

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



# OMAP

Online Massive Assessment Platform

**For stress-less school**



*Leo Park*  
*C.E.O*

**DUCOgen**

[ceo@ducowise.com](mailto:ceo@ducowise.com) [tose123@gmail.com](mailto:tose123@gmail.com)

82-10-4630-4376

Version 1.6

e-learning expo 2017 LASI ASIA

# **Index**

- 1. Definition**
- 2. Problem & Value**
- 3. Target**
- 4. Business model**
- 5. GO-TO Market plan**
- 6. SWOT analysis**
- 7. Who we are**
- 8. Financial projection**
- 9. Roadmap**

**What is the OMAP we propose?**

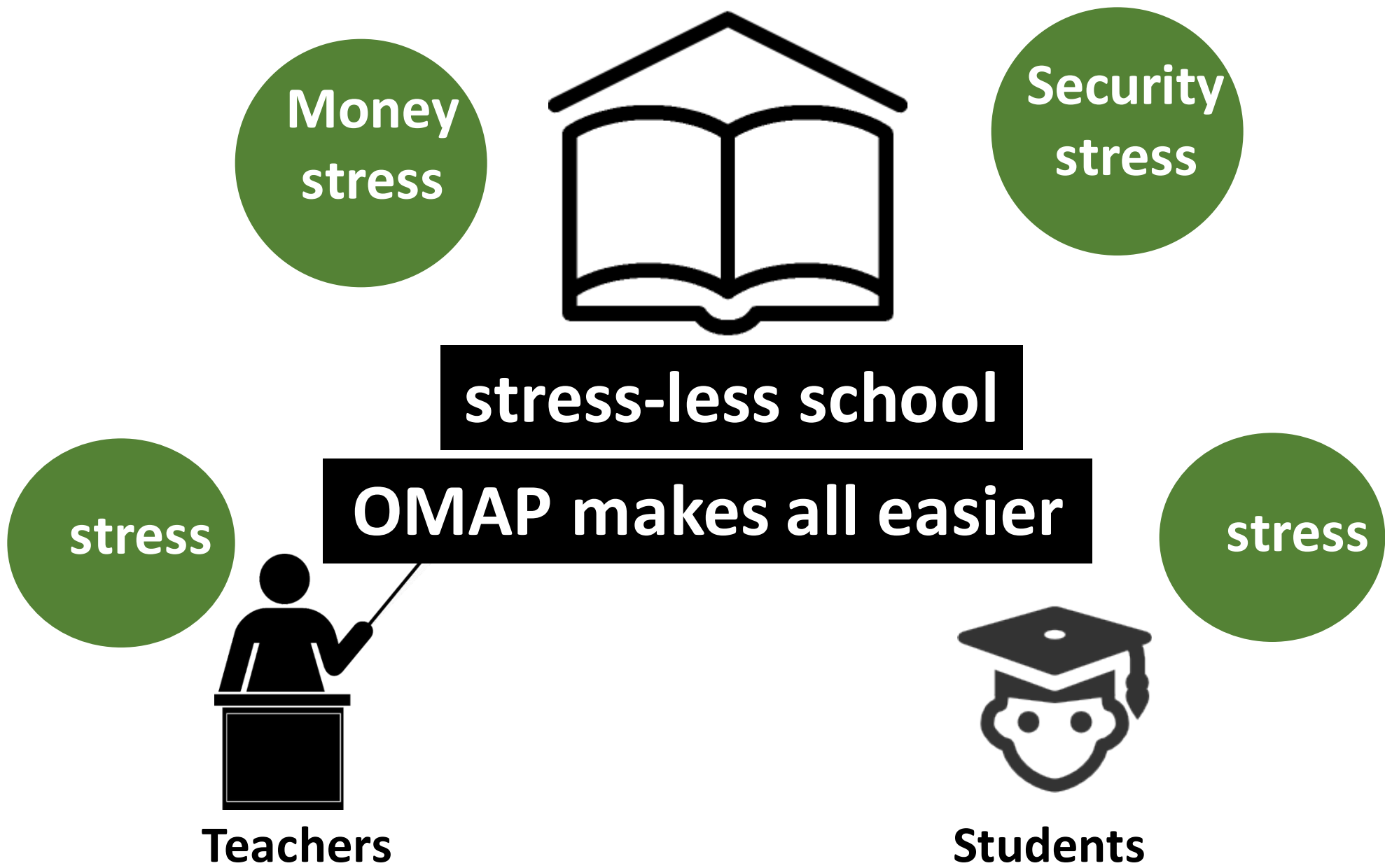
# Definition



**OMAP covers all the tests.**

**What does stress-less school mean?**

**Problem & Value**



**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**

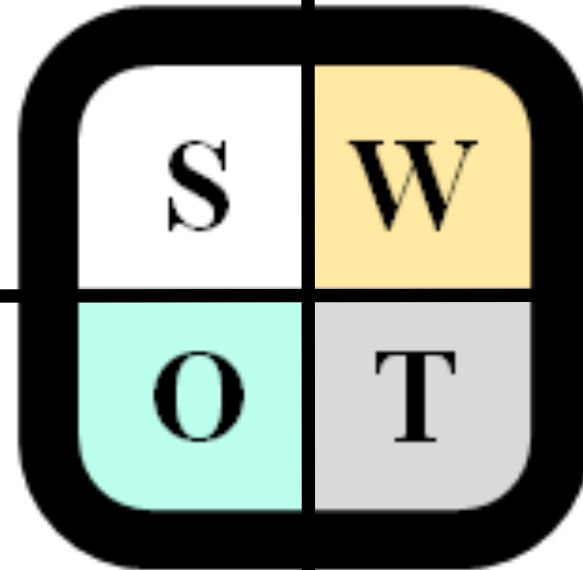


