

# Tackling Linguistic Difficulty through Serious Games

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iREAD  
personalised reading



JALTECALL  
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# Digital Game-Based Language Learning

Games can be great for language learning:

- 🎮 Inherently motivational
- 🎮 Input rich
- 🎮 Personalised support

A	R	I	S	E
R	O	U	T	E
R	U	L	E	S
R	E	B	U	S

Cornillie, F., Thorne, S. L., & Desmet, P. (2012). ReCALL special issue: Digital games for language learning: challenges and opportunities: Editorial Digital games for language learning: From hype to insight?. ReCALL, 24(3), 243-256.

c.f. Young et al. (2012), Yudintseva (2015), Poole & Clarke-Midura (2020) for research syntheses on DGBL and Language Learning

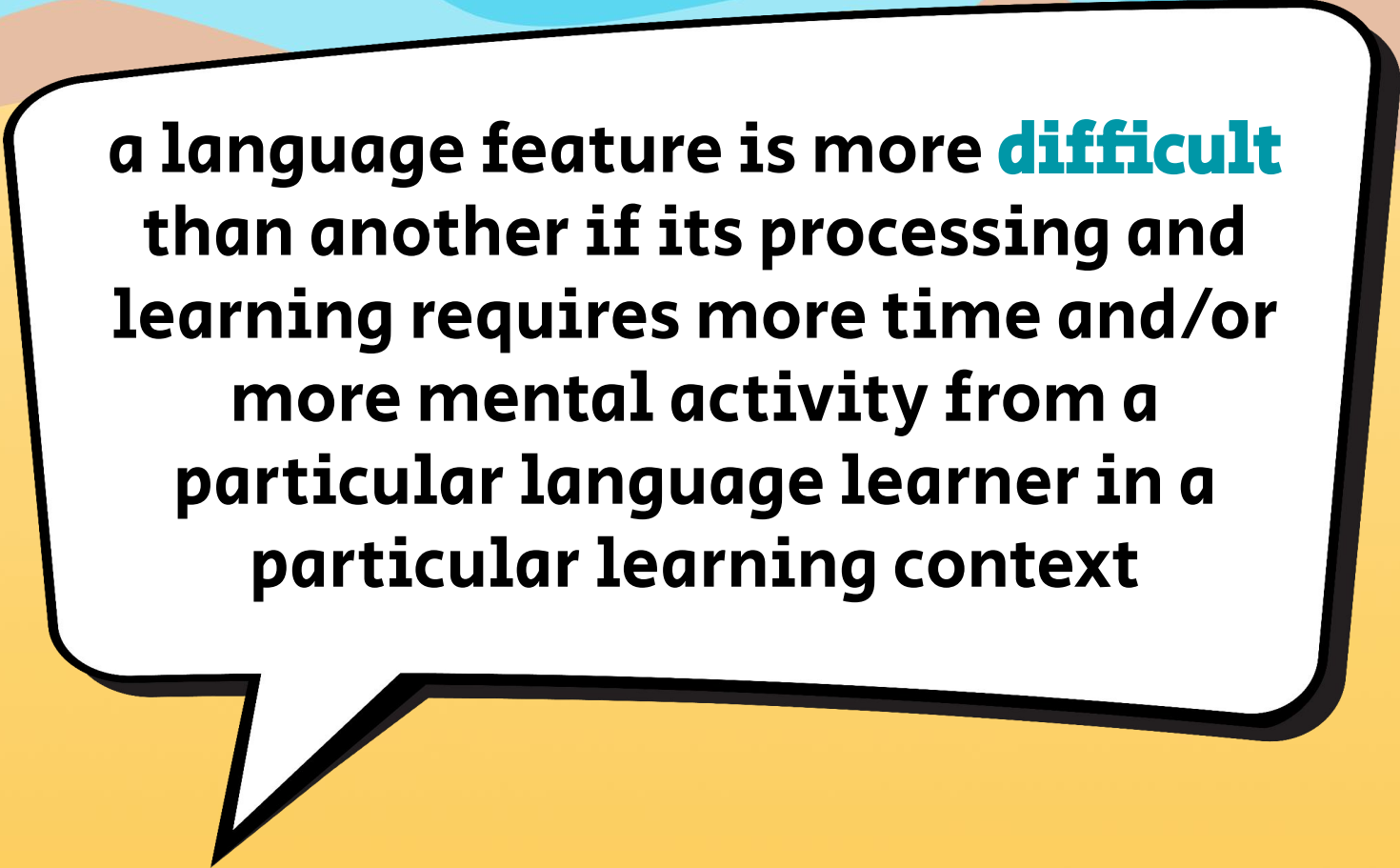
The background features a stylized landscape with rolling yellow hills and a light blue sky. The hills are represented by smooth, undulating shapes, and the sky is a solid light blue. The overall aesthetic is clean and modern.

# **Linguistic Difficulty**

# Linguistic Difficulty

- Language is undoubtedly difficult to acquire
  - and some language features are more difficult than others





**a language feature is more **difficult** than another if its processing and learning requires more time and/or more mental activity from a particular language learner in a particular learning context**





**Housen & Simoens, 2016**

# Linguistic Complexity





- Difficulty is a different construct than complexity
    - (but sometimes the words get interchanged)
  - Complexity has to do with the number of elements in a language system / text, and their relational patterns
- (Pallotti, 2015)
- Sometimes linguistic structures that are structurally more complex may actually be easier to learn and produce

# Linguistic Difficulty

## Subjective, Learner-Related Difficulty

-  individual differences in cognitive abilities
-  knowledge of previously learned languages
-  overall L2 proficiency and stage of L2 development
-  socioaffective and personality factors (e.g., motivation, extraversion, and anxiety)

## Objective, Feature-Related Difficulty

-  elaborateness (the number of concepts, steps, subrules, or criteria they contain)
-  conceptual clarity (amount and degree of technicality of metalanguage used)
-  scope (with general rules vs. specific rules)
-  reliability and truth value (number of exceptions) of the rules

# Linguistic Difficulty



## Subjective, Learner-Related Difficulty



individual differences in cognitive abilities



knowledge of previously learned languages



overall L2 proficiency and stages of L2 development



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## Objective, Feature-Related Difficulty



elaborateness (the number of concepts, steps, subrules, or criteria they contain)



formal clarity (amount and degree of technicality of metalanguage used)



scope (with general rules vs. specific rules)



reliability and truth value (number of exceptions) of the rules



# Why does difficulty matter?



## Research



do particular instructional approaches work better for easier/harder language features?



control for difficulty when investigating other topics



## Teaching



should teachers focus on simple or difficult features?

