



UNIVERSITY OF
CAMBRIDGE

Turning the page on paper-based assessments:

Three techniques and one technology to transform patient-reported assessments

Chris Gibbons

Director of Health Assessment and Innovation
The Psychometrics Centre

NIHR Research Fellow
Cambridge Centre for Health Services Research
University of Cambridge

TODAY'S TALK

Describe

- Three cutting-edge techniques and one open-source technology to improve patient reported outcome measures

Demonstrate

- How modern psychometrics can transform patient outcome and experience assessment

Discuss

- Where we might be heading in this world of big data and computational social science?

THE TERMS

- Patient-reported outcome and experience measures
- Psychometrics
- Item response theory
- Item banks
- eHealth
- Machine learning
- Computer adaptive testing
- Computational social science

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THE TERMS

ePsychometrics

- Item response theory
- Item banks
- eHealth
- Machine learning
- Computer adaptive testing
- Computational social science

PROMS AND PREMS

- Thousands, suitable for use in diverse conditions
- Many applications (trials, clinical practice, epidemiology, audit, quality, improvement)
- Patient-centred
- Ad hoc use (*motivated* clinicians)
- Take a *long time* to administer, score and interpret
- Scores can be *difficult* to understand
- Not much incentive

PROMS EVIDENCE

- Systematic reviews (e.g., Valderas *et al.*, 2008)
- Effect of PROM feedback to clinicians on processes and outcomes of care.
- 2/3rd of studies show impact on processes
- Half show impact on outcomes
- Heterogeneity (,of course)

WHAT SEEMS TO WORK?

- Specific PROM training/supported implementation
- Timely feedback
- Well aligned with practice
- Information integrated into available systems
- Information that is linked to specific clinical action
- Interventions with a formal, structured, feedback process perform better (Krageloh, 2014)

BMJ

RESEARCH

Patients' and doctors' views on depression severity questionnaires incentivised in UK quality and outcomes framework: qualitative study

Christopher Dowrick, professor of primary medical care,¹ Geraldine M Leydon, senior research fellow,² Anita McBride, research fellow,² Amanda Howe, professor of primary care,³ Hana Burgess, academic foundation trainee,² Pamela Clarke, research assistant,¹ Sue Maisey, research associate,³ Tony Kendrick, professor of primary medical care²

WHAT DO PATIENTS THINK?

- PROMS for Depression in primary care in the UK
- Patients favoured the PROMS
- They saw them as an efficient and structured supplement to medical judgement
- They saw them as evidence that clinicians were taking their problems seriously through full assessment

UPDATED REVIEW PLANNED 2016



Routine provision of information on patient-reported outcome measures to healthcare providers and patients in clinical practice (Protocol)

Gonçalves Bradley DC, Gibbons C, Ricci-Cabello I, Bobrovitz NJH, Gibbons EJ, Kotzeva A,
Alonso J, Fitzpatrick R, Bower P, van der Wees PJ, Rajmil L, Roberts NW, Taylor RS,
Greenhalgh J, Porter I, Valderas JM

PREM PROBLEMS

GP PATIENT SURVEY

How good is your GP at taking your problems seriously?


Very poor	Poor	Neither poor nor good	Good	Very good
9.1%		21.2%	69.7%	

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PREM PROBLEMS

People who said “Good” or “Very good”

“Taking your problems seriously? Well no he didn’t.”

“No, he didn’t listen to me.”

“Well no, he didn’t really ask about symptoms.”

IMPROVING MEASUREMENT IN CLINICAL PRACTICE

- Tackle the barriers!
- Make measurement more efficient
- Engage patients and clinicians with instant feedback
- Align feedback more closely with care
- Address response biases

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Hypothesis:
All possible using ePsychometrics

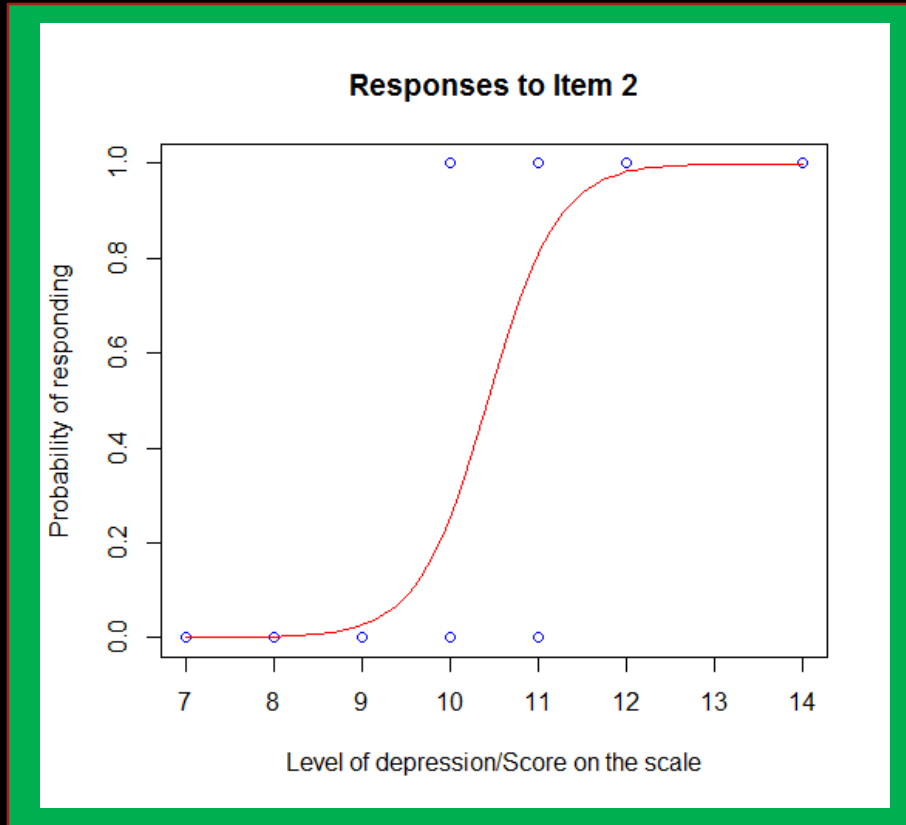
IMPROVING MEASUREMENT IN CLINICAL RESEARCH

ePsychometrics can also transform..

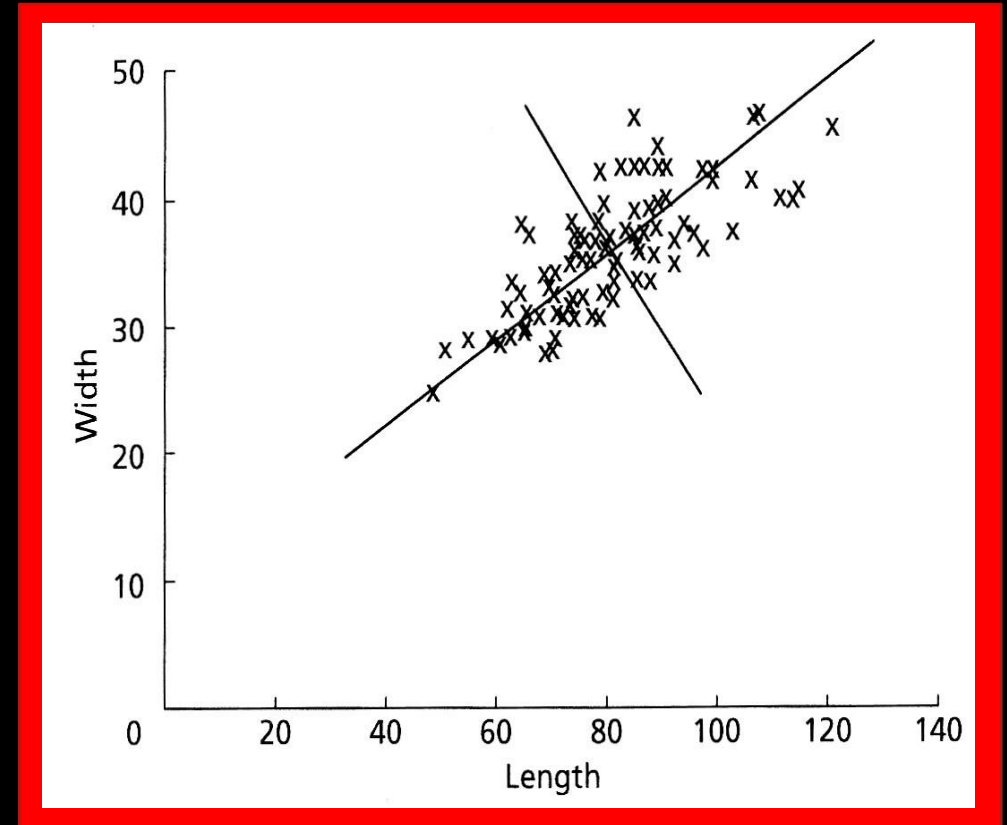
- Clinical trials (recruitment, reliability, attrition, comparability)
- Comparative research
- Patient-centred Big Data analytics

WHAT IS 'PSYCHOMETRICS'?

Item response theory



Classical test theory



The Psychometrics Centre

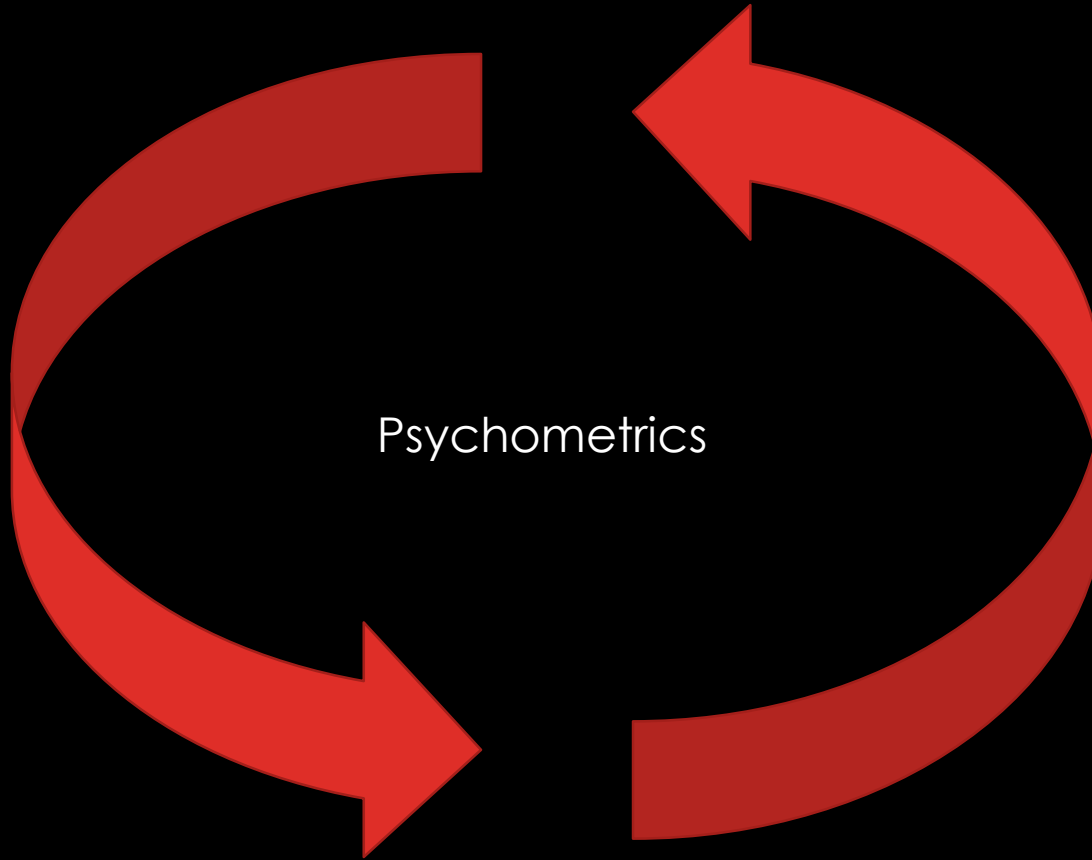


Psychometrics

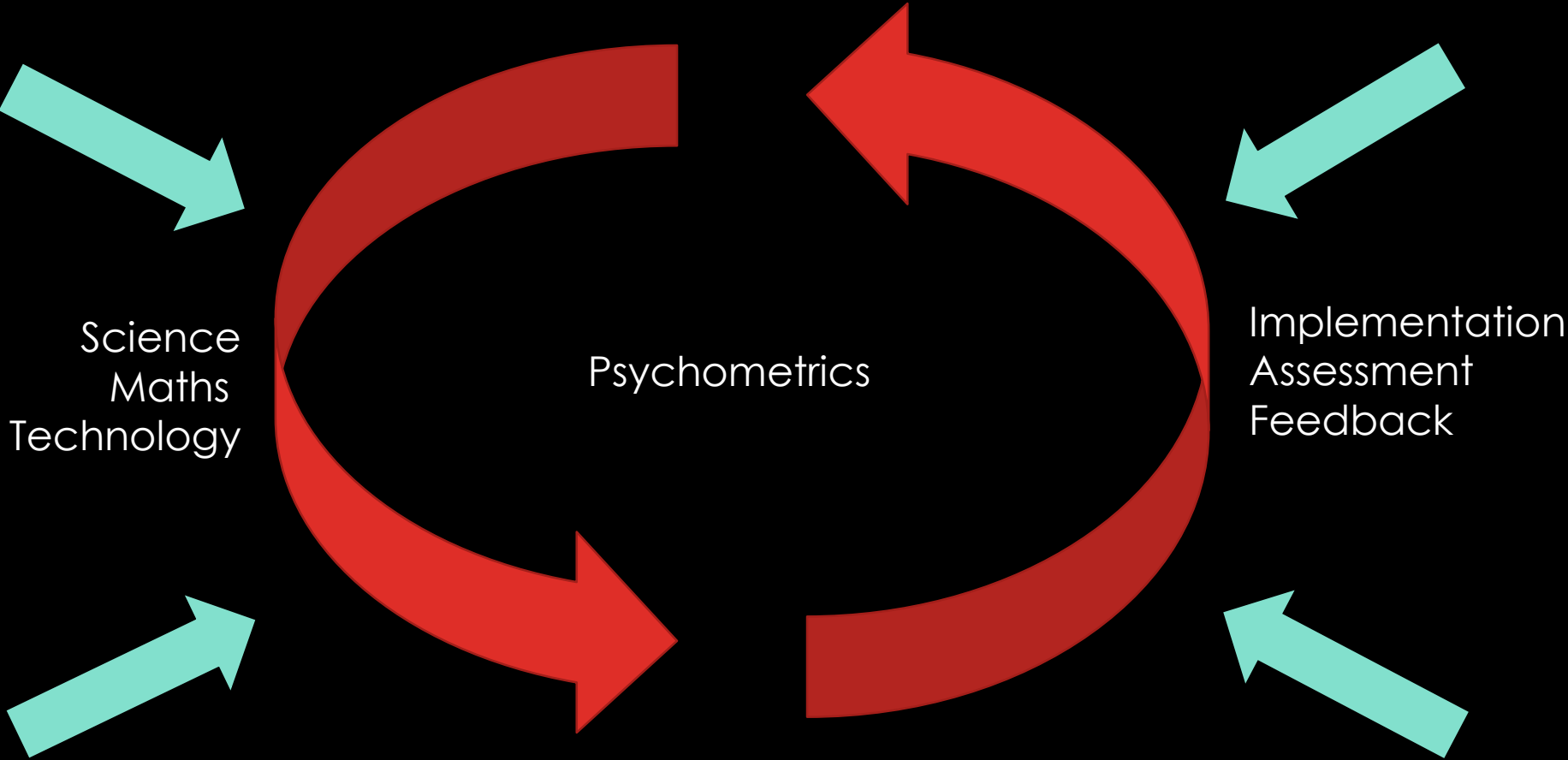
Science
Maths
Technology

Psychometrics

Implementation
Assessment
Feedback



The Psychometrics Centre



Research, leadership, innovation, support

THREE TECHNIQUES & ONE TECHNOLOGY

Computer
adaptive testing

(Efficiency and precision)