**Search Promotion & Sales  
partner**

**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

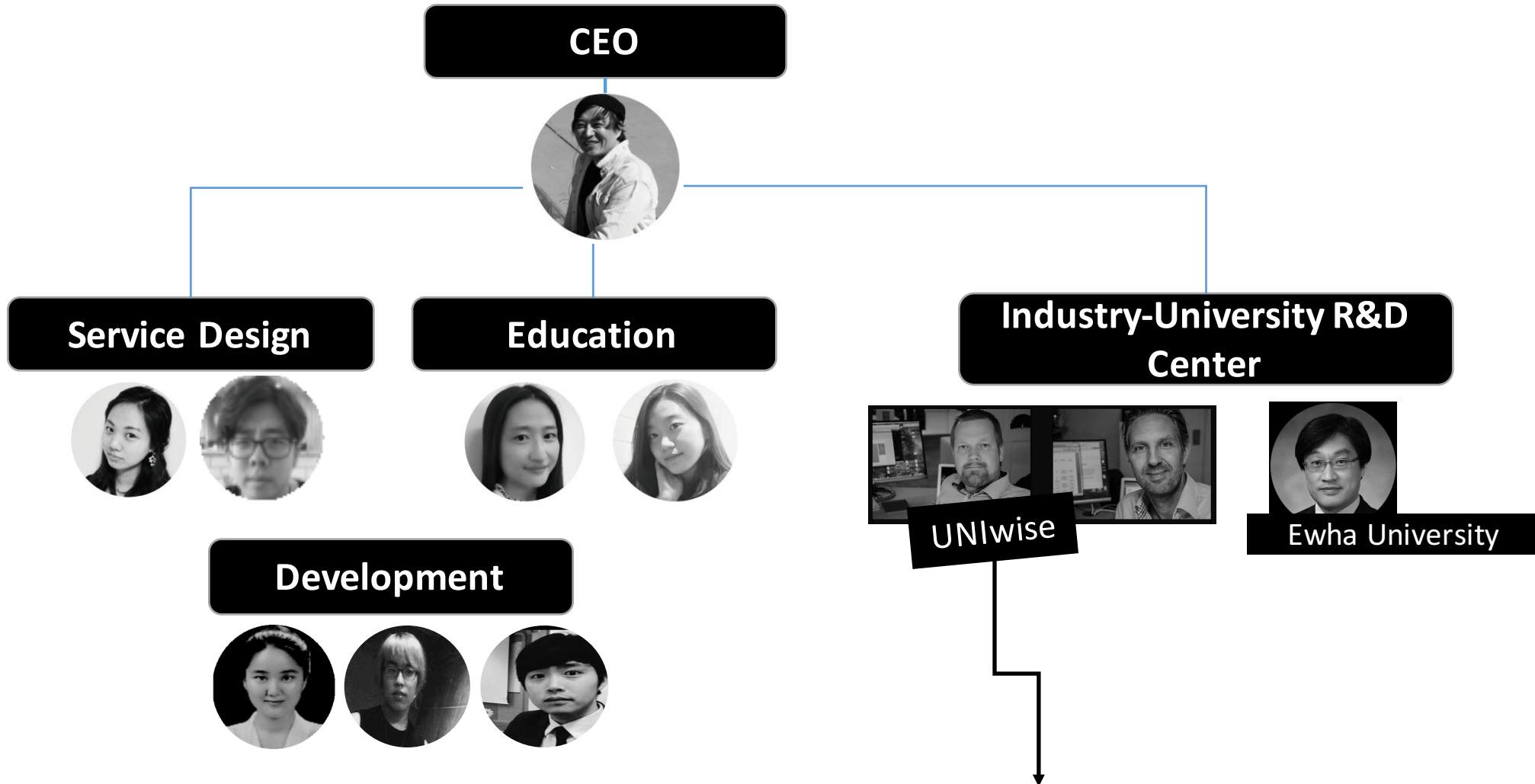
Teachers hate  
technologies

Schools want  
cost-saving

independence

**Who are we, DUCOgen?**

# WHO WE ARE



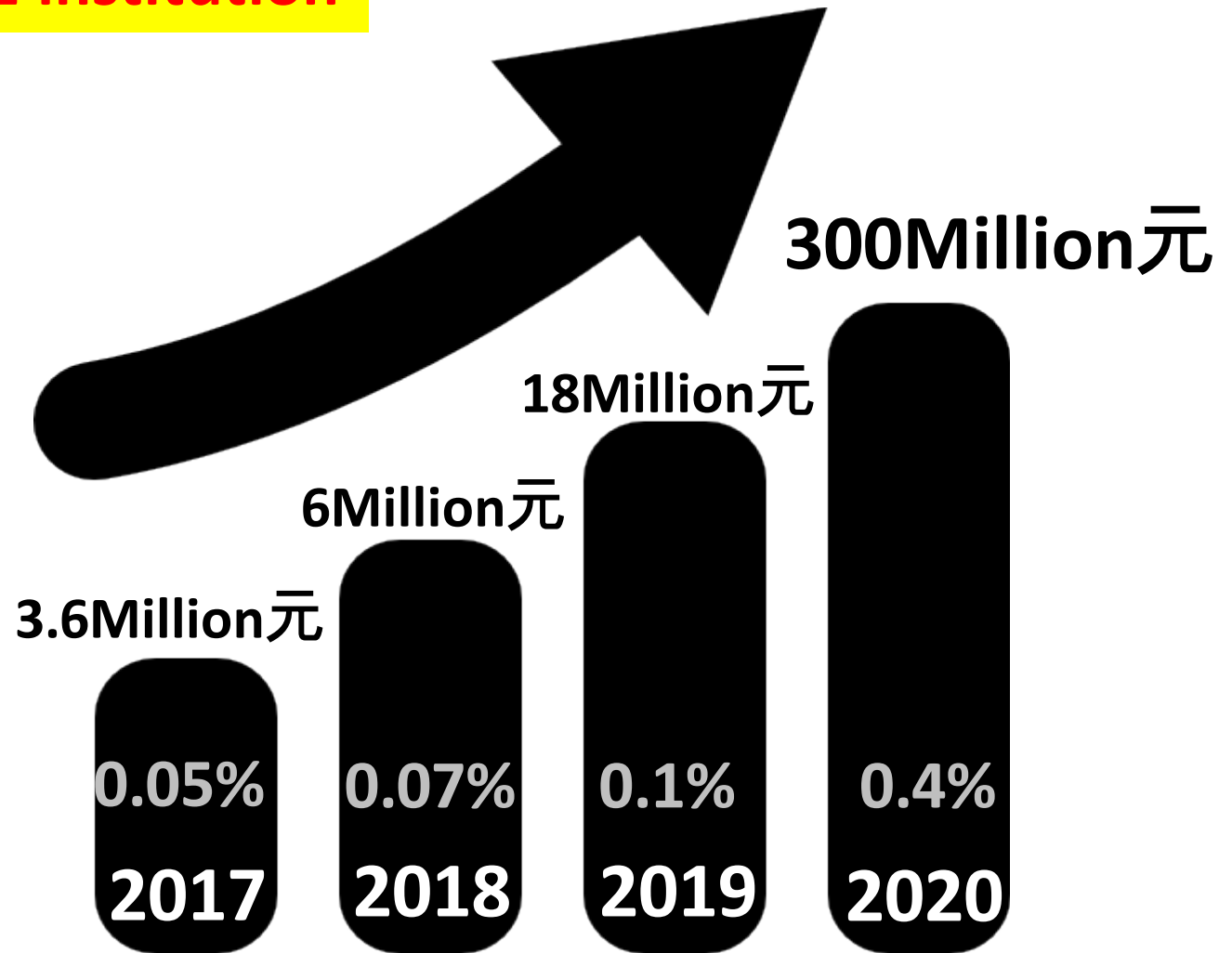
*UNwise* is an Scandinavian's leading online assessment services provider.