# How do we measure linguistic difficulty?



Until now some of the best indicators of feature difficulty are the judgments of experts (e.g. teachers) and perceptions of students. (Housen, 2014).



Some few studies that use reaction time measures or brain or eye-tracking methods to approximate cognitive difficulty



(e.g. Godfroid, 2016, Morgan-Short et al., 2010, Godfroid & Uggén, 2013)

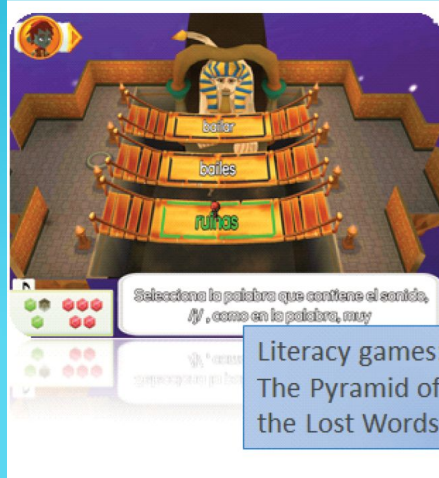


but they necessarily have to focus on a comparison of a small number of features

The background features a stylized landscape with rolling yellow hills and a light blue sky. The hills are represented by smooth, undulating shapes, while the sky is a solid light blue. The overall aesthetic is clean and modern.

# **Context: The iRead Project**

# iRead innovation project



Literacy games:  
The Pyramid of  
the Lost Words



e-Reader



Teacher  
Analytics  
Tool

- English
- English for dyslexia
- EFL
- German
- Greek
- Spanish

Thousands of children  
across Europe.



A 4-year EC Funded project (Jan 2017-Dec 2020). This project has received funding from the European Union's Horizon 2020 innovation programme under grant agreement No 731724.









# Challenges in interpreting game data

When a child makes a mistake, how do we disentangle whether the mistake is due to low proficiency, the game being challenging, or the feature itself being challenging?





**One of the big advantages of digital games is...**

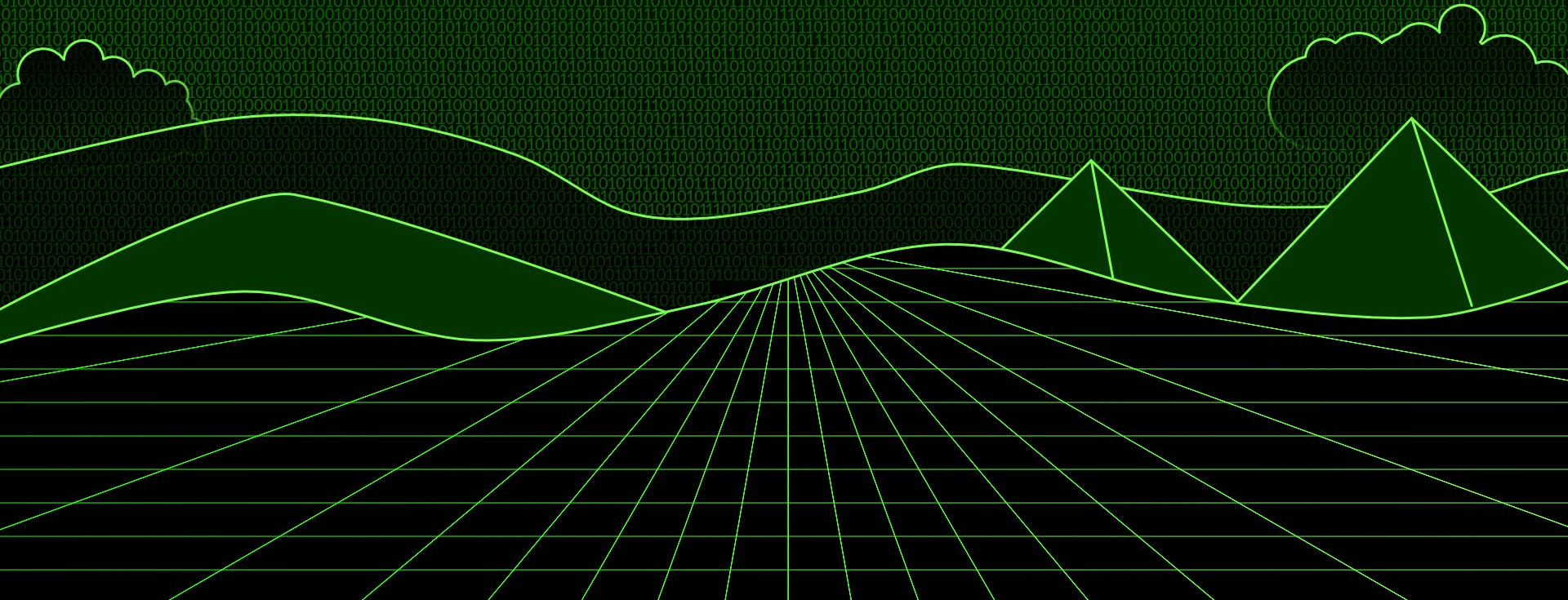


# DATA

From the games, we have tens of thousands of observations of hundreds of Spanish children playing minigames based on more than 200 features.