Prediction and
feedback

(Validation and engagement)

Machine learning

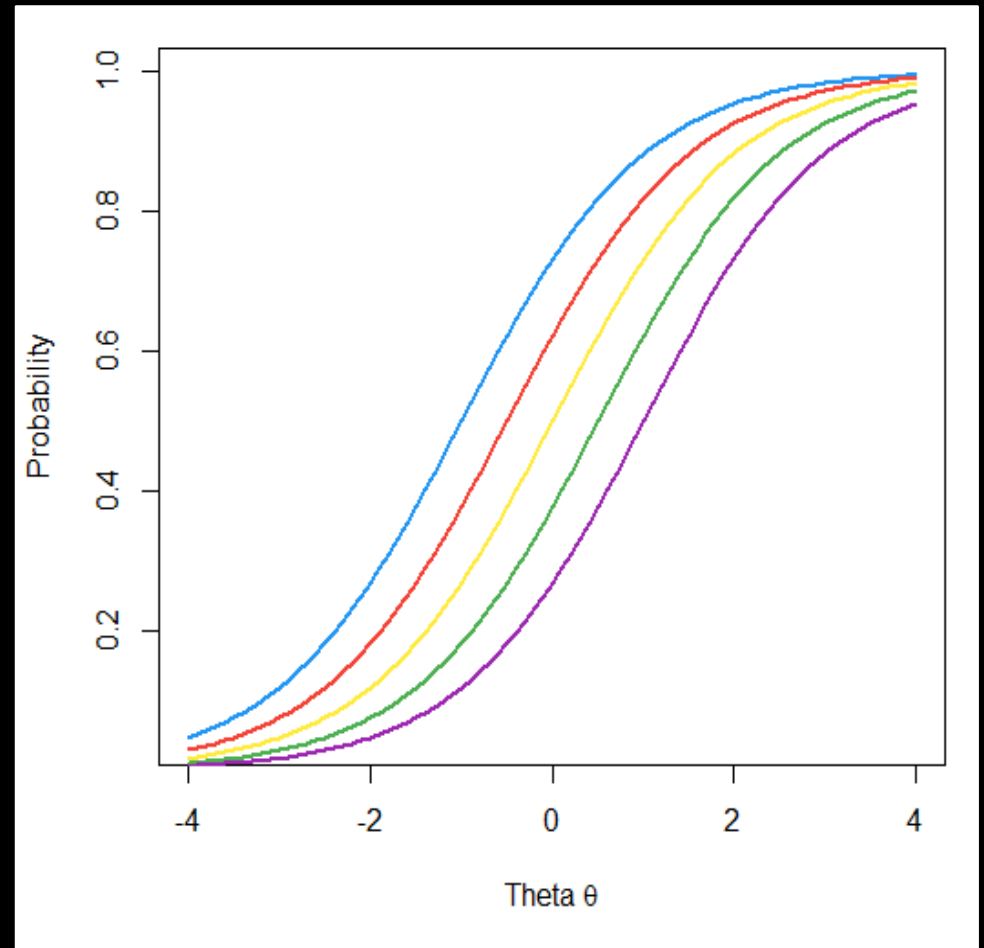
(New forms of data)



DEVELOPING A PREDICTIVE COMPUTER ADAPTIVE TEST FOR THE WHOQOL QUALITY OF LIFE MEASURE

ITEM RESPONSE THEORY

- Probabilistic psychometric theory
- The more you have the more you agree...
- Sample free and test free
- Individual reliability
- Permits computer adaptive testing
- Allows predictions

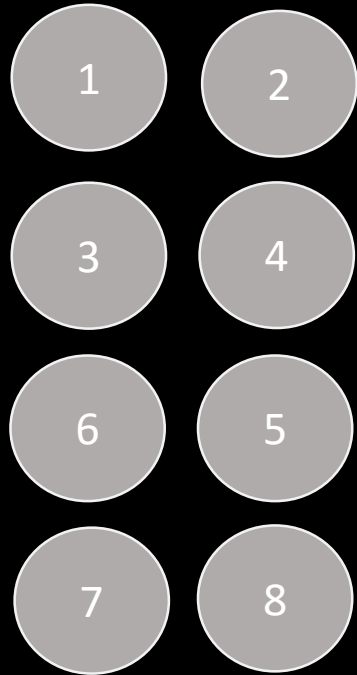


COMPUTER ADAPTIVE TESTING

- “Computer system which iteratively ‘learns’ about the test taker and selects the best item from a large ‘bank’ of items”
- Interfaces with item response theory
- Could be efficient, more reliable and better targeted
- Increasing use
- Used in international high stakes educational assessments (USA, UK and Australia)

