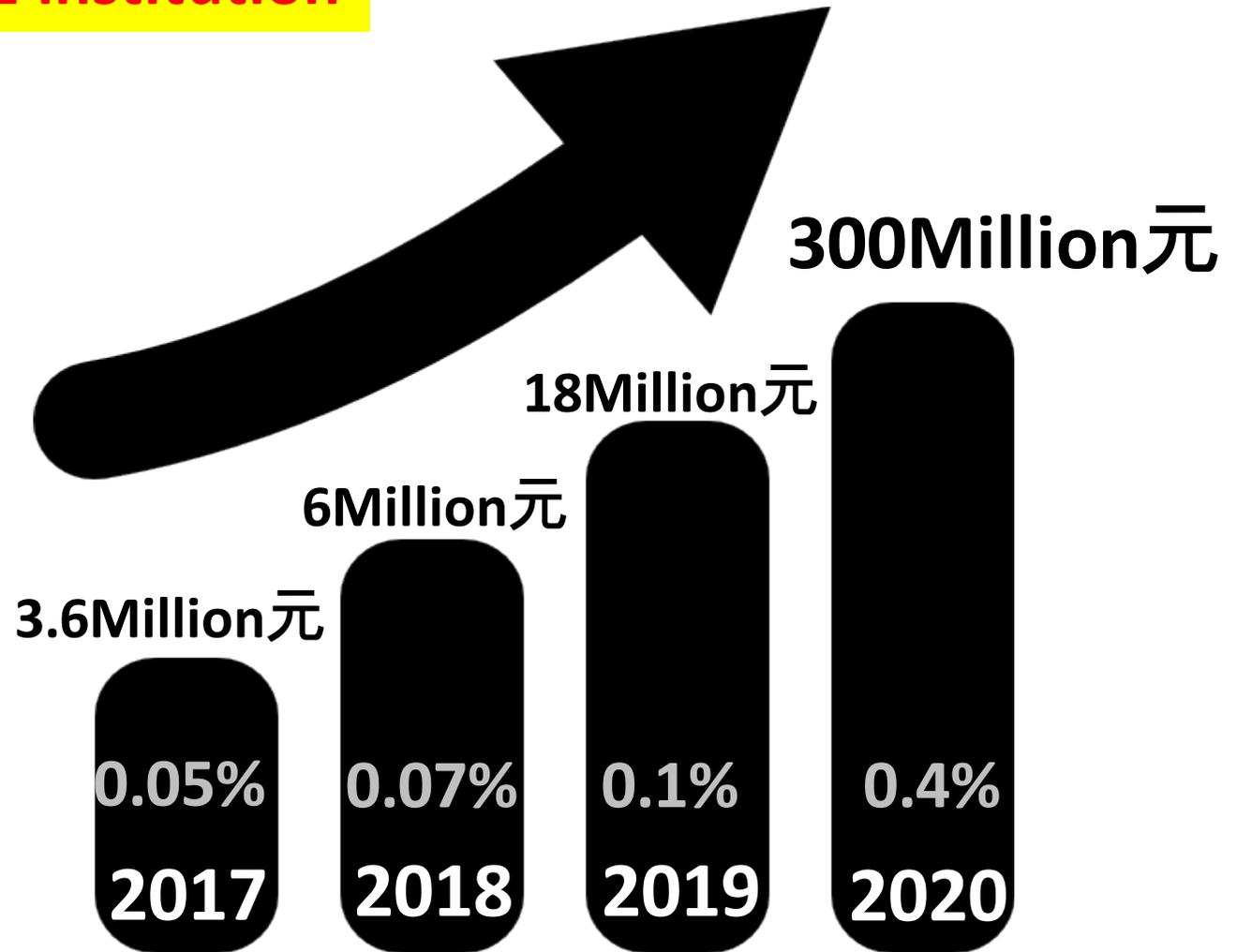
UNwise is an Scandinavian's leading online assessment services provider.

How much will you get money?