# Item Response Theory

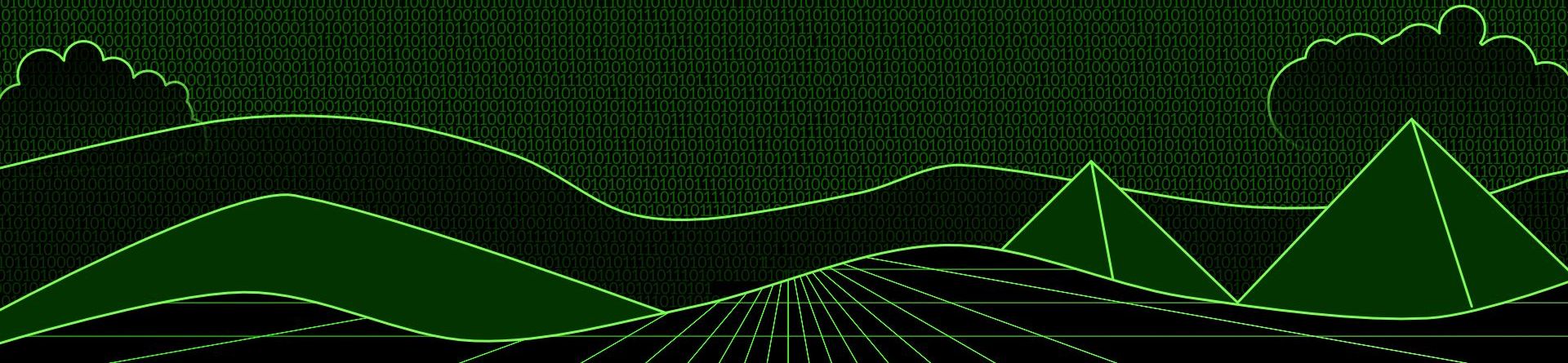




# Classical test theory



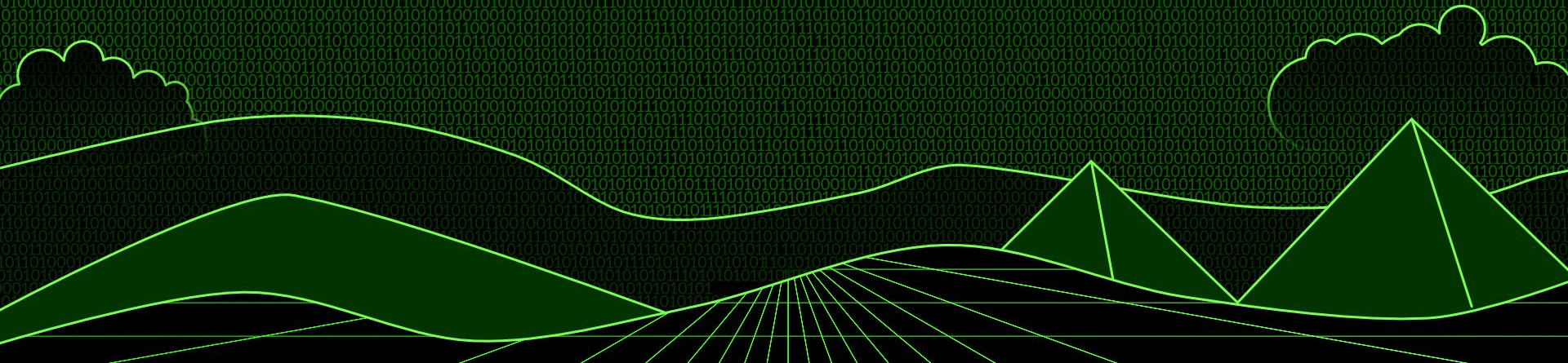
Measured ability = true ability + error





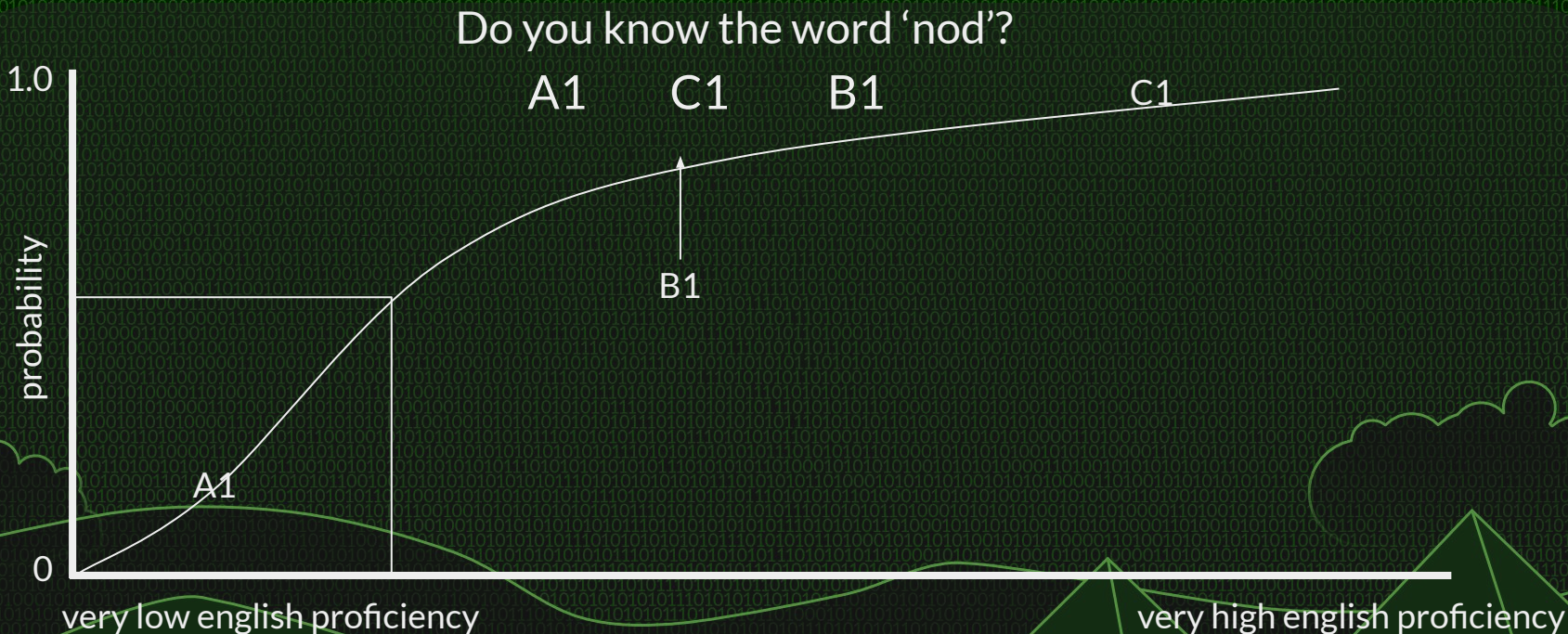
# Item response theory

An IRT model estimates the likelihood of different responses to items by people with different levels of the trait being measured.