COMPUTER ADAPTIVE TESTING

Item Bank



Computer
adaptive test

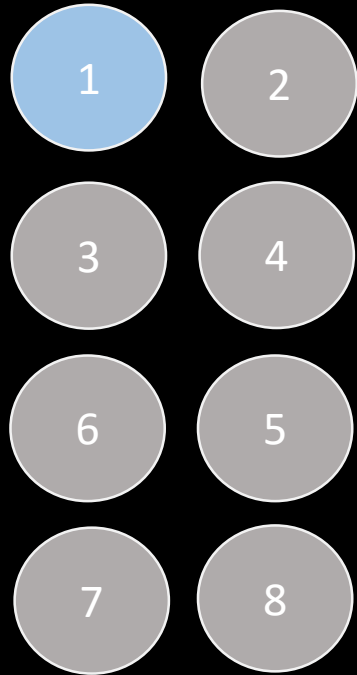


Participant



COMPUTER ADAPTIVE TESTING

Item Bank



Computer
adaptive test



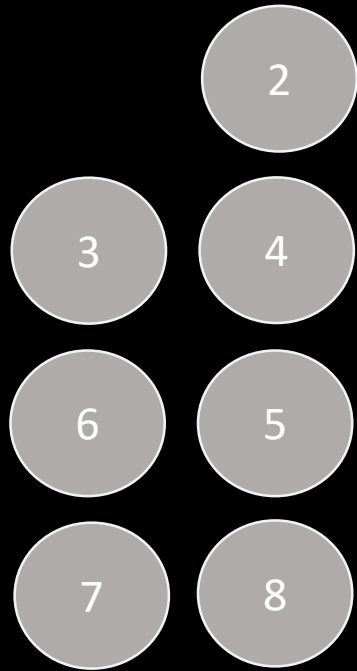
Best item

Participant



COMPUTER ADAPTIVE TESTING

Item Bank



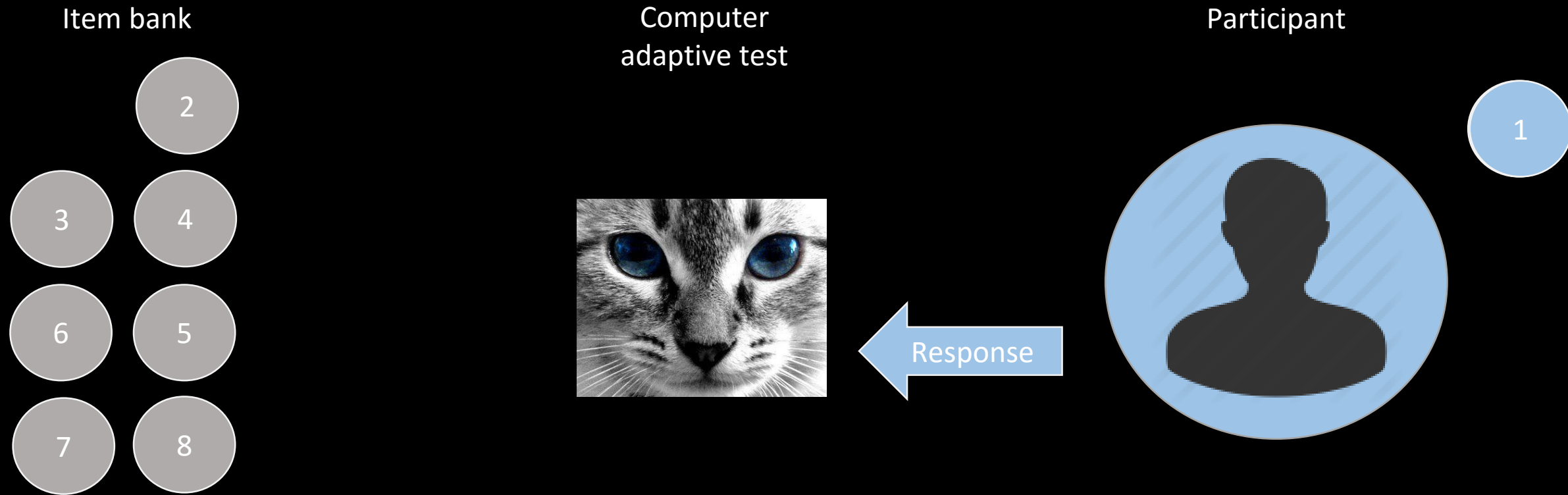
Computer adaptive test



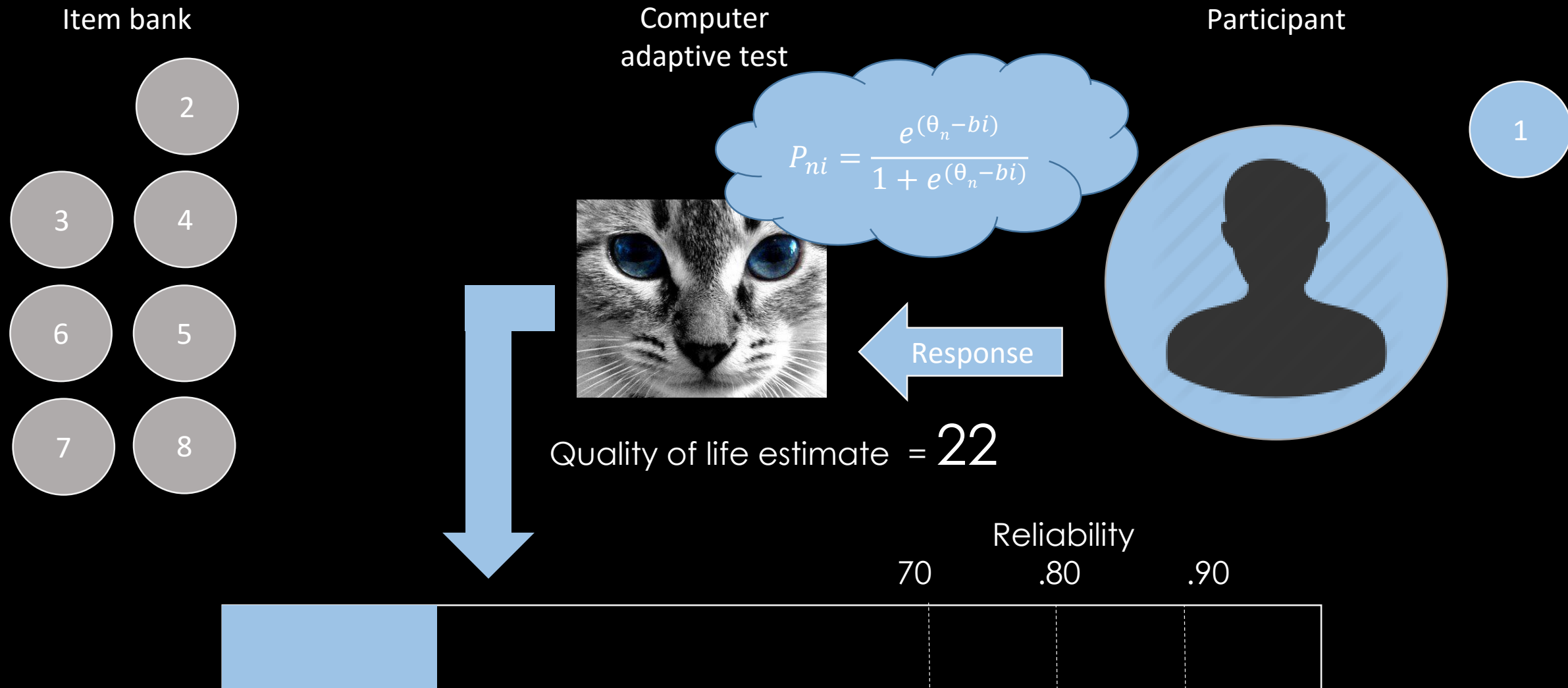
Participant



COMPUTER ADAPTIVE TESTING

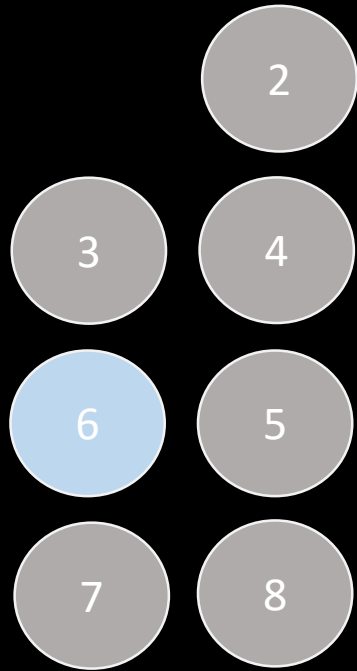


COMPUTER ADAPTIVE TESTING



COMPUTER ADAPTIVE TESTING

Item Bank



Computer adaptive test



Participant



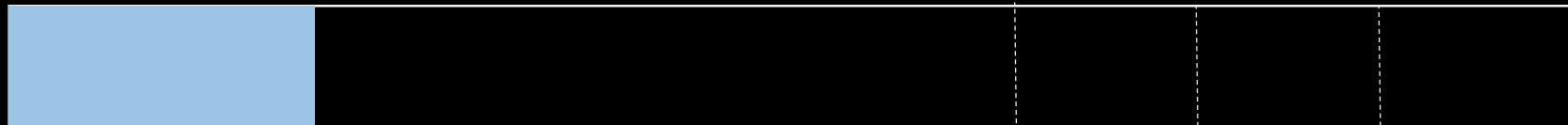
Quality of life estimate = 22

Reliability

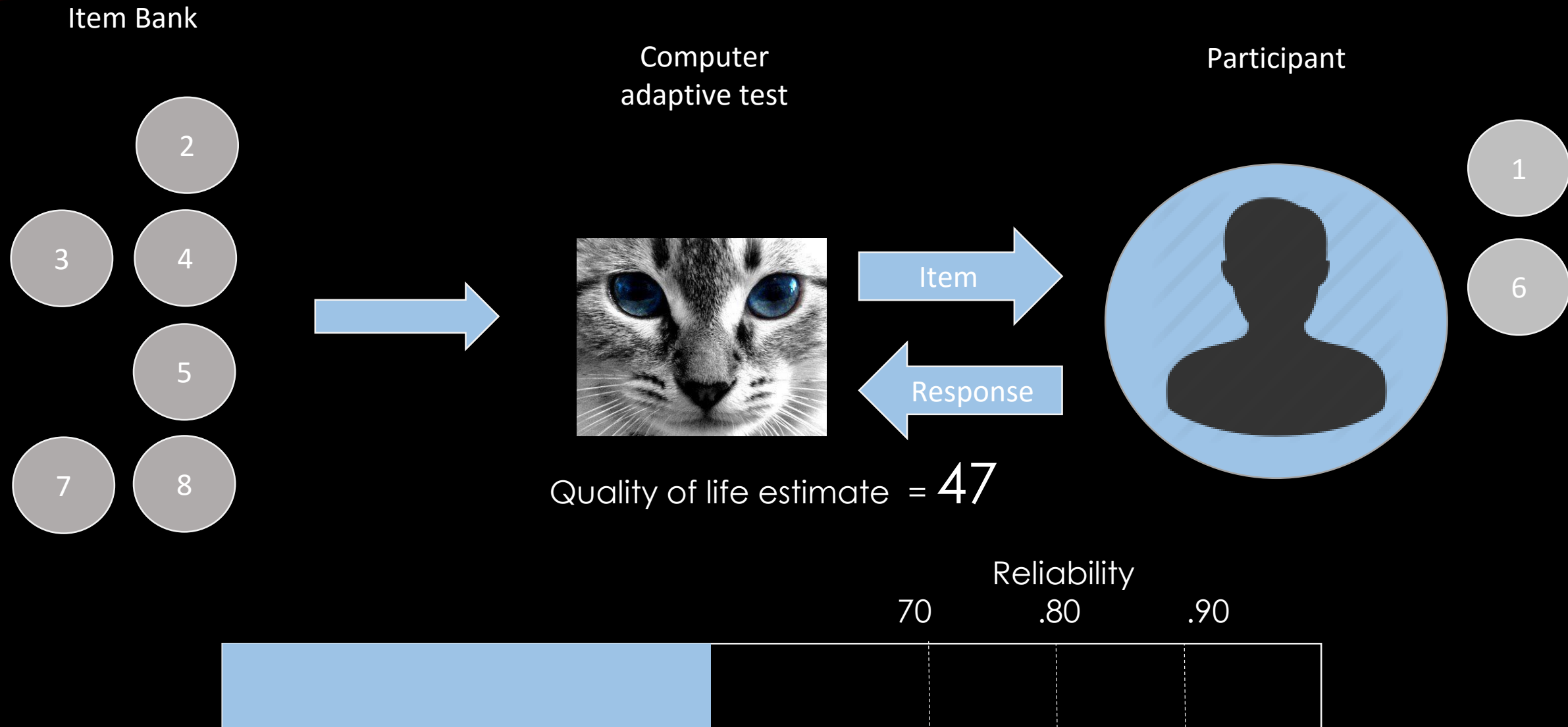
70

.80

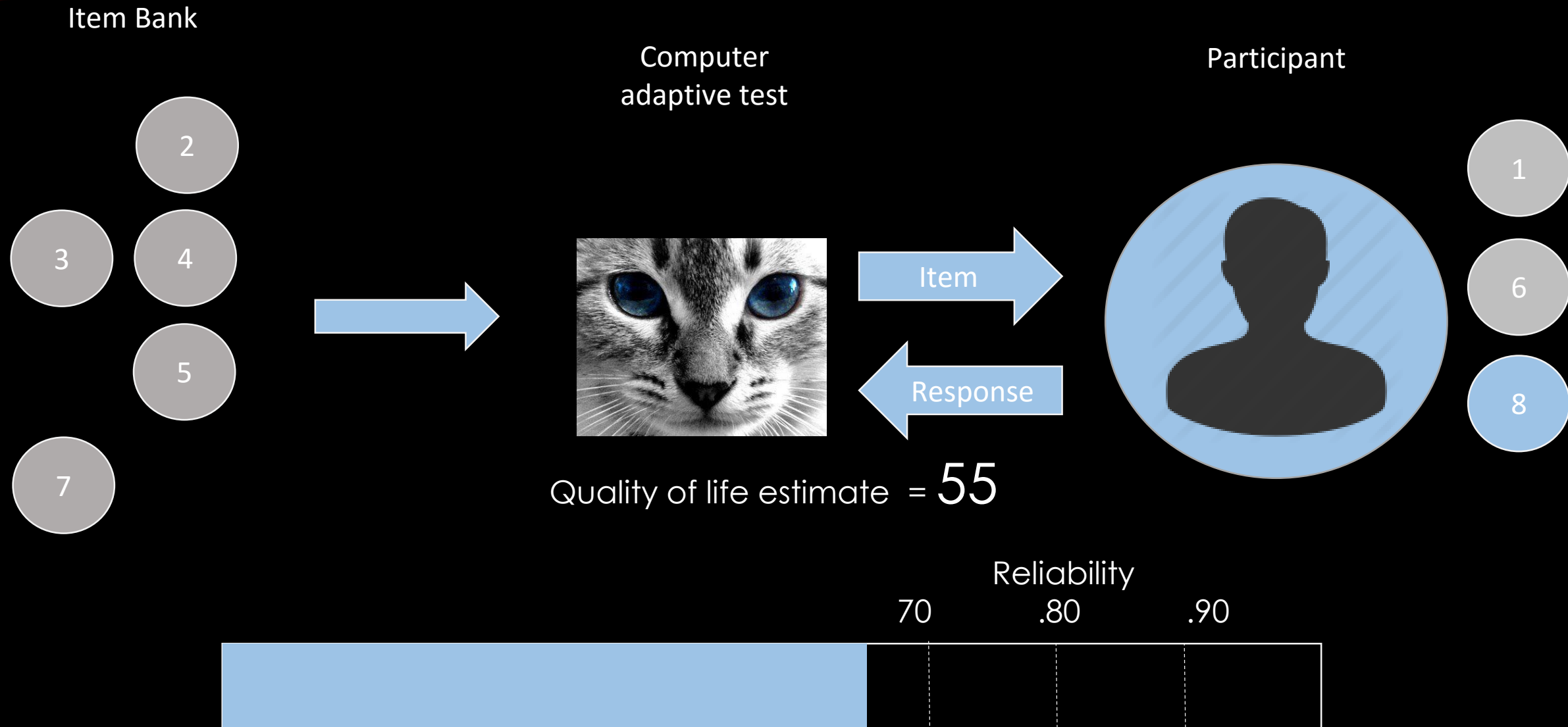
.90



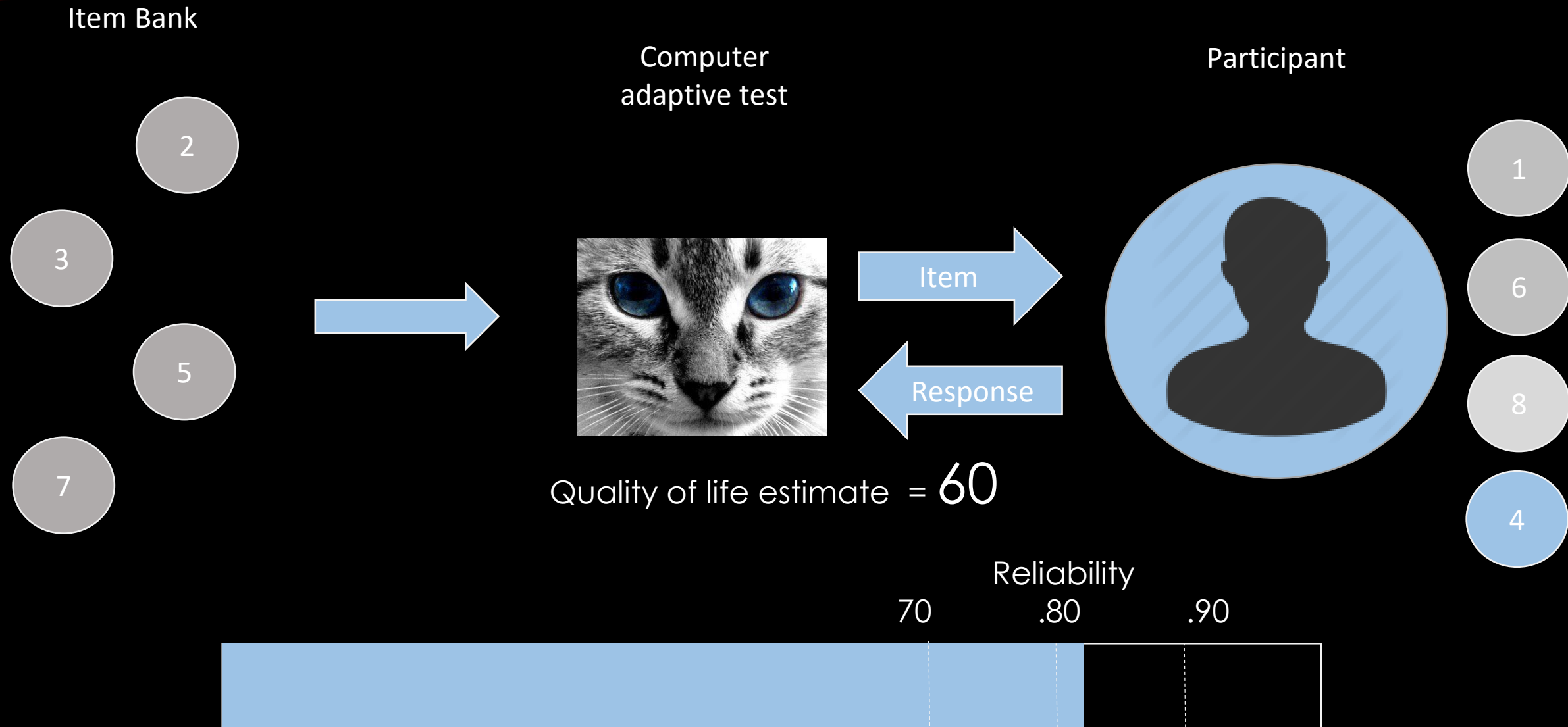
COMPUTER ADAPTIVE TESTING



COMPUTER ADAPTIVE TESTING



COMPUTER ADAPTIVE TESTING

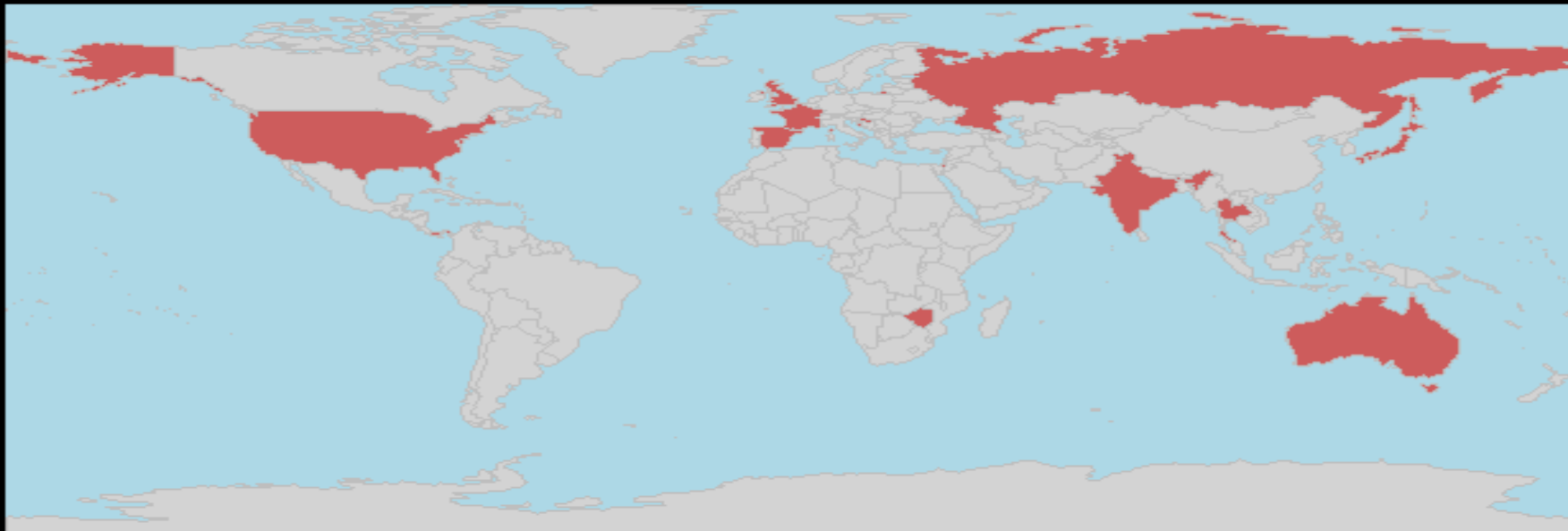


COMPUTER ADAPTIVE TESTING



The WHOQOL

- Multi-dimensional measure of global quality of life
- Physical, psychological, social and environmental domains
- Fifteen field centers