**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**

**C.P\_**+82-10-4630-4376 / [www.facebook.com/ducowise](https://www.facebook.com/ducowise)

**E-mail\_** [tose123@gmail.com](mailto:tose123@gmail.com) / [ceo@ducowise.com](mailto:ceo@ducowise.com)

**Office\_**

[Main]. 6F. DMC Nurikkumsquare Buld /No.1605

Sangam-dong, Mapo-gu, Seoul

[R&D]. 3F Global Lounge. DMC R&D Center / No.1649

Sangam-dong, Mapo-gu, Seoul



# Implications of Learning Analytics on OMAP

Eunji Lee

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
- ...

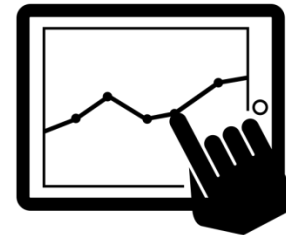
# Implications of Learning Analytics on OMAP

## 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

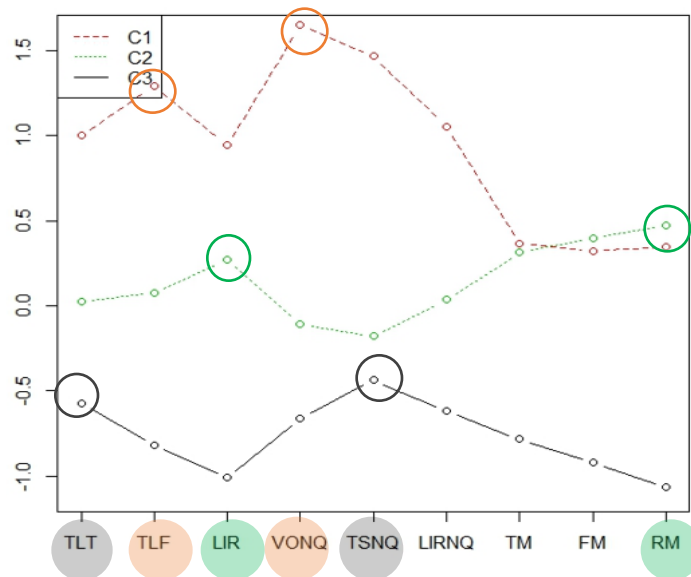
## 1) Uncovered Treasure: Big Data




- **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)

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*
$R^2(\text{adj. } R^2) = .222 (.201)$	

[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*
$R^2(\text{adj. } R^2) = .436 (.404)$	