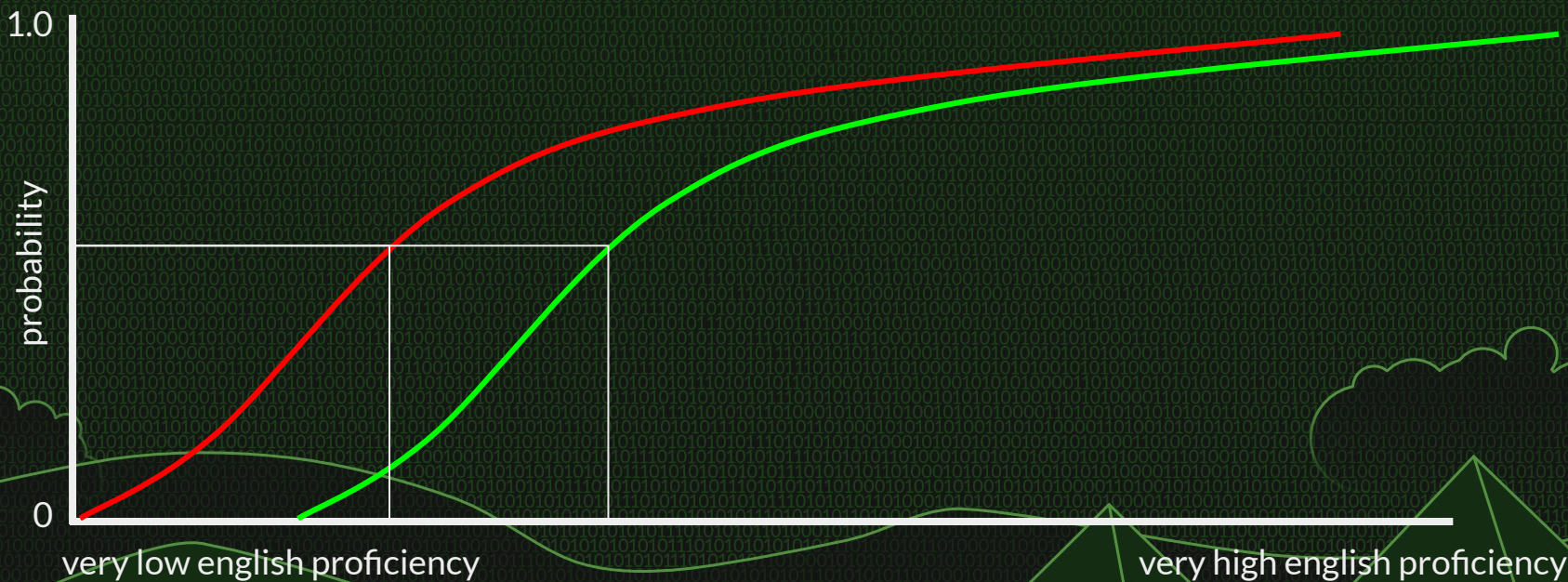
# Item response theory



based on Karon F. Cook's youtube series:  
A conceptual introduction to Item Response Theory



# Item response theory - item difficulty

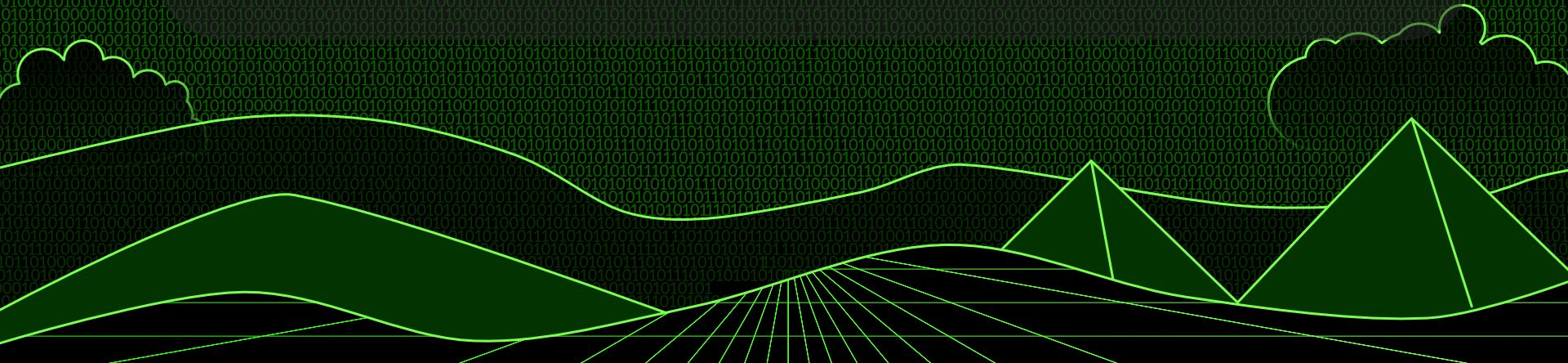


based on Karon F. Cook's youtube series:  
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# Advantages of IRT