Financial Projection

10,000 students = 1 institution

ILLUSTRATIVE



3 Hundred Million CNY income is expected

Let's take a real look at our current situation

ROADMAP



CHINA



OMAP

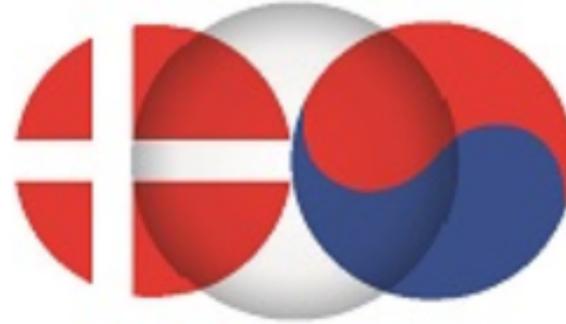
Establish 6 branch network

Welcome Pilot-study

**Welcome Promotion and Sales
partners**

further questions

creative partners



Ducogen
Nordic Passion

Learning by Doing

Park Jung Ho | CEO

DUCOgen/UNIwise Korea

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Implications of Learning Analytics on OMAP

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Ewha Womans University

Implications of Learning Analytics on OMAP

1) Uncovered Treasure: Big Data

- **Large Amount of Data from WISEflow**



- Grading
- Time spent on each item
- Frequency of modifying answer
- ...

Implications of Learning Analytics on OMAP

1) Uncovered Treasure: Big Data

▪ Necessity of Various Methods of Data Analysis



- Predictive modeling
- Regression
- Cluster analysis
- Machine learning
- Network analysis
- Time series analysis
- ...



Implications of Learning Analytics on OMAP

1) Uncovered Treasure: Big Data

- **Clustering of Online Students: Towards an Elaborated Prediction Model of Learning Achievement** (Lee, H., Sung, H., Park, Y. & Jo, I., 2015)

✓ Research Variables

Variables	Explanation
TLT (Total Login Time)	A time period from the beginning of login to the point of time the learner finish it
TLF (Total Login Frequency)	Adding up the total numbers of login
LIR (Login Regularity)	Standard deviation of average login interval
VOB (Visit on Board)	Adding up the total numbers of access to board
TSB (Time Spent on Board)	A time period from the beginning of the using board to the point of time the learner finish it
LIRB (Login Regularity on Board)	Standard deviation of average access on board interval
TM (Time on Movie)	A time period from the beginning of the playing movie to the point of time the learner finish it
FM (Frequency on Movie)	Adding up the total numbers of playing movie
RM (Regularity on Movie)	Standard deviation of average playing movie interval
TS (Total Score)	Adding up each evaluation score

Implications of Learning Analytics on OMAP

1) Uncovered Treasure: Big Data

- **Clustering of Online Students: Towards an Elaborated Prediction Model of Learning Achievement** (Lee, H., Sung, H., Park, Y. & Jo, I., 2015)

How learners' online behavior clusters certain groups with different characteristics?

✓ Cluster Analysis (K-means algorithm)

1. Standardize dataset → **scale difference**
2. Set the number of $k = 3$
3. Find clusters which **minimize** the within-cluster sum of squares based on **Euclidean distance**

Implications of Learning Analytics on OMAP

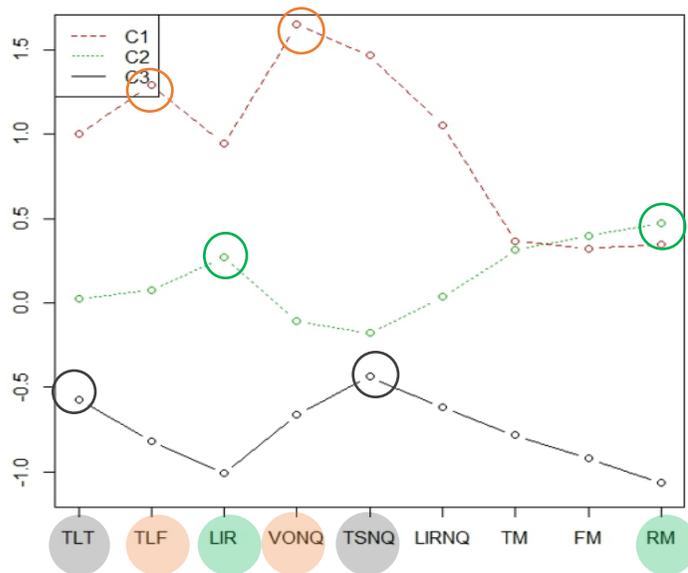
1) Uncovered Treasure: Big Data

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✓ Cluster Analysis (K-means algorithm)