Item bank development

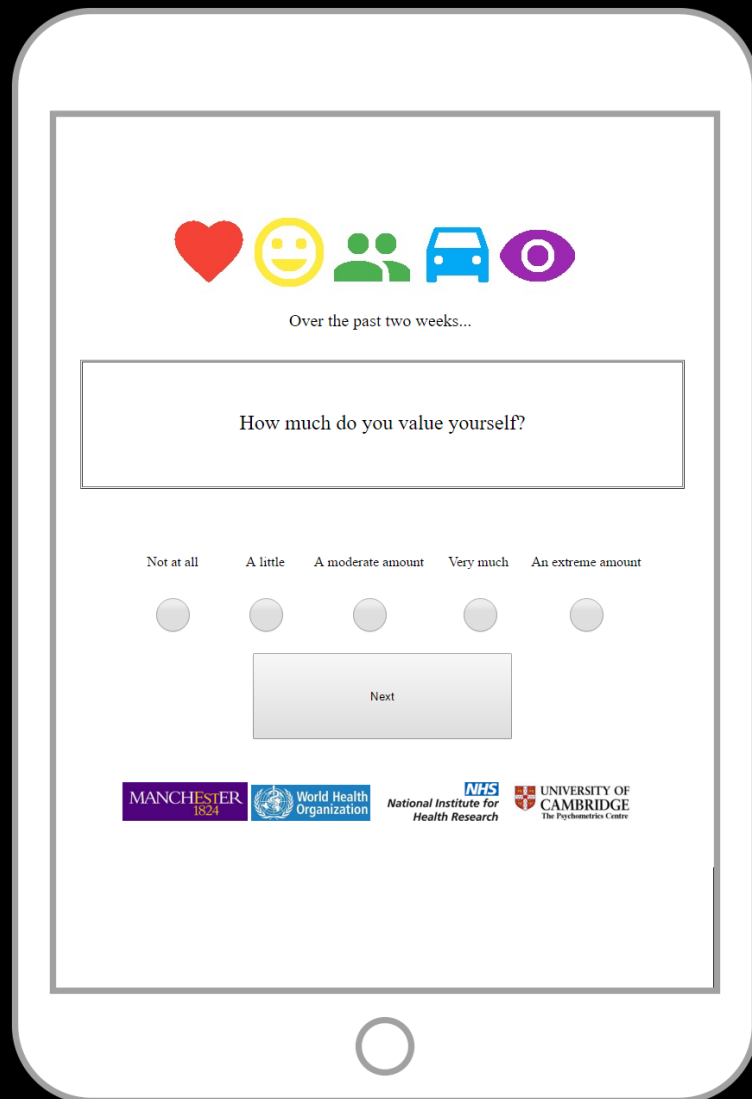
- IRT analysis of the WHOQOL-100 on UK sample ($n = 320$)
- Four domains fit the Partial Credit Model ($p > 0.5$)
- 52 items removed
- Mean 11 items per bank
- Banks were suitable for patients with long-term conditions

Computer adaptive test

- WHOQOL can be 82% shorter *and* more reliable
 - (Gibbons *et al.*, *In Press*. J Med Int Res)

- WHOQOL-CAT assessments are accurate and comparable between countries
 - (Gibbons *et al.*, 2015 .Qual Life Res)

Computer adaptive test







The image shows a smartphone screen displaying a computer adaptive test interface. At the top, there are five icons: a red heart, a yellow smiley face, two green people, a blue car, and a purple eye. Below the icons, the text reads "Over the past two weeks...". The main question is "How much do you value yourself?". Below the question, there are five radio buttons labeled "Not at all", "A little", "A moderate amount", "Very much", and "An extreme amount". A "Next" button is positioned below the radio buttons. At the bottom of the screen, there are logos for MANCHESTER 1824, World Health Organization, National Institute for Health Research, and UNIVERSITY OF CAMBRIDGE The Psychometrics Centre.



Personalised Quality of Life (QoL) feedback

Unique Identifier - S5QHJ

Scale		Score
Physical QoL		60
Psychological QoL		71
Social QoL		53
Environmental QoL		76

The scores above are worked out from the answers that you gave to the questions you have just completed. Your results are given on the chart above. A higher score means that you have a higher quality of life. The table below gives you a little bit more information about what each of the scores mean.

Physical Quality of Life

Your physical quality of life includes things like how well you are able to move around, how much energy you have or how much you are in pain.

Your score of 60 on this scale indicates that your physical quality of life is normal. The majority of people in the United Kingdom report a similar quality of life to you.

If you are worried about your physical quality of life then you should contact your doctor. You can [click here to find a local doctor](#) or [click here to access local emergency services](#)

PREDICATE STUDY

- Predictions in a Computer Adaptive Testing Environment
- Assess validity in the real-world
- CAT is completed until stopping rule is met
- An algorithm predicts what test-takers would answer to unseen questions
- Test-takers rate the predictions (Correct, close, wrong)
- PREDICTMYQOL.com

PREDICATE module



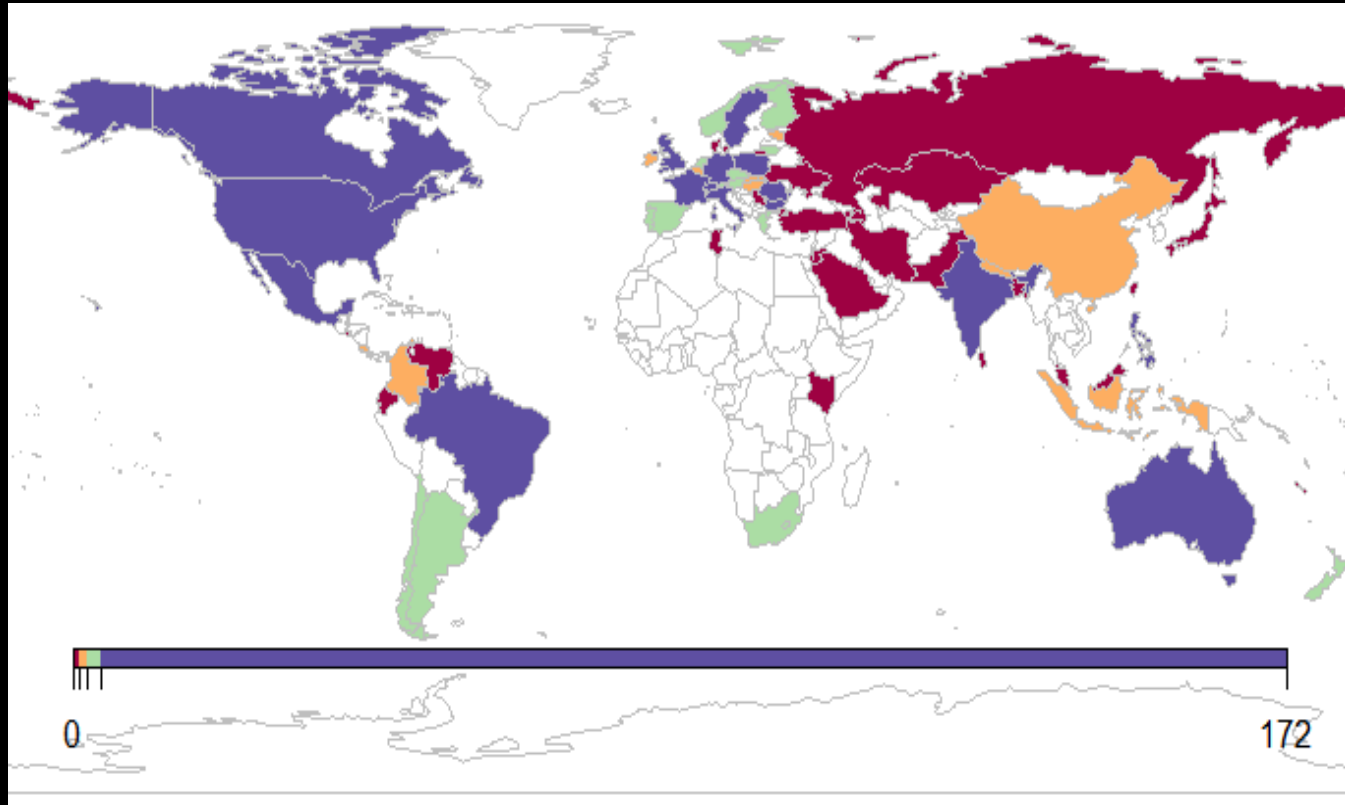
Here are the four questions and the answers we predict that you would have given, based on your previous responses.

Question 1	Do you have enough energy for everyday life?
Answer	Moderately

We are 69% sure that we're right, what do you think?

Exactly right, just what I was thinking!	<input type="radio"/>
Close, but it's not quite right	<input type="radio"/>
The prediction is wrong. I was thinking of a very different answer	<input type="radio"/>

PREDICATE Study



- 699 completions (52 countries)
- 173 participants from the UK
- Average age = 45 ± 10
- 56% male
- 50 patients from UK report long-term conditions (diabetes, depression, arthritis, chronic obstructive pulmonary disease, coronary heart disease)

PREDICATE results

- For the UK sample, results were, on average, 94% “right” or “close”
 - ‘Exactly right’ 69% of the time (predicted accuracy was 78%)
 - “Close, but not quite right” 25% of the time
 - “Wrong” 6% of the time [non-UK 10% wrong]
- Prediction performance was better in the UK (where item bank is validated) than in the rest of the world

ATLanTiC conclusions


- The item bank is brief and can be administered in 120 seconds
- As much as a 82% increase in efficiency compared to paper-based versions
- Simulations translated well to 'real world' CAT
- Prediction study supports the validity of the measurement model and 'backfilling' unanswered item bank items
- Feedback well received and instantaneous
- People like the predictions

USING MACHINE LEARNING TO MAKE SENSE OF GP PERFORMANCE DATA

OPEN TEXT FEEDBACK

- Many questionnaires include open-text elements to add further information
- May contain important information missed by questionnaires
- Typically underused / ignored
- Time-consuming to use human analysts

PATIENT EXPERIENCE

 **Jen Practice**
@JPract Follow

Reception staff at the GP are SO rude
#howrude

Reply Retweet Favorite More

4:47 AM - 4 Apr 16 · Embed this Tweet

 **Mr. A Patient**

Waiting ages for GP visit, and then it was over before it got started! The NHS has gone to the dogs!

Like · Comment · 9 minutes ago · 

Like 24 people like this.

 Write a comment ...