- IRT models can work with missing data
- IRT models can make data comparable between different instruments
- IRT models can give you information about individual questions
  - e.g. how well they discriminate and at what levels they discriminate best





# Item response theory

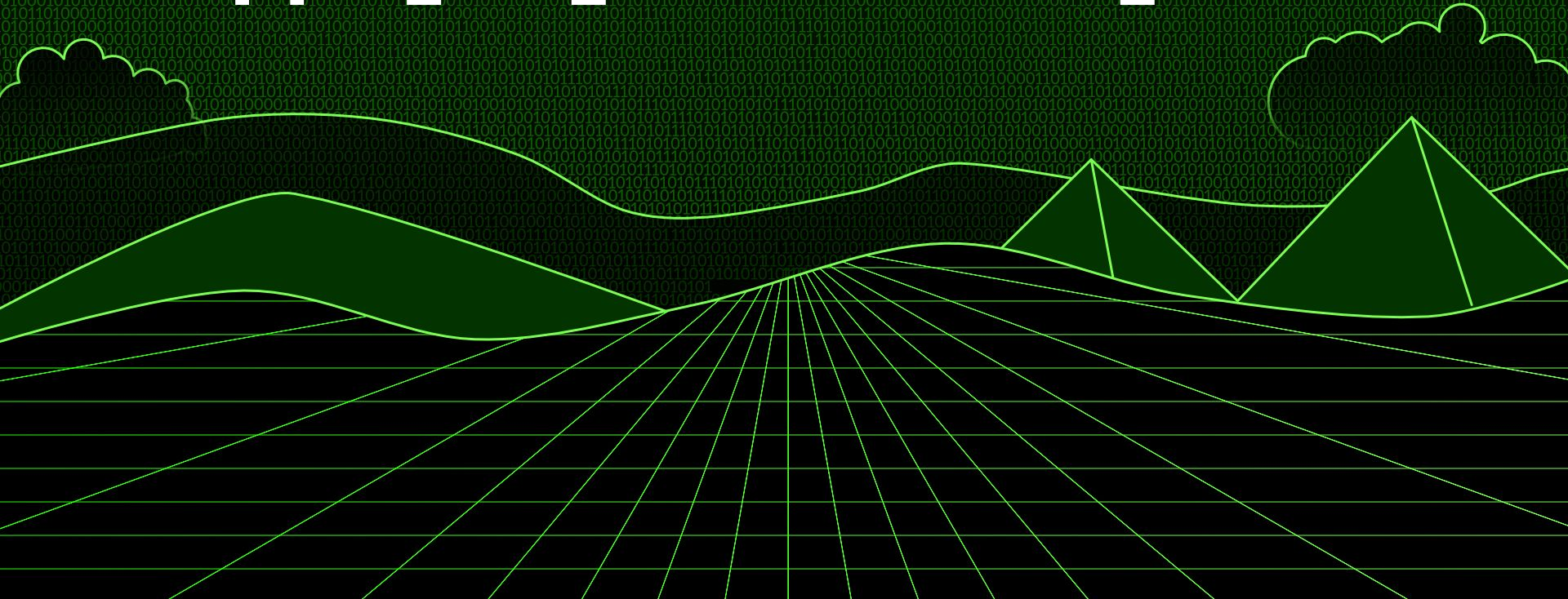
In practice, we can construct (some types of) IRT models using a binomial form of a GLMM.

De Boeck et al., 2011

```
IRTmodel = glmer(response ~ -1 + fixedeffect +  
(1|randomeffect) + (1|randomeffect), family=binomial("logit"),  
data=datasource
```



Applying this to games





# Item Response Theory

- “the probability of a correct response to an item is a mathematical function of person and item parameters”
- Each question posed by a minigame is an item





triumph

batch

utograph

sketch

paragraph

fantastic

scratch





# Item Response Theory

- “the probability of a correct response to an item is a mathematical function of person and item parameters”
- Each question posed by a minigame is an item
- Item parameters include minigame type and feature



# IRT – difficulty model



ability + minigame difficulty + feature difficulty



This allows us to disentangle feature and minigame difficulty



Actually utilises the large dataset - ~432000 item responses



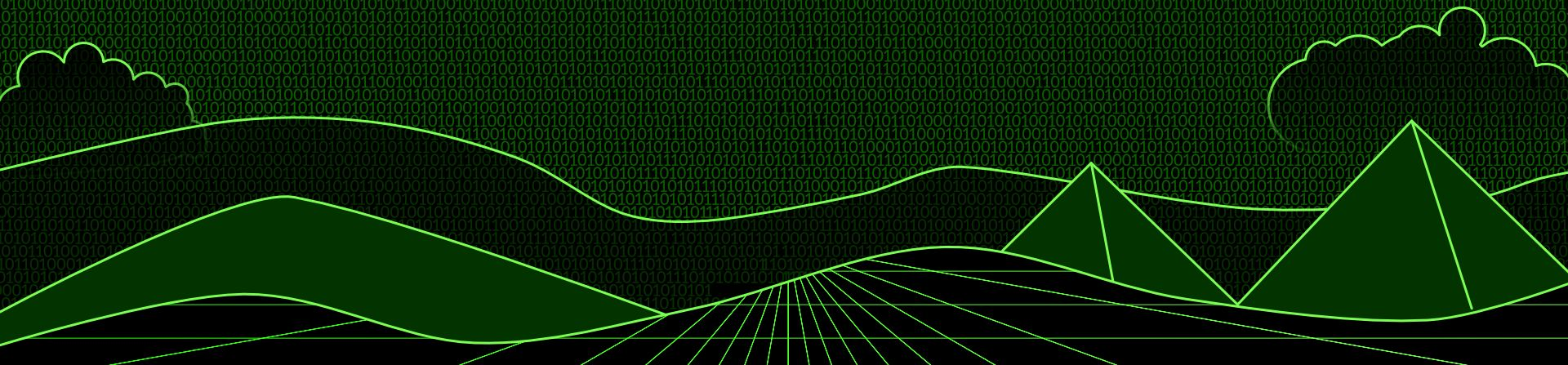
Can create an objective list of relative difficulties of various features for EFL learners