Cluster mean vectors of K-means algorithm (N = 194)



	Number of Students	High value
C1	30	 Busy sparrow
C2	107	 Steady ant
C3	57	 Inactive sloth

Implications of Learning Analytics on OMAP

1) Uncovered Treasure: Big Data

- **Clustering of Online Students: Towards an Elaborated Prediction Model of Learning Achievement** (Lee, H., Sung, H., Park, Y. & Jo, I., 2015)

✓ Conclusion

- **First**, each cluster showed different participation degree
- **Second**, predictors for learning achievement were different by cluster-based prediction models
- **Third**, the explanation ratio increased gradually in clustered prediction model than single model

[Whole group regression]

(N = 194)	
	t
(constant)	10.358
LIR	-4.294*
TSB	2.552*
TLT	-2.536*
LIRB	3.240*
VOB	2.681*
$R^2(\text{adj. } R^2) = .222 (.201)$	

x3

[Clustered regression]

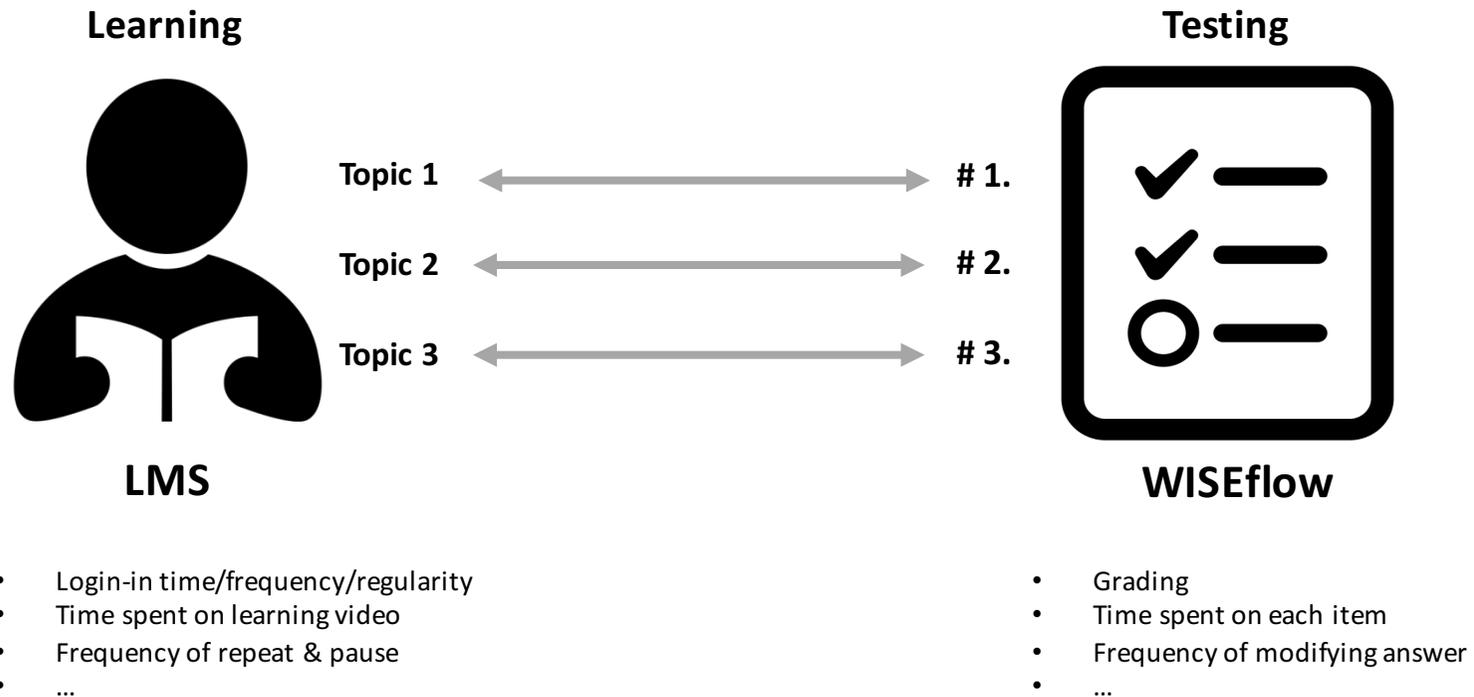
(n = 30)	
	t
(constant)	4.692
TLT	-6.788*
TM	3.510*
TSB	3.298*
$R^2(\text{adj. } R^2) = .650 (.610)$	
(n = 107)	
	t
(constant)	4.908
RM	2.105*
$R^2(\text{adj. } R^2) = .040 (.031)$	
(n = 57)	
	t
(constant)	.349
TLF	3.230*
TSB	3.014*
TM	2.257*
$R^2(\text{adj. } R^2) = .436 (.404)$	

