# Improving an online diagnostic test via item analysis 

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Diagnostic tests are widely used in university mathematics departments. Here, we demonstrate a set of methods for evaluating the performance of such a test, based on the example of improving the test used by the University of Edinburgh.

## Aims of the test

- Inform staff about the preparedness of incoming undergraduates.
- Show students the sort of skills we expect them to have, and give them a chance to self-diagnose topics requiring further study.
- Inform decisions about course choice, e.g. direct entry to Year 2.


## Project timeline

2011 Original 32 question test created using Maple T.A.
2012 Test is reduced to current form of 20 questions
2013 Project evaluates performance of the new test

2016 There are now $\approx 1000$ test attempts each year
2017 With the test moving to the STACK online assessment system, the current project re-evaluates its performance and implements changes


## Test format

- 20 questions on core high-school mathematics
- 90 minute time limit, one attempt allowed (a practice quiz is available)
- Questions are a mixture of numerical response and multiple choice
- All questions make use of randomisation
- Feedback is limited to marks and whether correct (no worked solutions)


## Example questions

Q3



Q6

## 

 Find hin a vues of $A$docima p pases

Q13


$\qquad$

Q14 $\qquad$

Q20

## Evaluation methods

## 1. Content

The MATH taxonomy (Smith et al., 1996) was applied to judge the type of mathematical skill assessed by each question. The results show the test is predominantly based on "routine procedures", which is perhaps not surprising for a test of this type.
To make the test more balanced, three Group A questions were removed, and replaced by two Group B and one Group C question.

|  | Old | New | New Group C question |
| :---: | :---: | :---: | :---: |
| Group A <br> Routine Procedures | 14 | 11 |  |
| Group B <br> Using knowledge in new ways | 6 | 8 | $\min (\max (10,8), 12)=$ $\qquad$ $\max (\min (a, b), c)>\min (\max (a, b), c)$. |
| Group C Mathematical arguments | 0 | 1 | $b=\square$ $c=\square$ |

## 3. Psychometric analysis

Assuming the test is measuring a single trait, e.g. mathematical ability, item response theory uses the response data to estimate the trait level for each student
For each question, the expected score for any given ability is modelled by a logistic (S-shaped) curve, with two parameters:

- difficulty - horizontal shift (left=easier)
- discrimination - the maximum slope


The three items selected for replacement were among the least discriminating items. The information curve for each question shows where it is most discriminating on the ability scale. Their sum, the test information function, shows the test is most informative at the lower end of the ability scale.

## References

Revelle, W. (2018). How To: Use the psych package for Factor Analysis and data reduction. [online] Available at Revele, W. (2018). How To. Use the pssch hackage for Factor Analysis and dal
http://personality-project.orgg//psych
Smith, G. et al. (1996) 'Constructing mathematical examinations to assess a range of knowledge and skills'. Internatio Journal Iof Mathematical Education in Science and lechnology. Taylor \& Francis Group, 27(1), pp. 65-77.

## 2. Factor analysis

Factor analysis aims to explain observed scores in terms of underlying factors or latent traits, e.g. mathematical ability.
Results from the R package psych (Revelle, 2018) showed:

- good loading on a general factor, g, indicating that all of the questions are assessing a common construct
- three subsidiary group factors:

F1 Predominantly questions from MATH Group B
F2 Chain rule, perhaps calculus more generally

F3 Questions requiring the use of a calculator


## 4. Predictive validity

Diagnostic Test scores are used in practice to help inform course choice, so it is important to understand their relationship with subsequent performance. The scores correlated moderately with the first course results in three different tracks:


Interestingly, slightly higher correlations were found using the sub-score on the MATH Group B questions, e.g. $r=0.53$ for Year 1 Mathematics. This contributed to the decision to rebalance the test in favour of Group B.

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