 **SMI Patient**
@SMIpatient Follow

I've got my own care plan FINALLY
#woohoo #mentalhealth #NHS

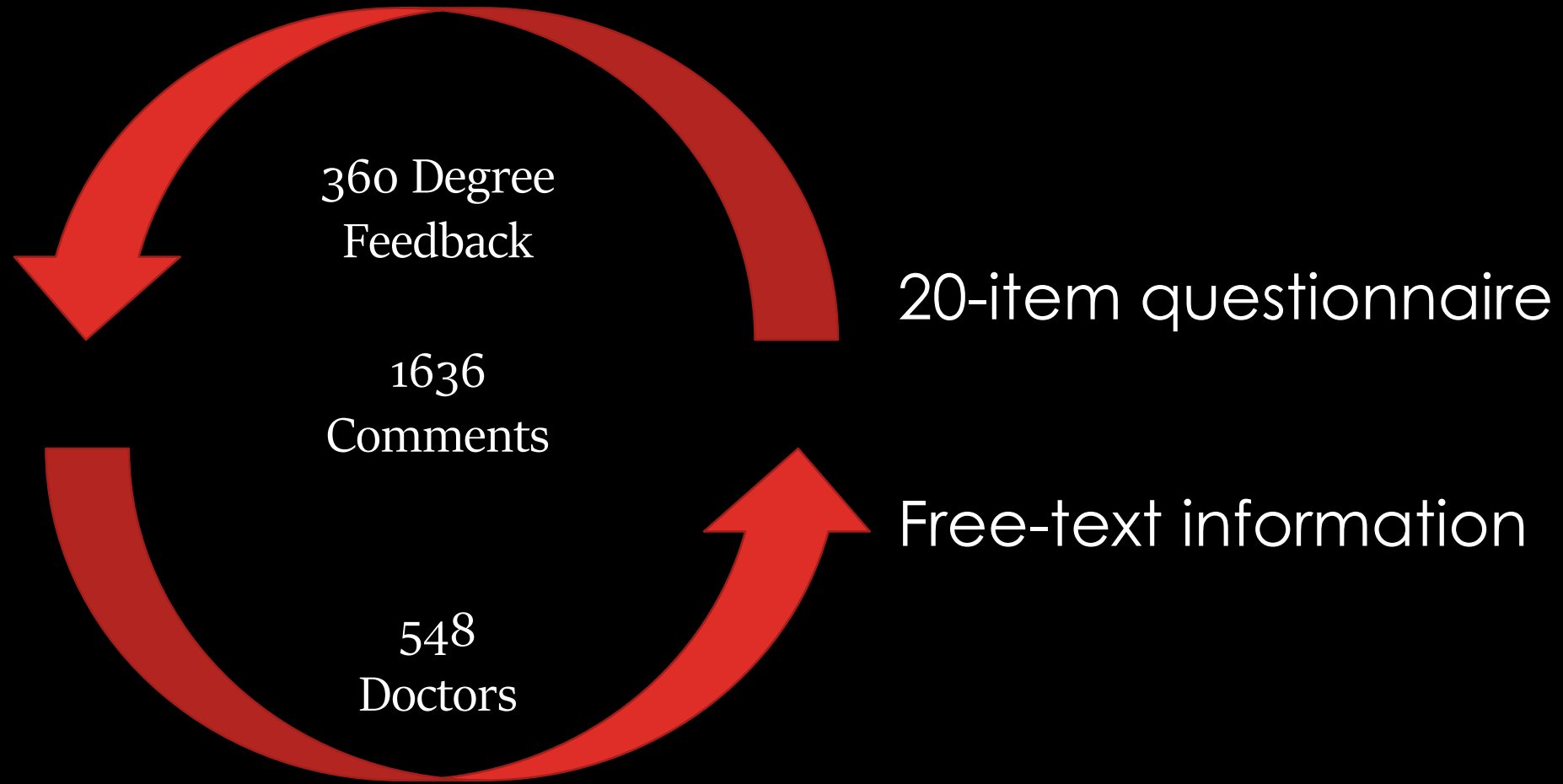
Reply Retweet Favorite More

4:42 AM - 4 Apr 16 · Embed this Tweet

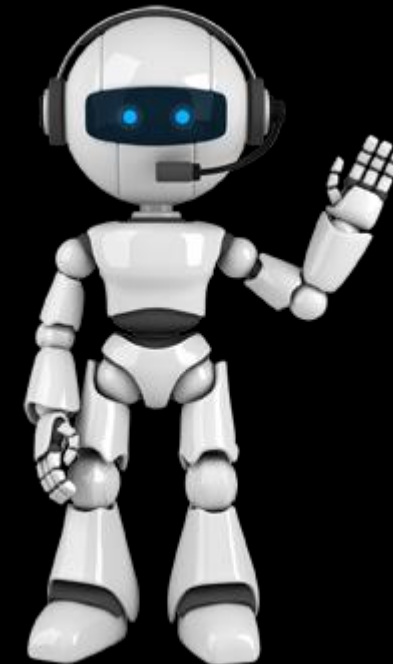
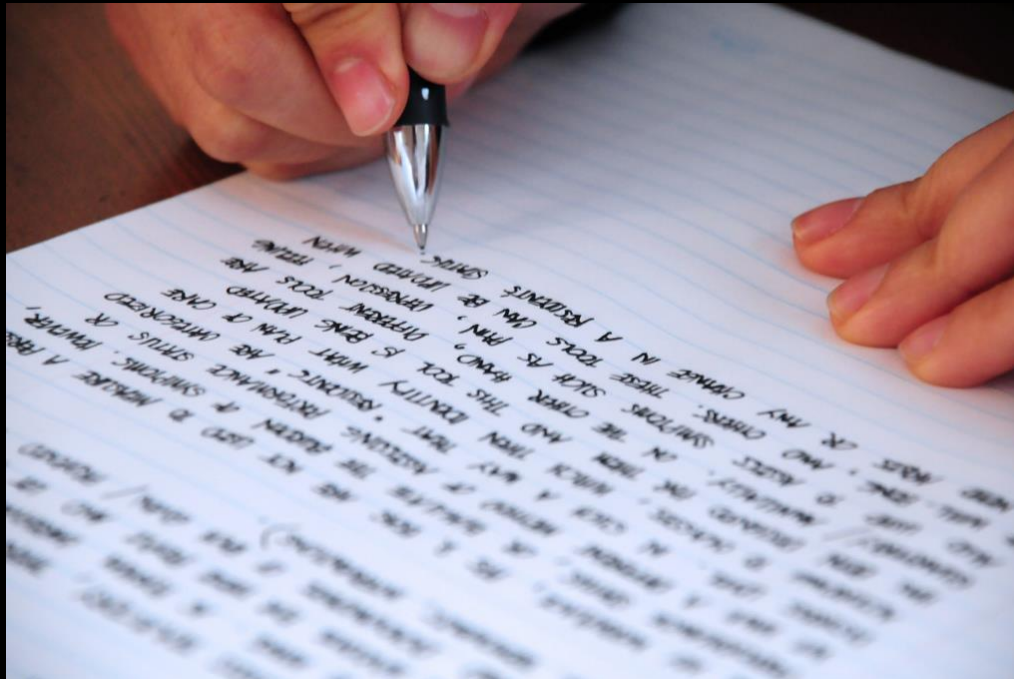
GP PATIENT SURVEY

We can make *sense* of this data using machine learning and natural language processing

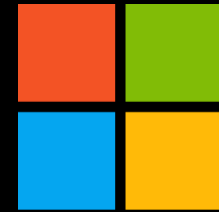
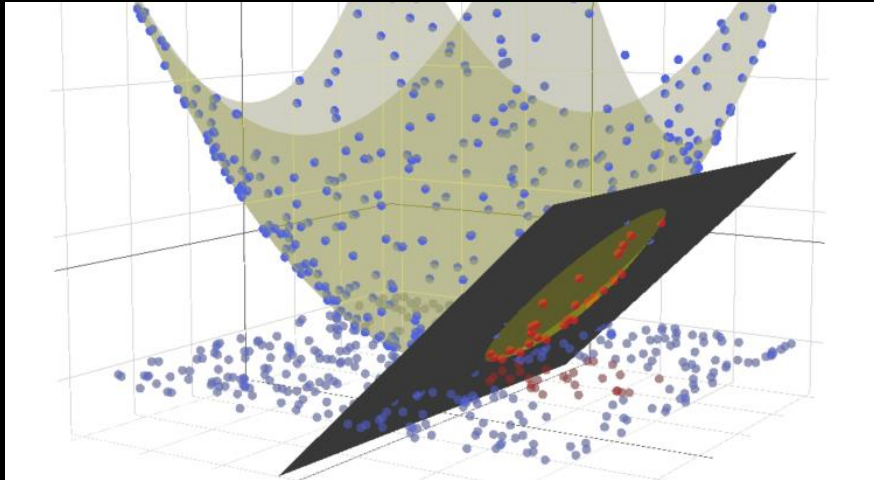
GMC Colleague Questionnaire



MACHINE LEARNING



MACHINE LEARNING

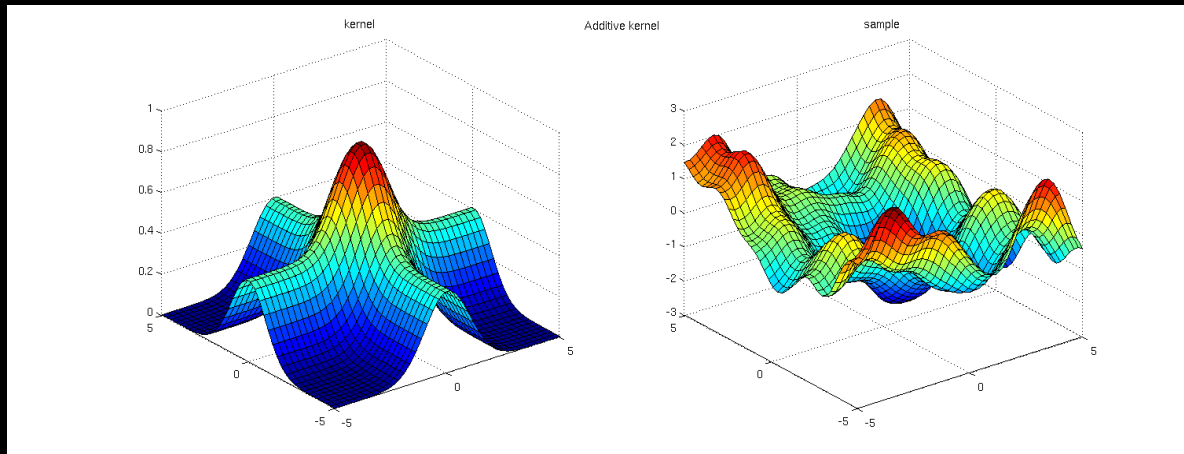


Microsoft

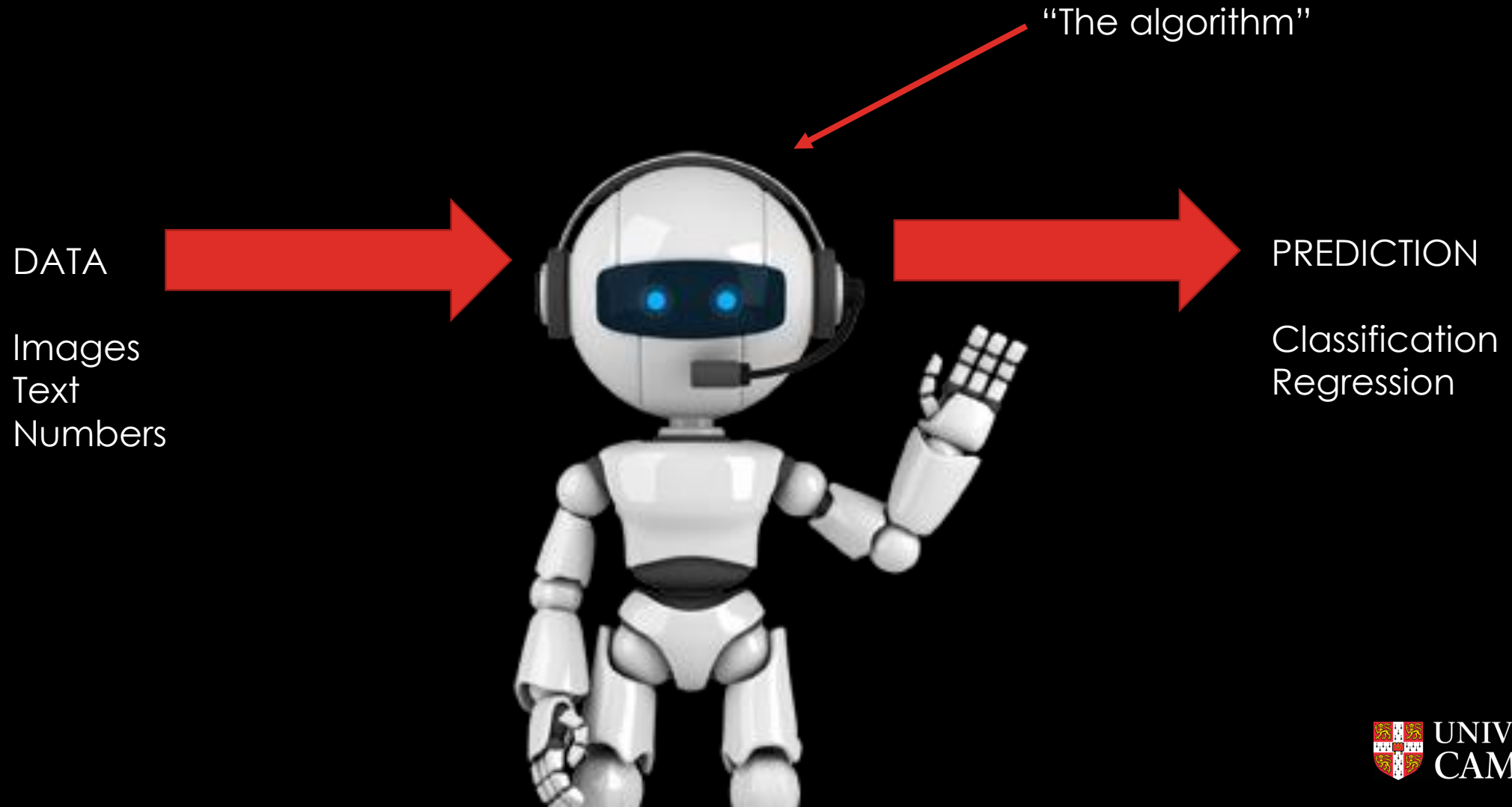


Google

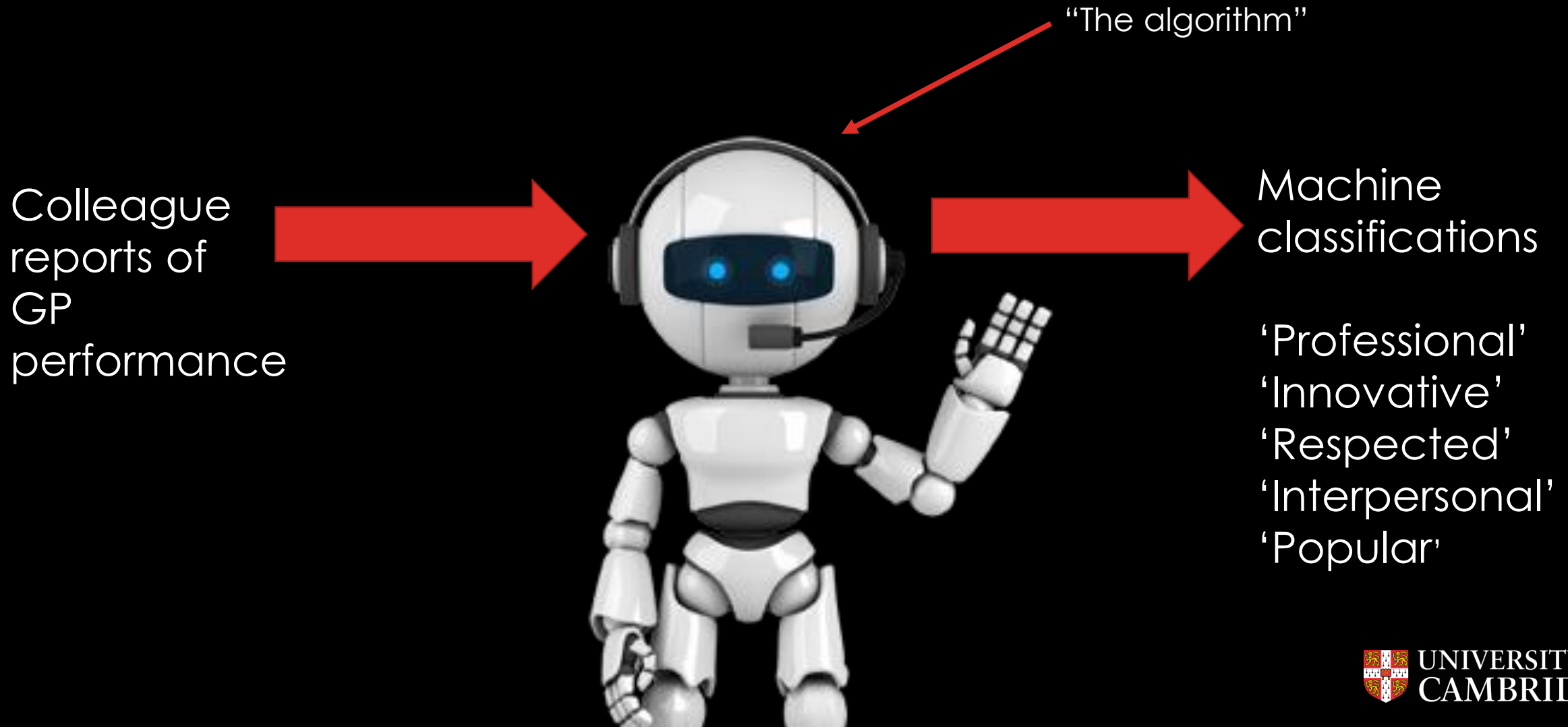
amazon.com[®]