# IRT – difficulty model



ability + minigame difficulty + feature difficulty



game\_name effect plot





game\_name effect plot



# IRT – difficulty model



ability + minigame difficulty + feature difficulty

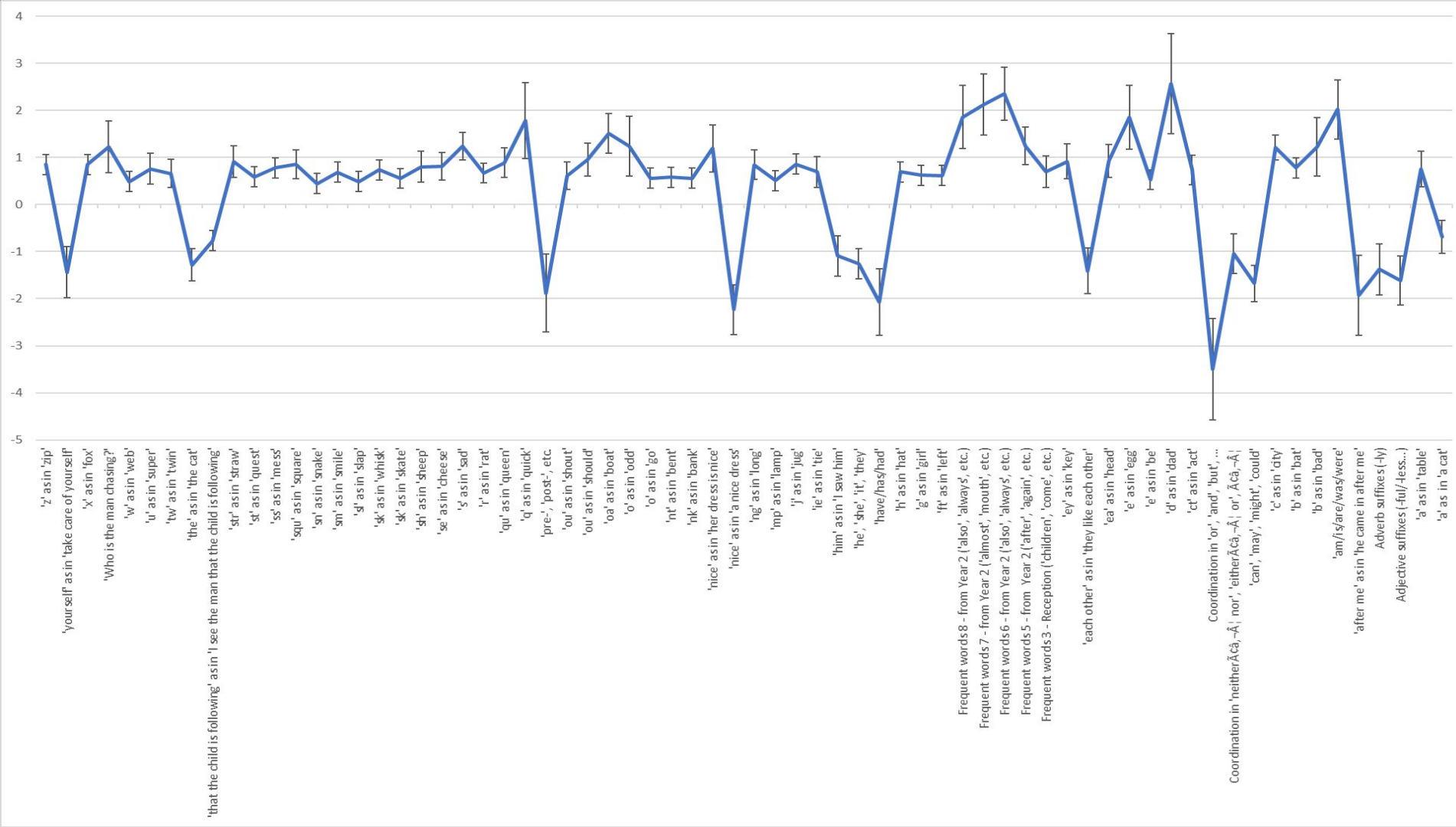


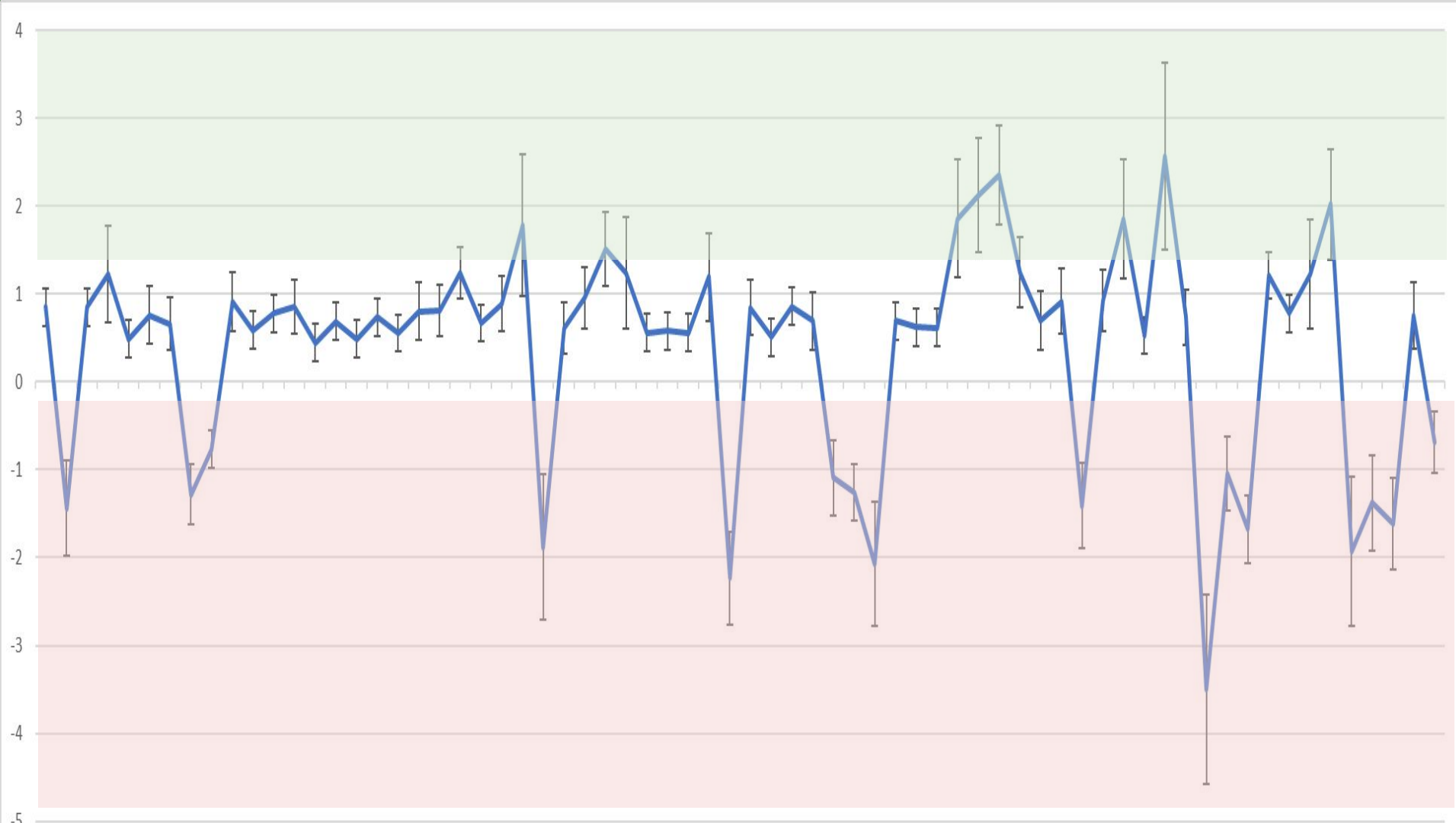
# IRT - difficulty model



ability + minigame difficulty + feature difficulty

	A	B	C	D	E	
1		Estimate	Std. Error	z value	Pr(> z )	
2	description(CC)+S+V+DO+OI(CC)	0.6185341326	0.335053338	1.846076616	0.0648810816	
3	description/a/-a	1.272109561	0.185686308	6.850852769	0.0000000000	
4	description/aa/-aa	0.1474663877	0.210519334	0.7004885698	0.4836222409	
5	description/ae/-ae	0.269436068	0.187413734	1.437653801	0.1505323076	
6	description/ai/-aÄ	0.3433298742	0.187181510	1.834208265	0.0666230851	
7	description/aj/-ai	0.2837153053	0.185918709	1.526018048	0.1270053699	
8	description/aj/-ay	0.2170620969	0.184154865	1.178693253	0.2385203433	
9	description/ao/-ao	-0.018903032	0.221220198	-0.085448943	0.9319044839	
10	description/au/-aÄ²	0.3907264532	0.187493342	2.083948413	0.0371648600	
11	description/aw/-au	0.3181904956	0.185990995	1.710784415	0.0871209176	
12	description/b/-b	0.5442931531	0.187686552	2.900011455	0.0037314902	
13	description/b/-v	0.6372239257	0.191081174	3.334833633	0.0008535045	
14	description/d/-d	1.033677336	0.191290039	5.403717522	0.0000000653	
15	description/e/-e	1.253558949	0.187229133	6.695319937	0.0000000000	
16	description/Ê/-ch	0.6934631719	0.188049770	3.687657634	0.0002263279	