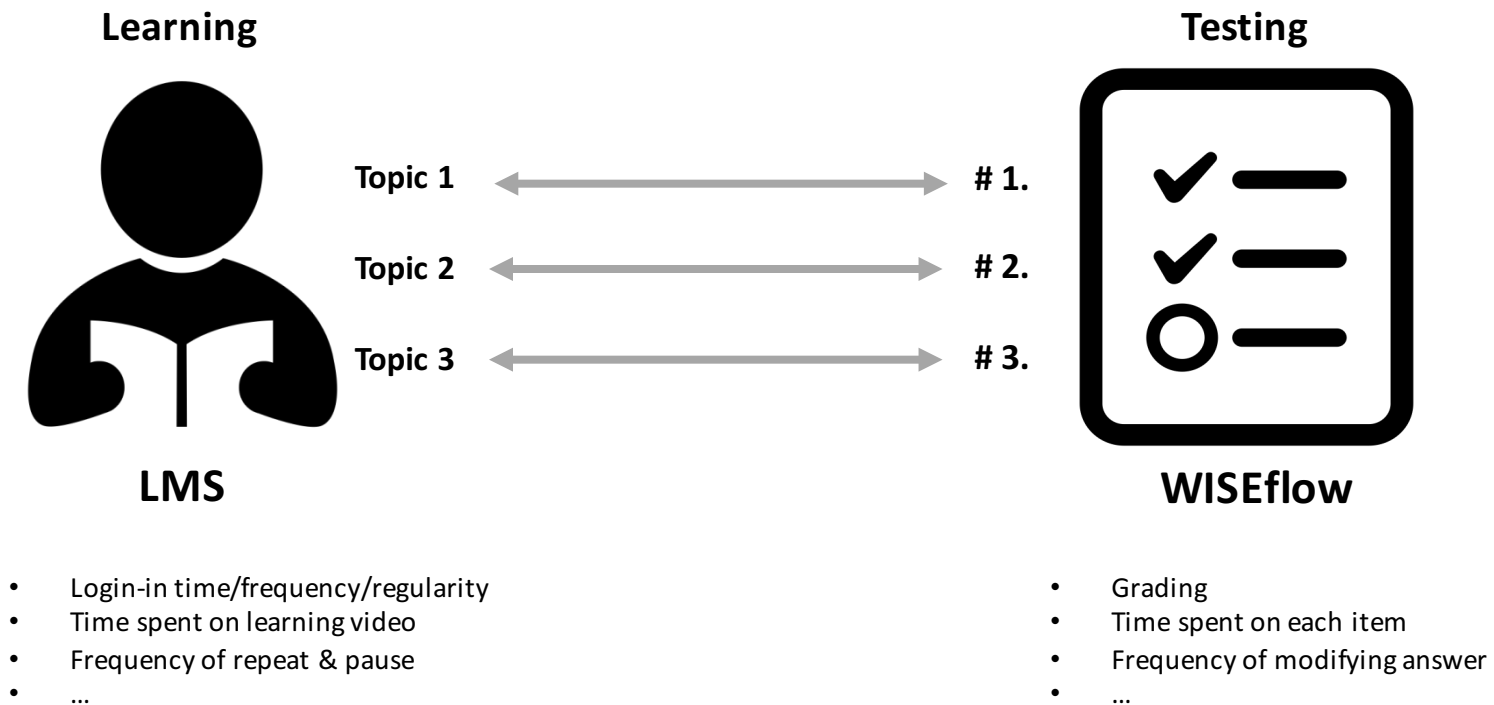
x3

x2

# Implications of Learning Analytics on OMAP

## 2) Combining learning data with WISEflow

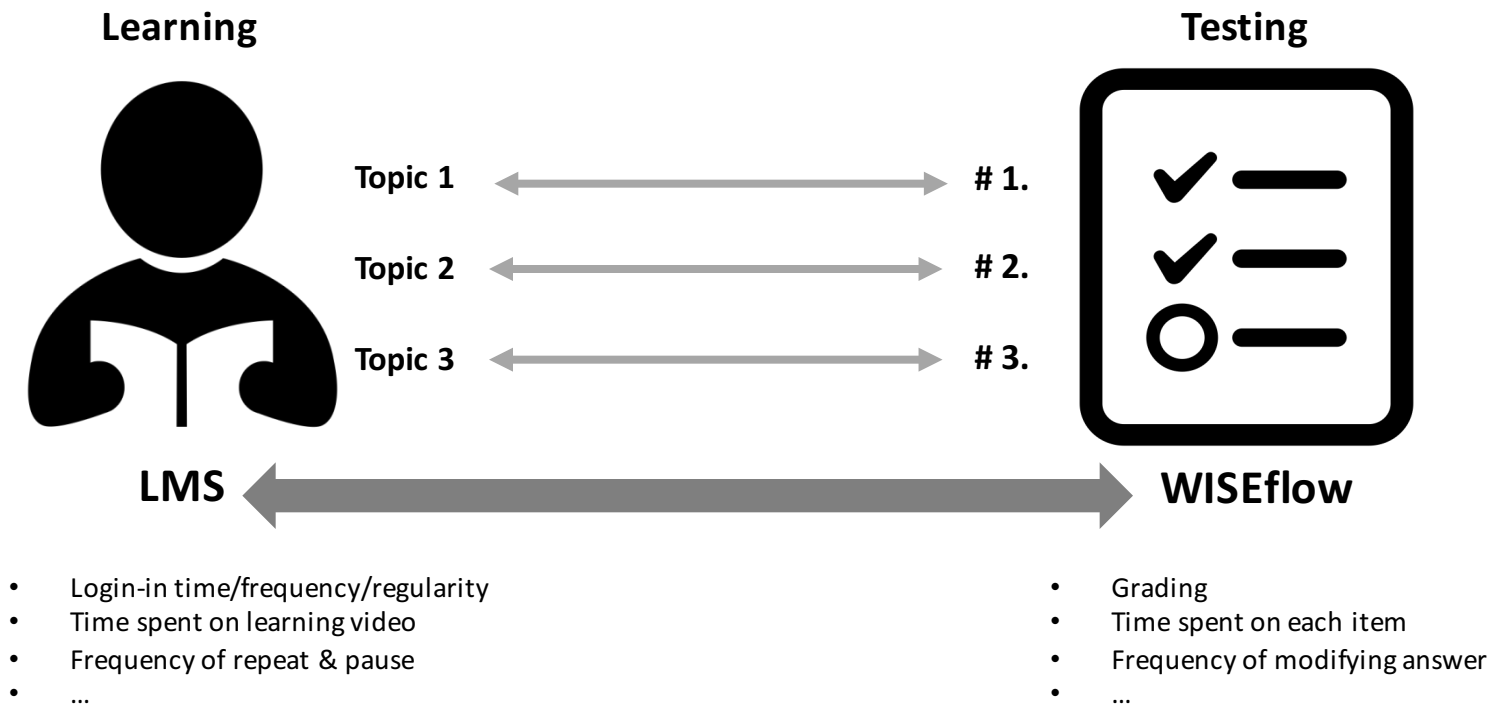
### ▪ Matching Learning data to Each Test item