x2

Implications of Learning Analytics on OMAP

2) Combining learning data with WISEflow

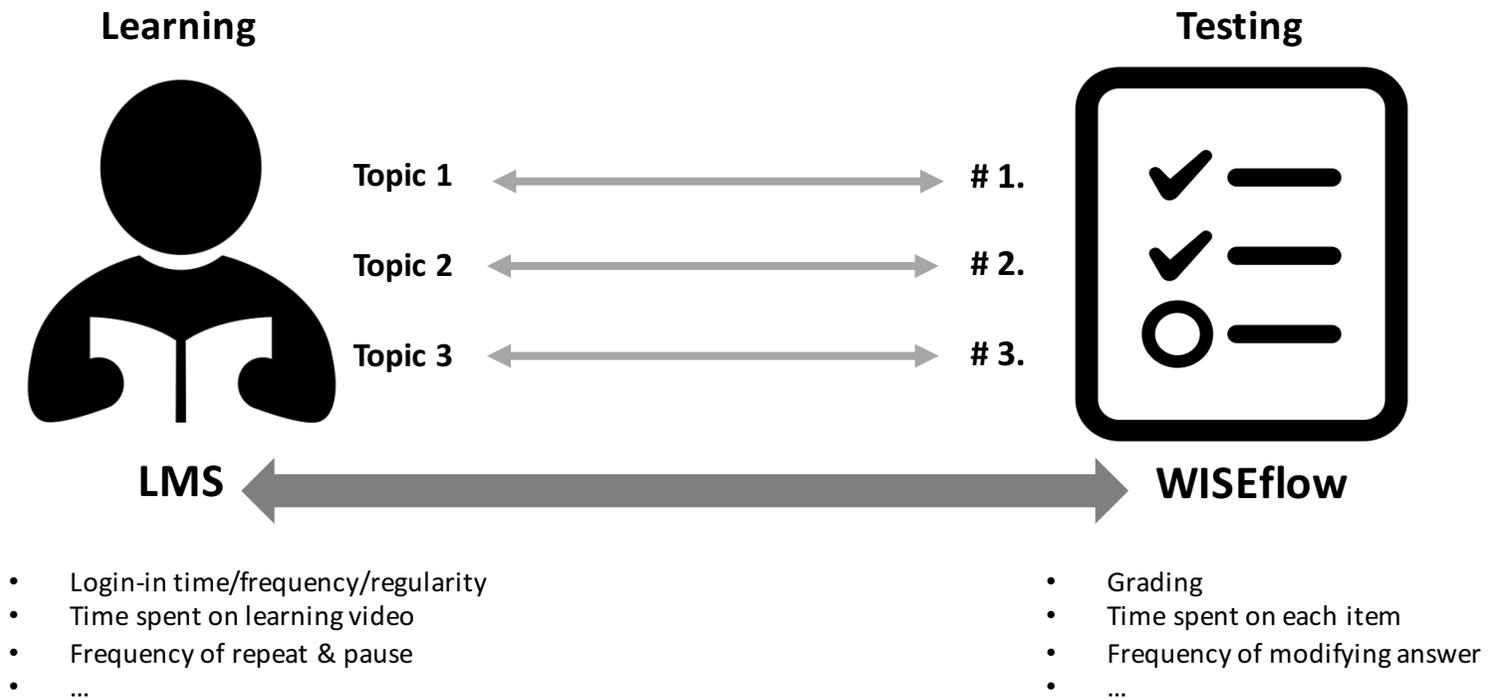
- Matching Learning data to Each Test item