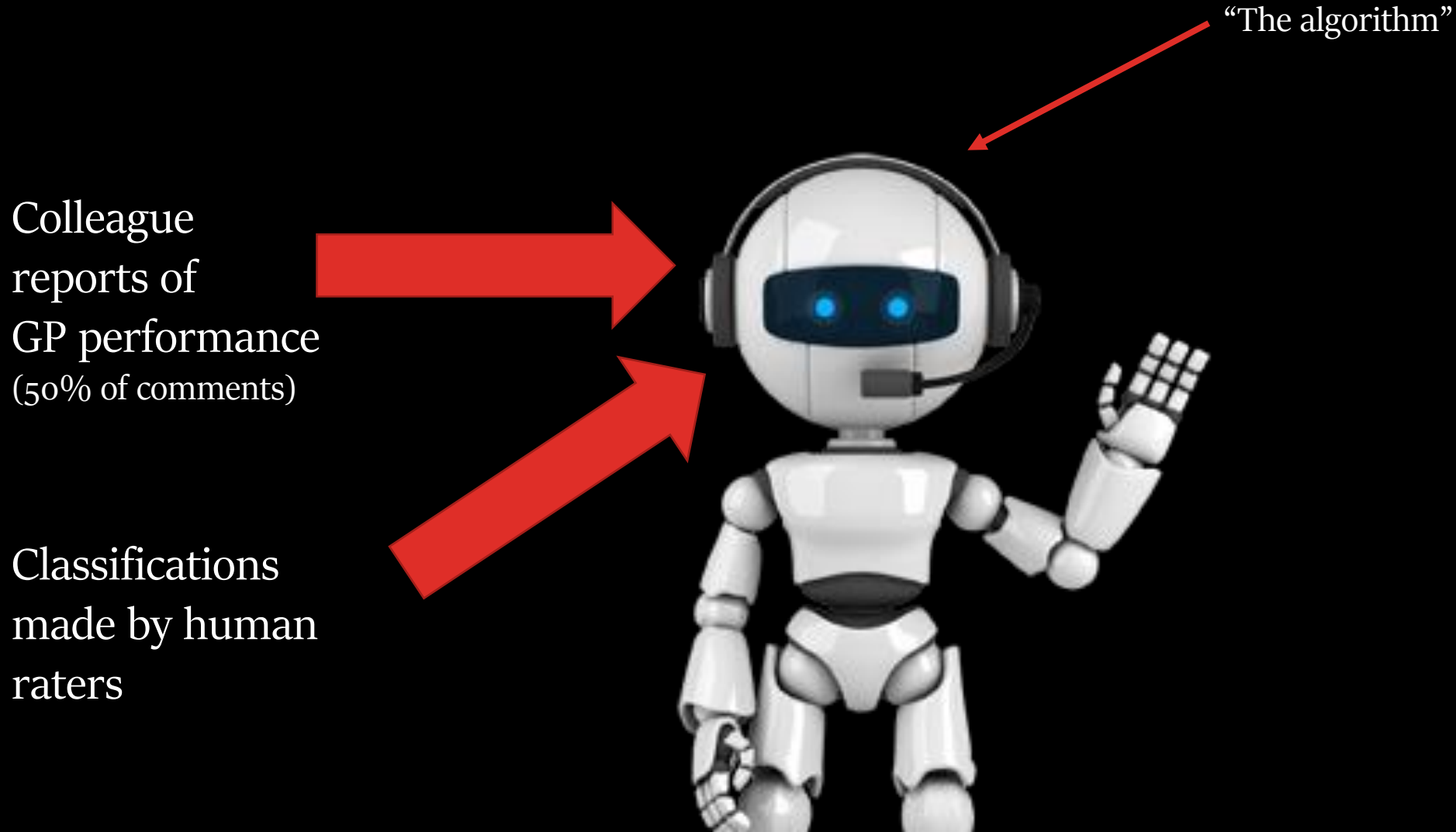
MACHINE LEARNING



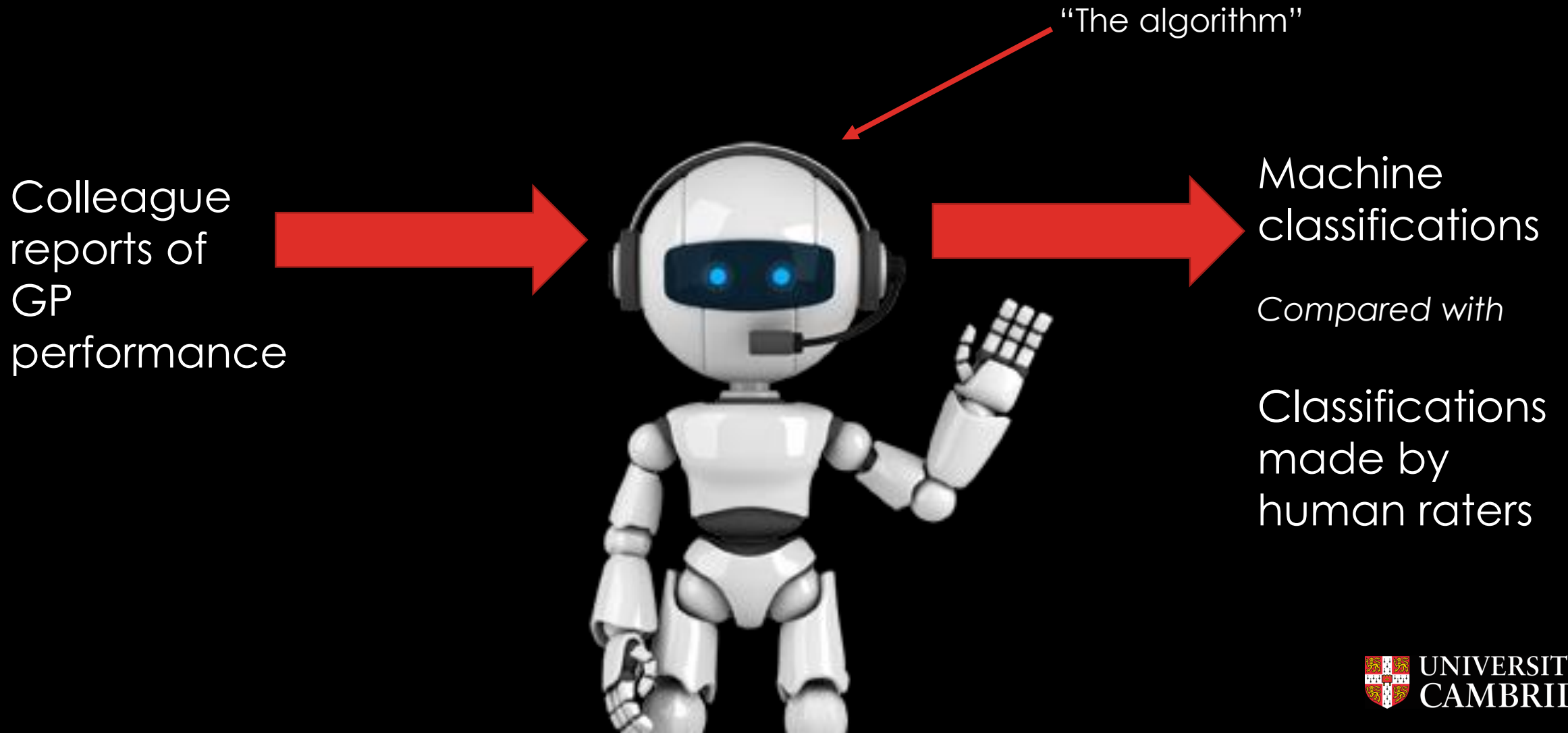
MACHINE LEARNING



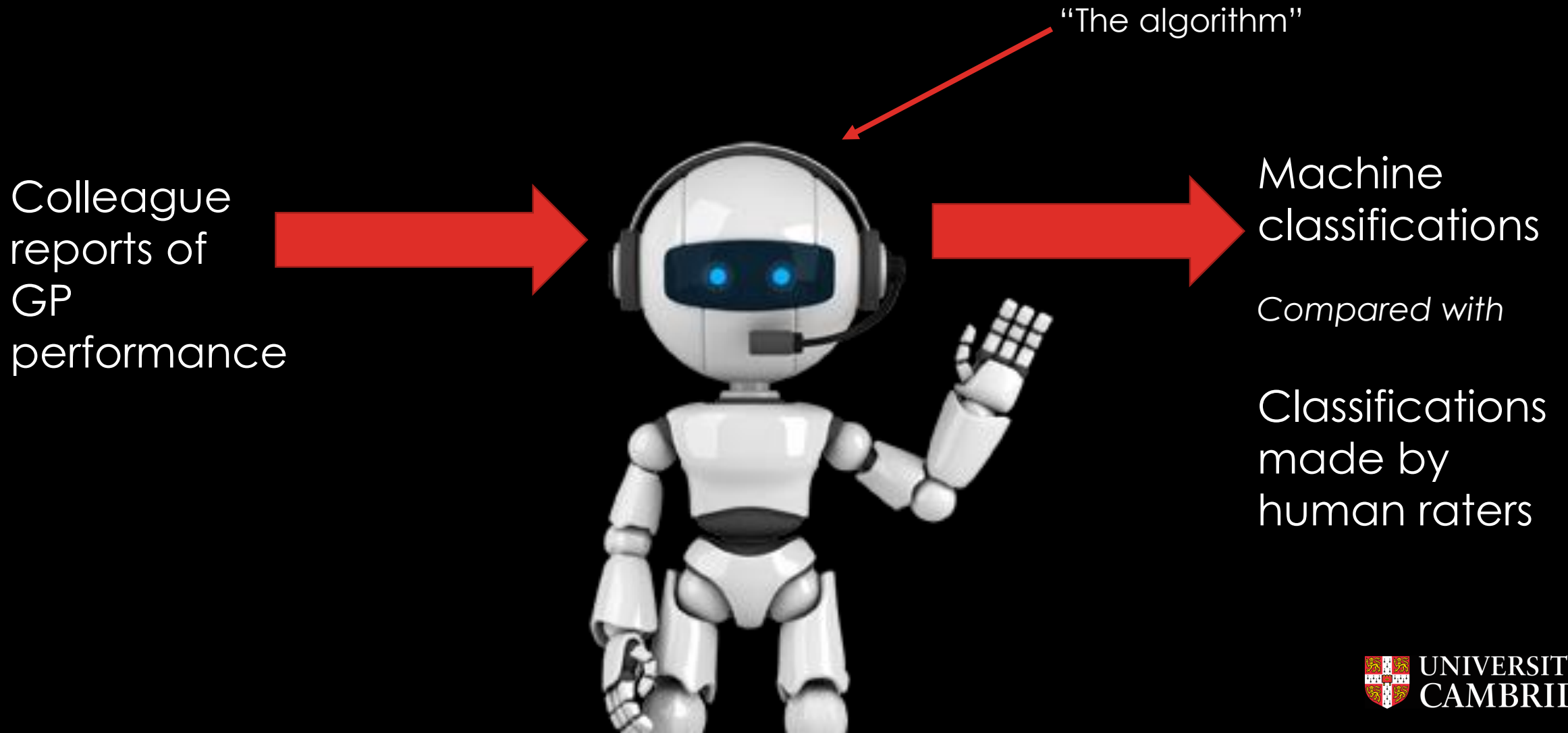
TRAINING



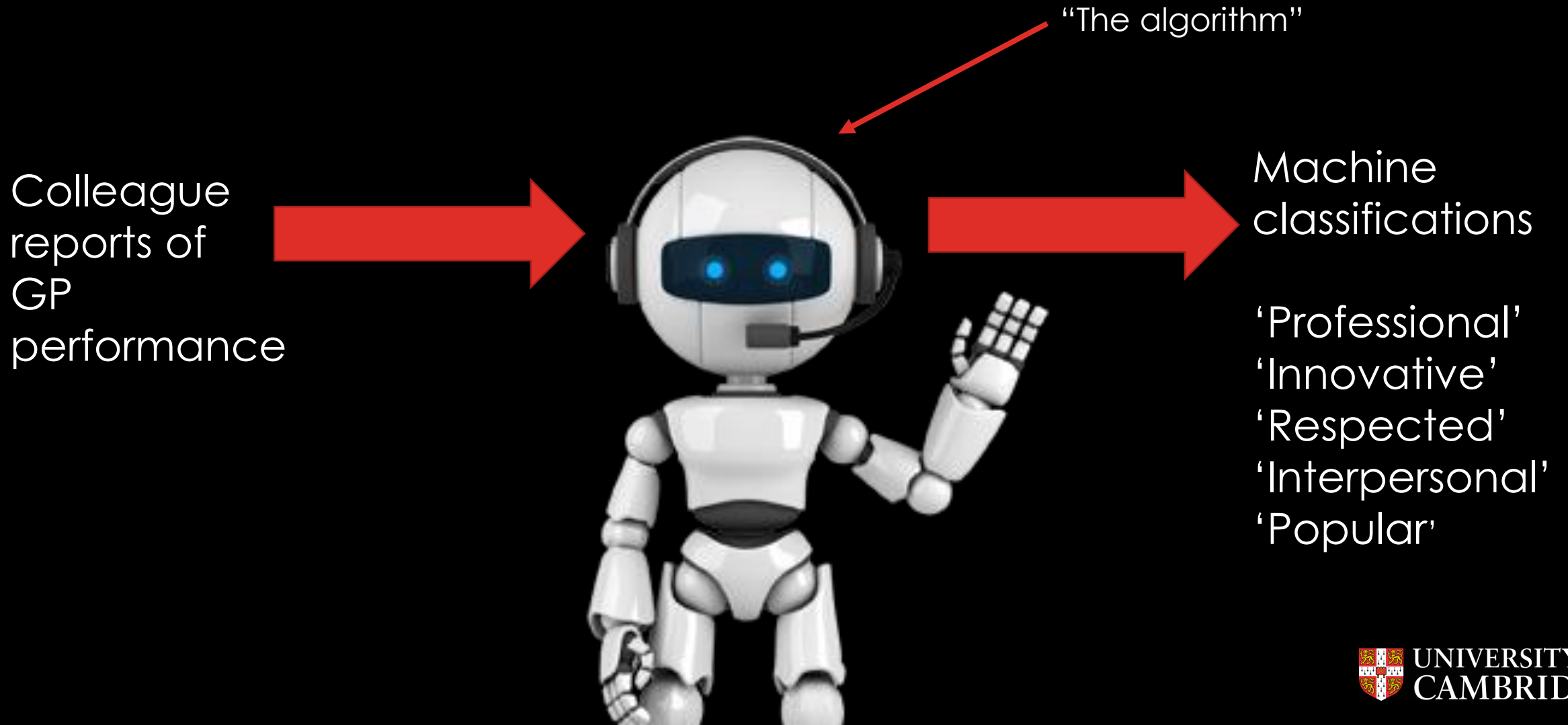
VALIDATING



VALIDATING



DEPLOYMENT



COMMENTS

Theme	Comment
Innovator	<p>"It is clear from the advice he gives that he is aware of [the] current good practice, is highly motivated, very practical and very much a team player. His advice, when working with consultant colleagues was respected, and he recognised where practice/primary care limitations were and yet looked for opportunities for change and improvement."</p>
	<p>"She has an admirable level of commitment and enthusiasm for her patients and her work. She has been instrumental in promoting change and improvement in her department. She is a great asset to the department and the hospital."</p>
Interpersonal	<p>"She is a very good, committed colleague always keen to improve, very liked by her patients and highly valued by all who work with her."</p>
Popular	<p>"Very approachable and professional."</p>
	<p>"Excellent well liked and easy working colleague."</p>
Professional	<p>"Very popular doctor. Works to high standards."</p>
	<p>"I find this doctor to be very efficient, caring, honest and very professional."</p>
Respected	<p>"I find that he very easy and helpful to work with, he always has time for patients and staff."</p>
	<p>"A first class colleague."</p> <p>"Pleasant and valued colleague."</p>