17	description/Ë/-ll	0.6798365367	0.186156271	3.65196687	0.0002602395	
18	description/Ê/-y	0.5233539483	0.189097285	2.767643901	0.0056463112	
19	description/É/-r	-0.026412240	0.185074730	-0.142711217	0.8865182573	







# Easier features

feature	Estimate
'd' as in 'dad'	2.566385
Frequent words 6 ('also', 'always', etc.)	2.354535
Frequent words 7 ('almost', 'mouth', etc.)	2.129937
'am/is/are/was/were'	2.021187
Frequent words 8 ('also', 'always', etc.)	1.862003
'e' as in 'egg'	1.854427
'q' as in 'quick'	1.780006
'oa' as in 'boat'	1.506819



phonology that matches  
with L1



common words



the verb to be



low frequency consonants









# Harder features

feature	Estimate
Coordination in 'or', 'and', 'but', ...	-3.50077
'nice' as in 'a nice dress'	-2.23587
'have/has/had'	-2.07429
'after me' as in 'he came in after me'	-1.93248
'pre-', 'post-', etc.	-1.87932
'can', 'may', 'might', 'could'	-1.6747
Adjective suffixes (-ful/-less...)	-1.61517
'yourself' as in 'take care of yourself'	-1.4372

feature	Estimate
'each other' as in 'they like each other'	-1.4078
Adverb suffixes (-ly)	-1.37697
'the' as in 'the cat'	-1.28361
'he', 'she', 'it', 'they'	-1.25597
'him' as in 'I saw him'	-1.09317
Coordination in 'neither/nor', 'either/or', ...	-1.04798
'that the child is following' as in 'I see the man that the child is following'	-0.7685
'a' as in 'a cat'	-0.68278



# Harder features

-  conjunctions and coordinators
-  anaphora and pronouns
-  prefixes / suffixes
-  articles
-  the verb 'to have'
-  modal verbs