Implications of Learning Analytics on OMAP

2) Combining learning data with WISEflow

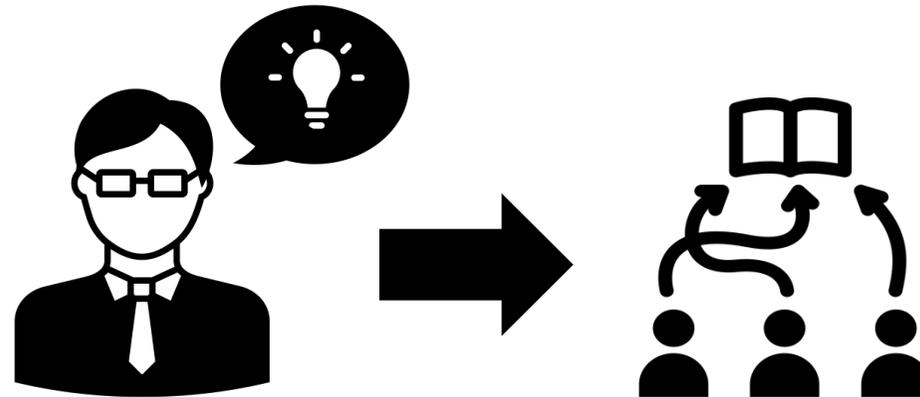
- Matching Learning data to Each Test item



Implications of Learning Analytics on OMAP

2) Combining learning data with WISEflow

- **Providing Insights for Teaching and Exam methods**

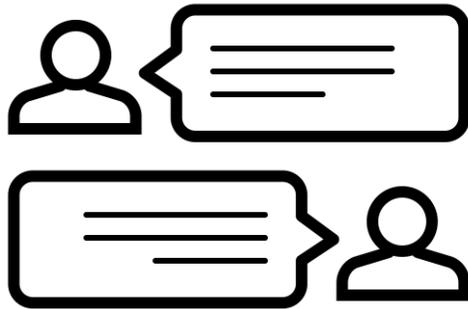


- Analyzing the connection between teaching methods and learning outcomes.
- Institutions can gain insights into which teaching and exam methods the students perform best in.