# Implications of Learning Analytics on OMAP

## 2) Combining learning data with WISEflow

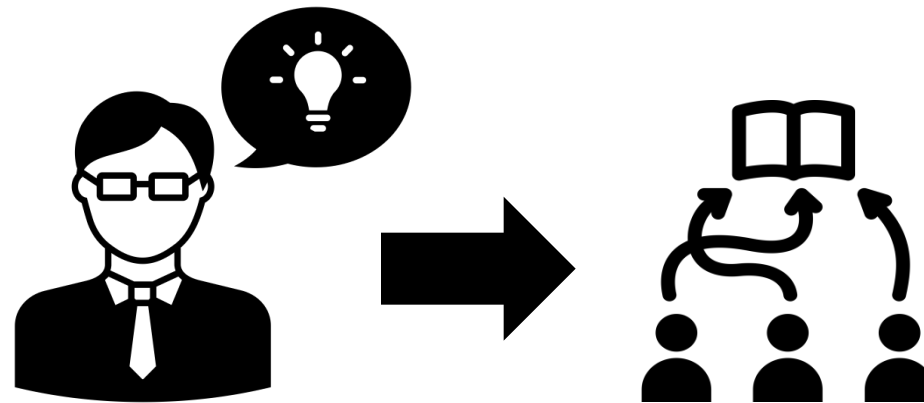
### ▪ Matching Learning data to Each Test item



# Implications of Learning Analytics on OMAP

## 2) Combining learning data with WISEflow

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

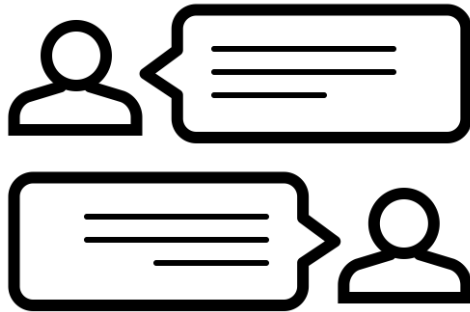


- Analyzing the connection between teaching methods and learning outcomes.
- Institutions can gain insights into which teaching and exam methods the students perform best in.