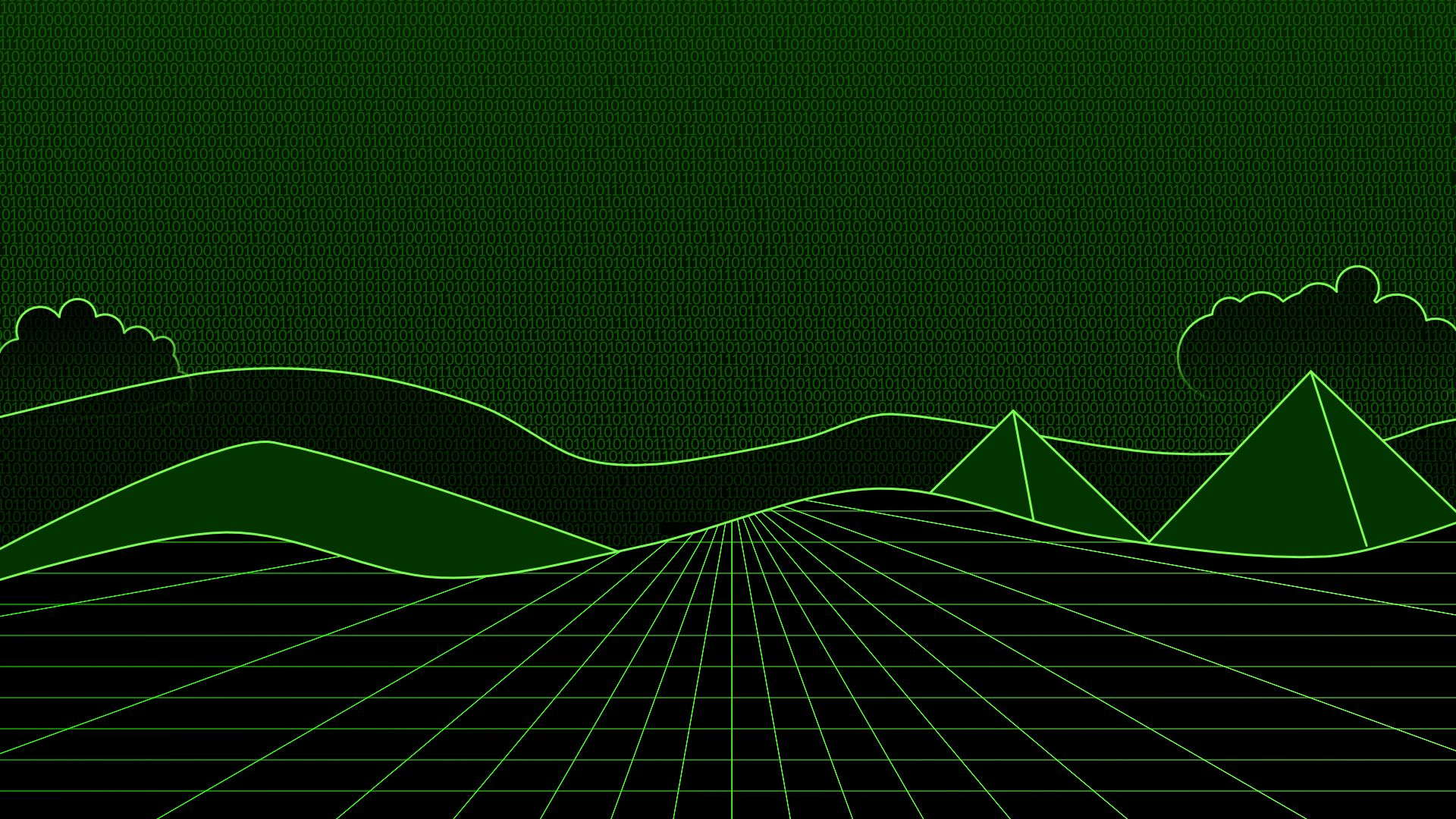
# Comparison - L1 transfer

feature	estimate
'nice' as in 'her dress is nice'	1.192034
'nice' as in 'a nice dress'	-2.23587

Her dress is nice - su vestido es bonito

A nice dress - un vestido bonito







# Conclusions and Discussion

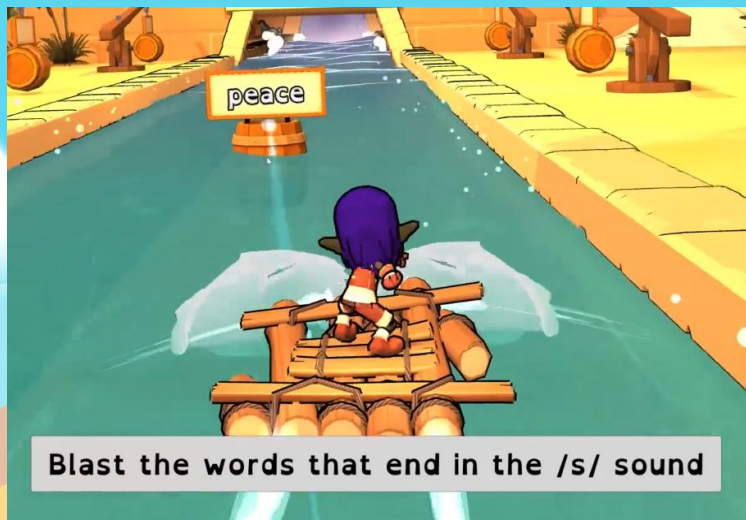
- Using the vast amounts of data from the Navigo games
- and an IRT model
- we can show that some linguistic features are demonstrably easier or harder than others
- for this group of learners (relatively low proficiency, Spanish/Catalan 11-year olds)

# Conclusions and Discussion

- Clearly, there would be differences with other populations.
- The way the features are tested may still make a difference.
- We have only limited explanations as to *why* these features are difficult.
- But, conceptually, this seems to be an objective method for investigating linguistic feature difficulty.
- Any other thoughts or comments?



# Thank you!



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