INTER-RATER AGREEMENT

<u>Code</u>	<u>Agreement (Kappa)</u>
Innovator	.98
Interpersonal	.80
Popular	.97
Professional	.82
Respected	.87

PREDICTING PERFORMANCE

T-tests comparing doctors with a rating in a category vs those with no ratings

<u>Code</u>	<u>Performance</u>		
	<u>N</u>	<u>T-test</u>	<u>Sig.</u>
Innovator	48	.00	.99
Interpersonal	435	1.98	.04
Popular	107	-.88	.38
Professional	643	2.51	.01
Respected	264	3.75	<.001
Any category	1069	.77	.001

PREDICTING PERFORMANCE

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MACHINE LEARNING

- Machine learning algorithms can classify open-text reports of doctors' performance with **human-level** accuracy
- Machine-made classifications can signal **significant** differences in doctors' performance

SUMMARY SO FAR

- Computer adaptive testing is an acceptable way to make assessments shorter and more reliable
- Machine learning may be used to make sense of open-text data in questionnaires

The problem



Concerto

Open source tool for the development of online assessments utilising computer adaptive testing and tailored feedback



The solution



Research, tools, training, and support

People



Prof John Rust
Director
Psychometrics Centre



Dr David Stillwell
Deputy Director
Psychometrics Centre



Dr Michal Kosinski
Computational Psychology
Stanford University



Przemyslaw Lis
Software Engineer
University of Cambridge



Dr Tomoya Okubo
Psychometrician
NCUEE Tokyo



Dr Chris Gibbons
Director of eHealth Assessment
Psychometrics Centre

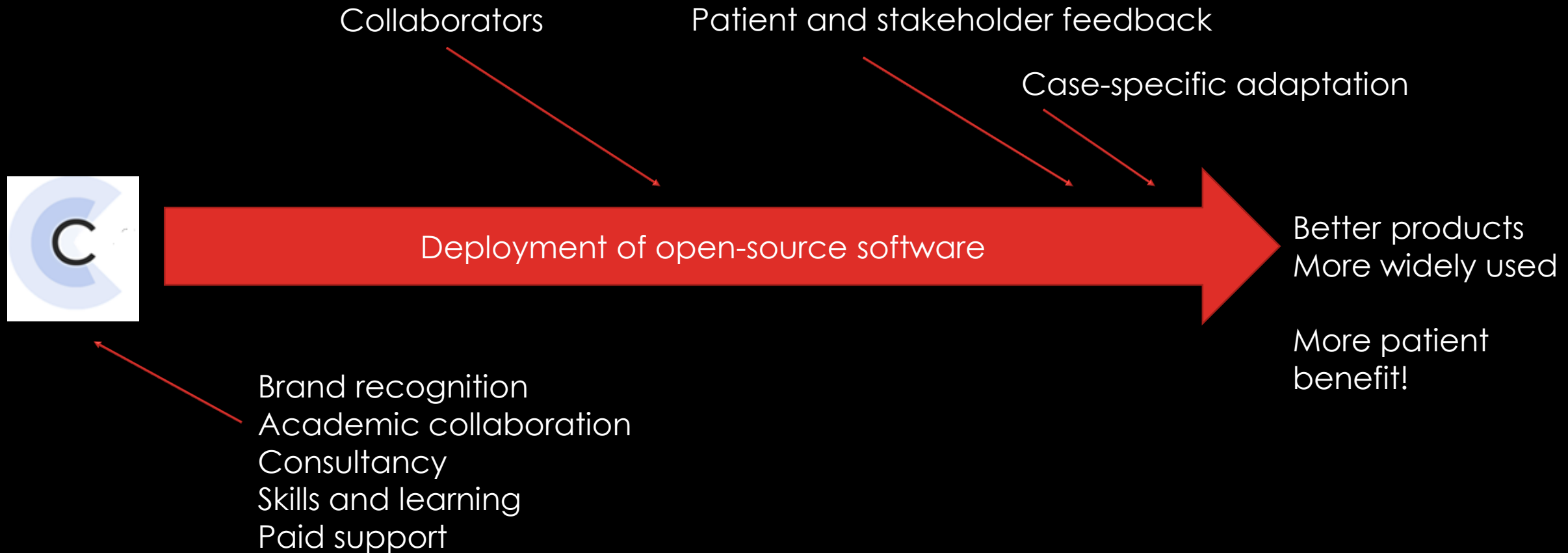
Vision

- Adaptive and electronic measurement should be widely available to non-experts
- It should always be free and open-source
- Assessment experience should be made as enjoyable as possible
- Data security protocol should be dictated by the test-developer (**your** servers, not ours)
- Where expertise is needed, it should be readily available as consultancy, training, or free guides online

open source_



open source_



In practice



Flexibility

- Item type (text, pictures, sounds, movies)
- Test type (Computer adaptive test, short-form questionnaire)
- Scoring (IRT scoring or sum scores)
- Feedback (patient/clinician/graphs/text/links/e-mail)
- Longitudinal assessment
- 'Log on' details
- Facebook connection
- Data security
- Data storage

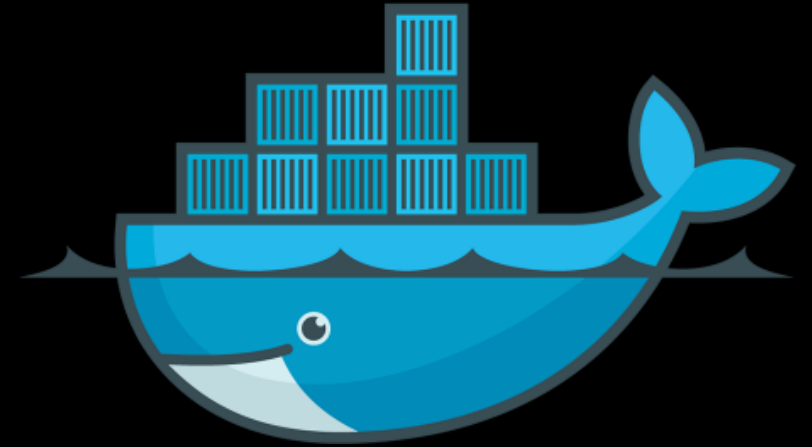


Concerto



Installable on different platforms

- Cloud servers (free trial on Amazon)
 - Local server
 - (at your University/hospital)
 - Docker
 - Windows 10 tablet
 - Linux Installation
-
- User-defined data security
 - Flexible speed and data-storage for scalable tests (up to millions of users)





BONUS STUDY

PREDICTING PERSONALITY FROM DIGITAL FOOTPRINT

MYPERSONALITY

- Example of Big Data analytics in social science, using personality questionnaires
- Big Data widely used by Big Business
- Google, Facebook, Apple, Amazon, Barclays...
- “Inevitable application in health care...” (Murdoch & Detsky, 2013, JAMA)
- What might big data analytics using questionnaires or PROMs look like?

MYPERSONALITY

Psychometric tests (BIG-5) hosted on Facebook

Developed and managed by psychometrics centre members

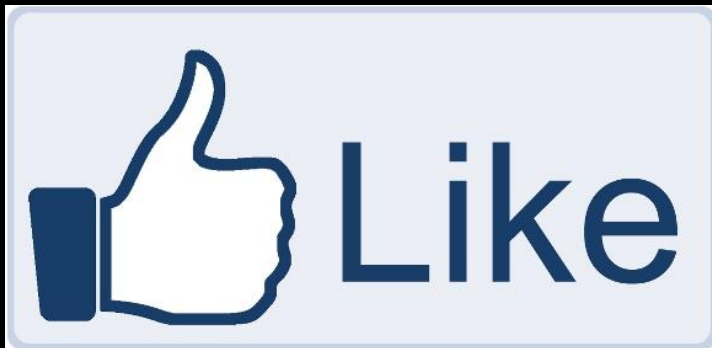
More than 3,100,000 people have completed the Big-5 questionnaire

Large proportion of users shared their Facebook 'Like' information

MYPersonality

Question:

Can you build a predictive model of a psychometric test (BIG-5) using Facebook 'likes'?



LIKES AND ITEMS

University of Birmingham

Chris Home

UNIVERSITY OF BIRMINGHAM

UNDERGRADUATE OPEN DAYS
#hellobrum

University of Birmingham ✓
@unibirmingham

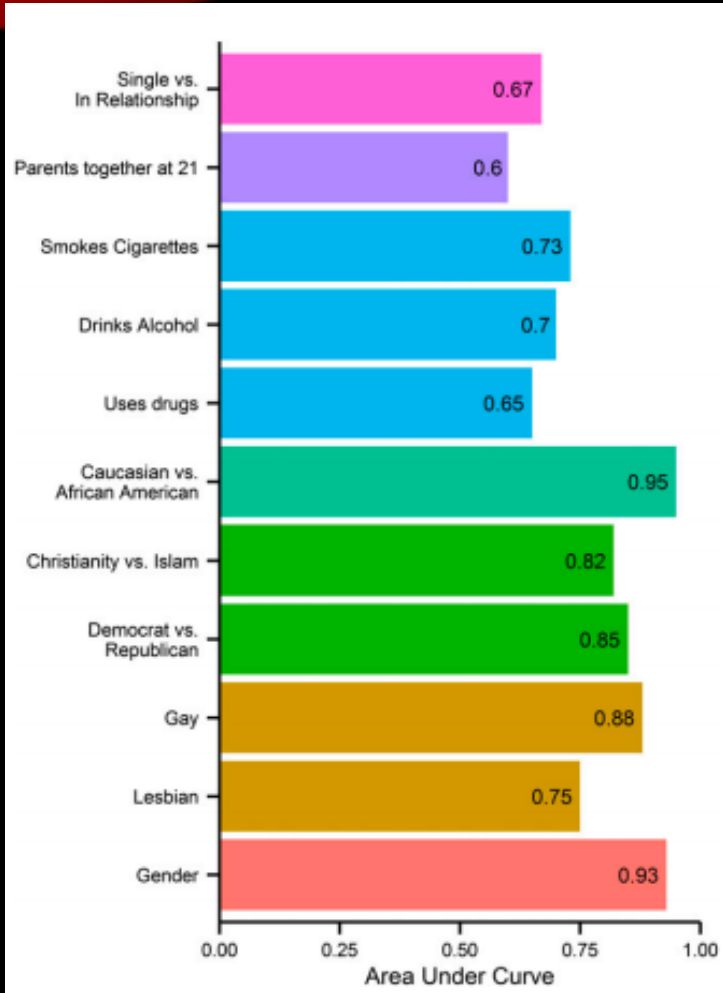
Home
About
Photos

Liked Message More

Send Message

Do you like University of Birmingham? Yes | No

MY PERSONALITY



Private traits and attributes are predictable from digital records of human behavior

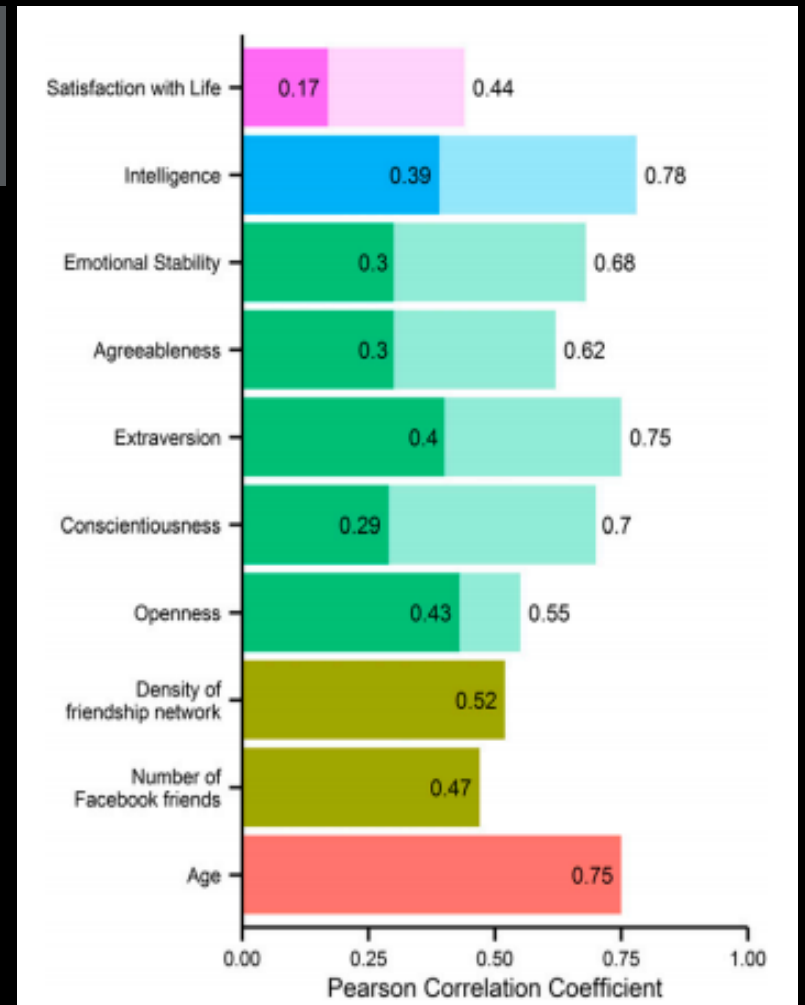
Michal Kosinski^{a,1}, David Stillwell^a, and Thore Graepel^b

^aFree School Lane, The Psychometrics Centre, University of Cambridge, Cambridge CB2 3RQ United Kingdom; and ^bMicrosoft Research, Cambridge CB1 2FB, United Kingdom

Edited by Kenneth Wächter, University of California, Berkeley, CA, and approved February 12, 2013 (received for review October 29, 2012)

We show that easily accessible digital records of behavior, Facebook Likes, can be used to automatically and accurately predict a range of browsing logs (11–15). Similarly, it has been shown that personality can be predicted based on the contents of personal Web sites (16).