# Implications of Learning Analytics on OMAP

## 3) Using Data for Action and for Impact

### ▪ Automated Message feedback

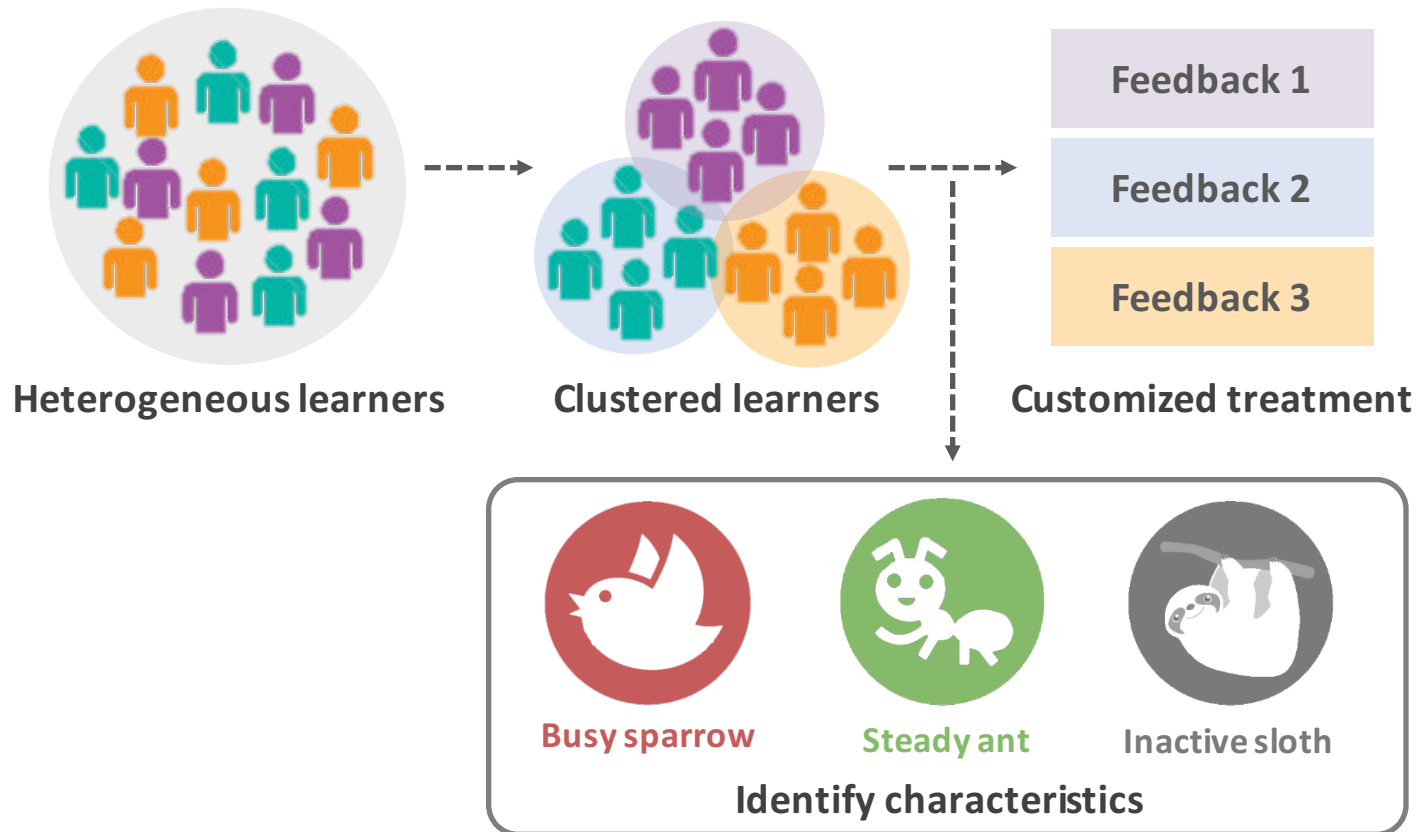


- Highlighting 'feedback' function in WISEflow.
- Customize the messages and control the flow of info.

# Implications of Learning Analytics on OMAP

## 3) Using Data for Action and for Impact

- Automated Message feedback

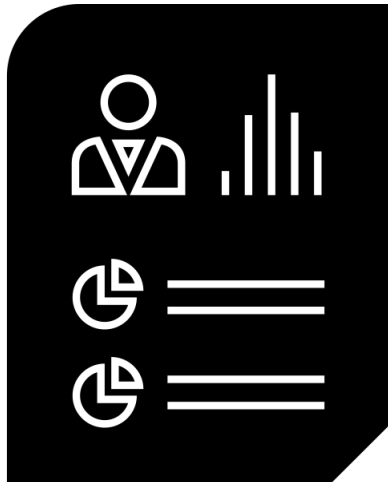


[ Facilitate active participation + enhance learning performance ]

# 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 !