Implications of Learning Analytics on OMAP

3) Using Data for Action and for Impact

- **Automated Message feedback**

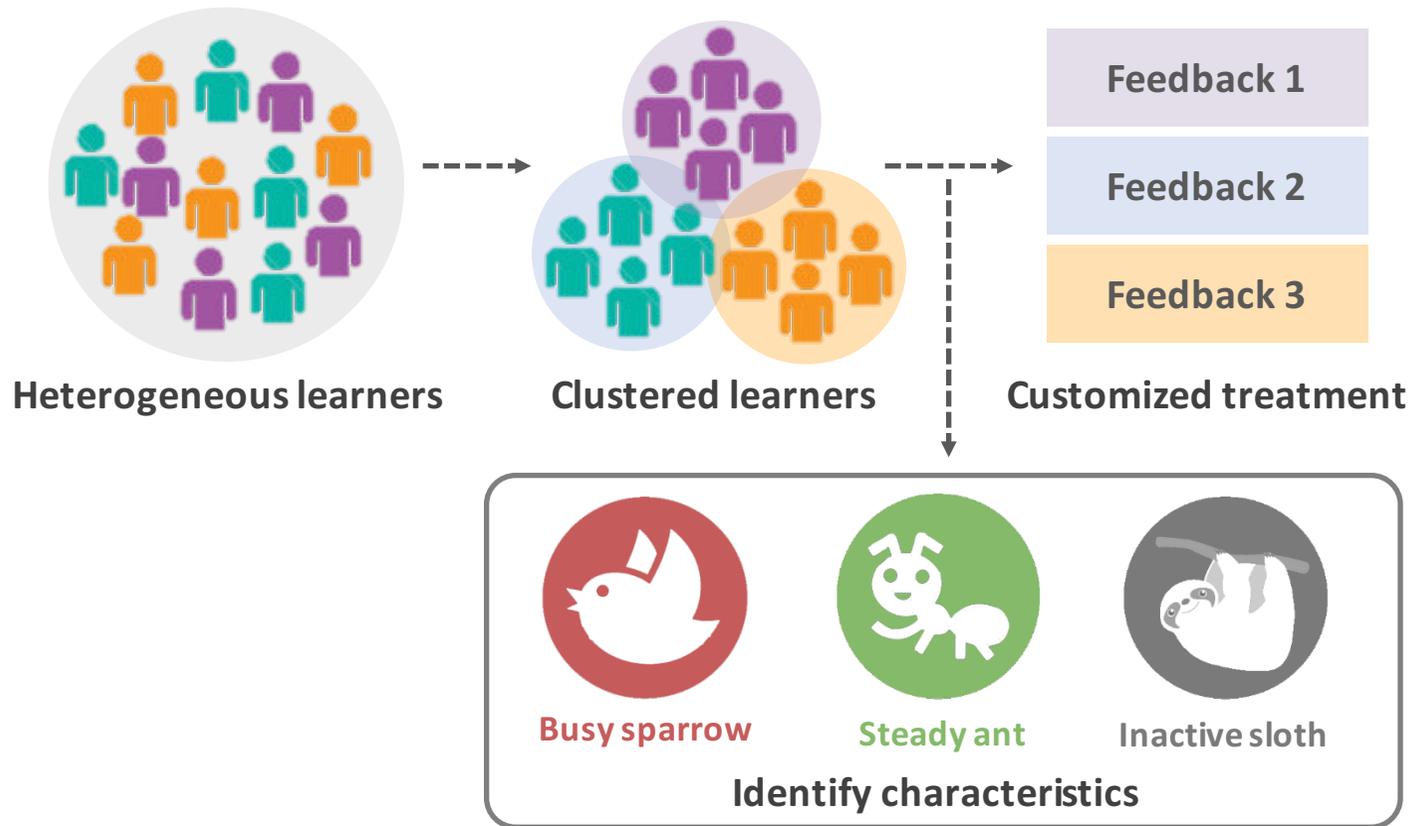


- Highlighting 'feedback' function in WISEflow.
- Customize the messages and control the flow of info.

Implications of Learning Analytics on OMAP

3) Using Data for Action and for Impact

- Automated Message feedback

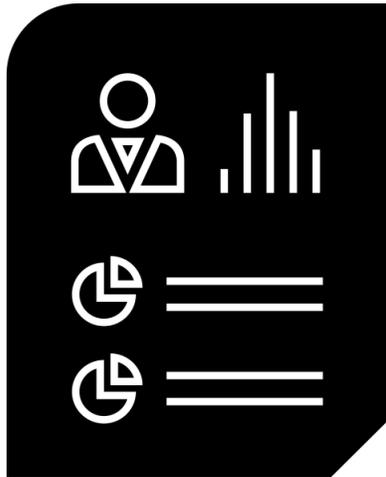


[Facilitate active participation + enhance learning performance]

Implications of Learning Analytics on OMAP

3) Using Data for Action and for Impact

▪ Learning Analytics Dashboard



- A visualized and intuitive display derived from the results of educational data-mining for the purpose of supporting students' learning and performance improvement
(Yoo, Lee, & Park, 2014)
- LAD includes visual elements such as charts, graphs, indicators and alert mechanisms (Podgorelec & Kuhar, 2011).

Q & A

Thank you !