PNAS



MY PERSONALITY

PNAS

Private traits and attributes are predictable from digital records of human behavior

Michal Kosinski^{a,1}, David Stillwell^a, and Thore Graepel^b

^aFree School Lane, The Psychometrics Centre, University of Cambridge, Cambridge CB2 3RQ United Kingdom; and ^bMicrosoft Research, Cambridge CB1 2FB, United Kingdom

Edited by Kenneth Wächter, University of California, Berkeley, CA, and approved February 12, 2013 (received for review October 29, 2012)

We show that easily accessible digital records of behavior, Facebook Likes, can be used to automatically and accurately predict a range of browsing logs (11–15). Similarly, it has been shown that personality can be predicted based on the contents of personal Web sites (16).

Trait		Selected most predictive Likes		
IQ	<i>High</i>	The Godfather	Jason Aldean	<i>Low</i>
		Mozart	Tyler Perry	
		Thunderstorms	Sephora	
		The Colbert Report	Chiq	
		Morgan Freemans Voice	Bret Michaels	
		The Daily Show	Clark Griswold	
		Lord Of The Rings	Bebe	
		To Kill A Mockingbird	I Love Being A Mom	
		Science	Harley Davidson	
		Curly Fries	Lady Antebellum	
Satisfaction With Life	<i>Satisfied</i>	Sarah Palin	Hawthorne Heights	<i>Dissatisfied</i>
		Glenn Beck	Kickass	
		Proud To Be Christian	Atreyu (Metal Band)	
		Indiana Jones	Lamb Of God	
		Swimming	Gorillaz	
		Jesus Christ	Science	
		Bible	Quote Portal	
		Jesus	Stewie Griffin	
		Being Conservative	Killswitch Engage	
		Pride And Prejudice	Ipod	



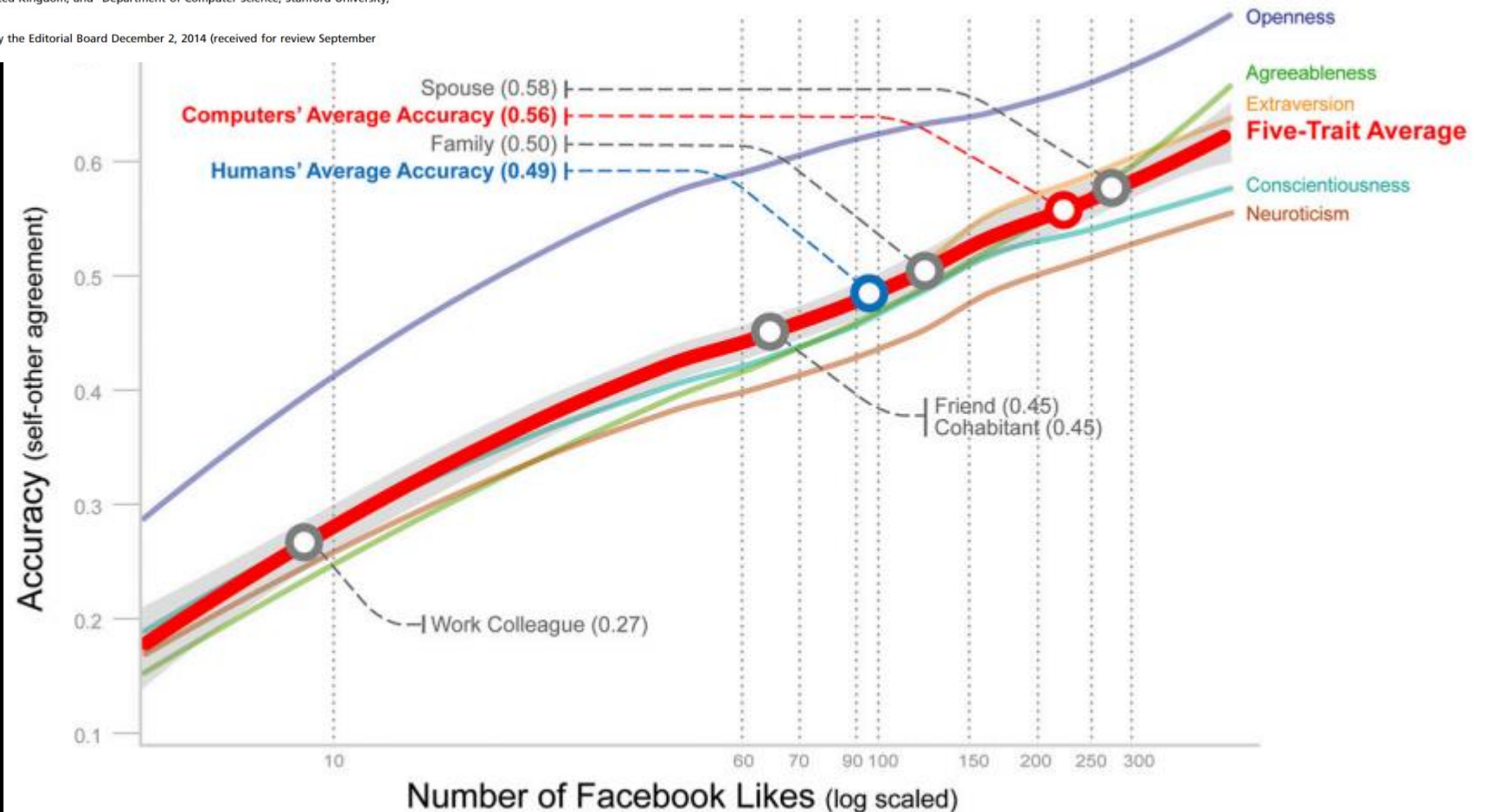
Computer-based personality judgments are more accurate than those made by humans

Wu Youyou^{a,1,2}, Michal Kosinski^{b,1}, and David Stillwell^a

^aDepartment of Psychology, University of Cambridge, Cambridge CB2 3EB, United Kingdom; and ^bDepartment of Computer Science, Stanford University, Stanford, CA 94305

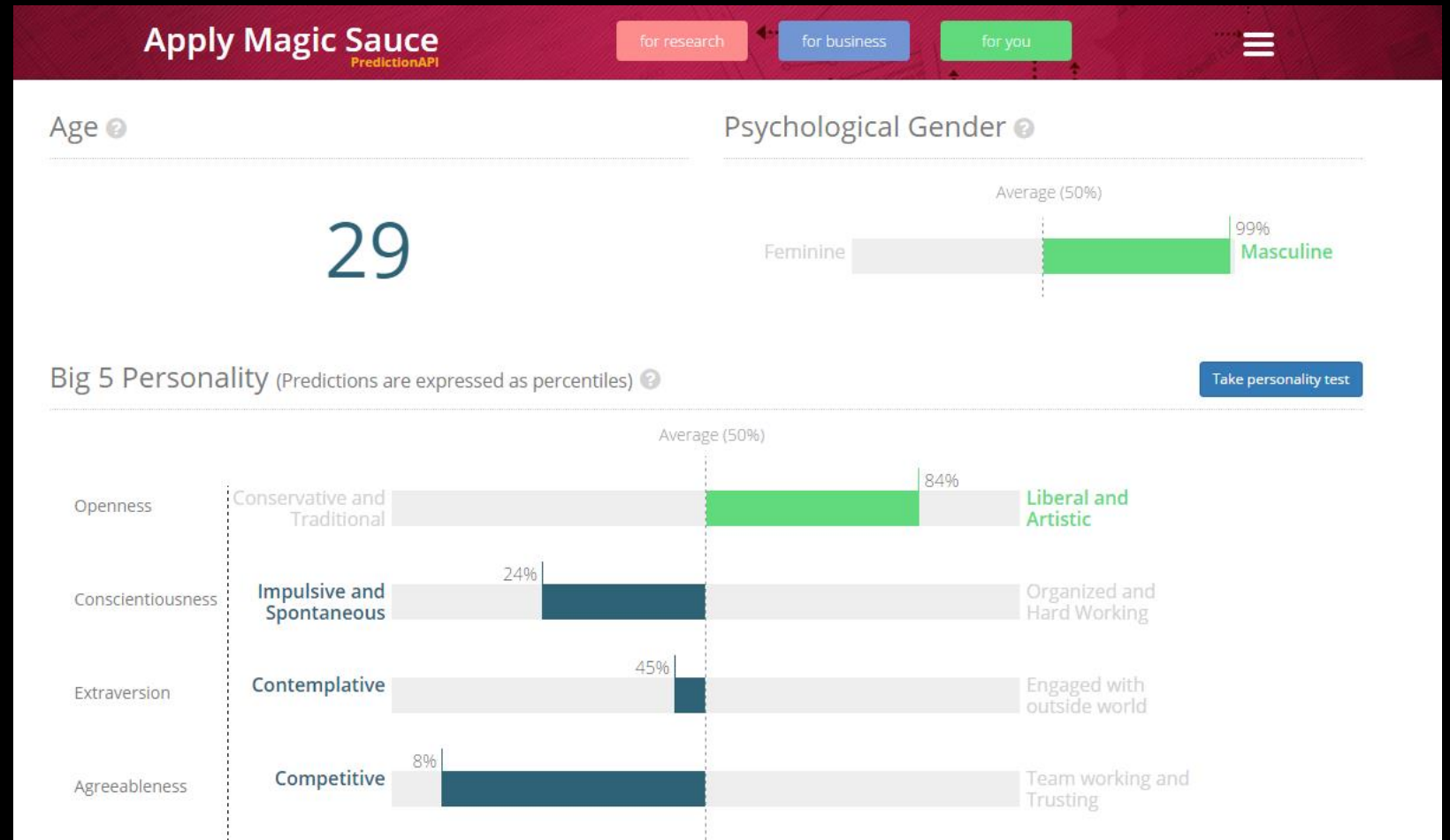
Edited by David Funder, University of California, Riverside, CA, and accepted by the Editorial Board December 2, 2014 (received for review September 28, 2014)

DOES IT WORK?



APPLYMAGICSAUCE.COM

- API which allows companies to make personality predictions at scale
- Hilton, Wrigleys, Barclays
- Using psychological information to tailor information improves conversion rate and satisfaction (Matz, 2015)



LEARNING POINTS

On feedback

- 3,100,000 people completed a personality test that had up to 336 items!
- No monetary incentive - just feedback
- People want to learn about themselves!
- Research planned – how does feedback affect response rates, reliability etc

On data sharing and collaborative / open science

- All these data (apart from Likes) are available for free to anyone (since 2012)
- No regrets!

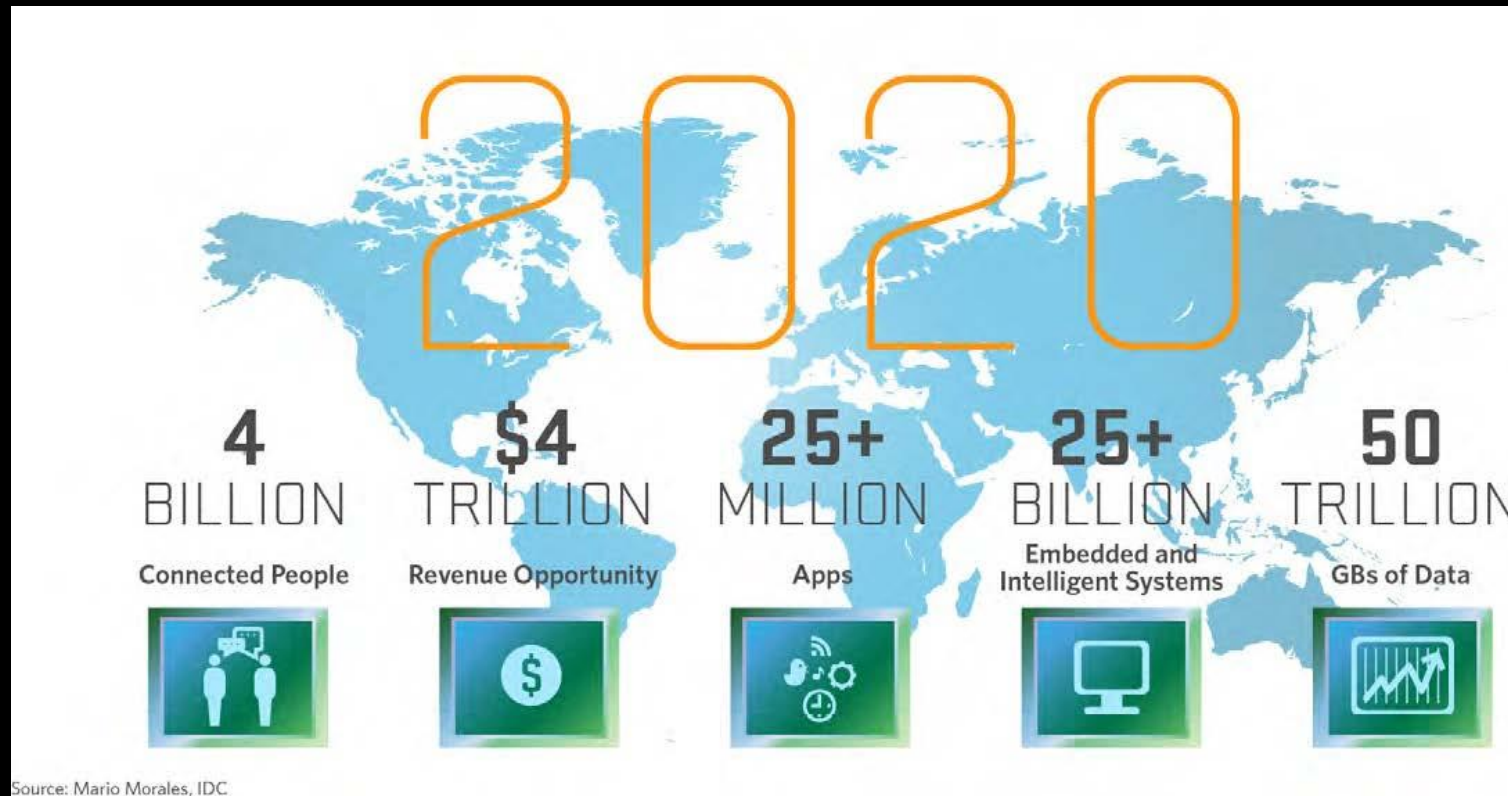
WHY MIGHT THIS 'INFERENTIAL' APPROACH BE USEFUL FOR HEALTH SCIENCES

- Data exist on a massive scale (and increasingly so)
- 'Learning data cities' are being planned (cf. Buchan)
- May be *less biased* than questionnaire items
- Real-time monitoring/assessment/feedback
- No measurement latency (can be done anywhere)

WHAT MIGHT WE USE FOR 'INFERENTIAL PSYCHOMETRICS'?

- Facebook likes
- Twitter 'follows'
- Twitter updates
- Images (image recognition – e.g., Facebook profile photos)
- Instagram
- 'Wearables'
- Geo-location data
- Movement data
- Phone activity
- Interactions with friends (proximity of two devices)

INTERNET OF THINGS



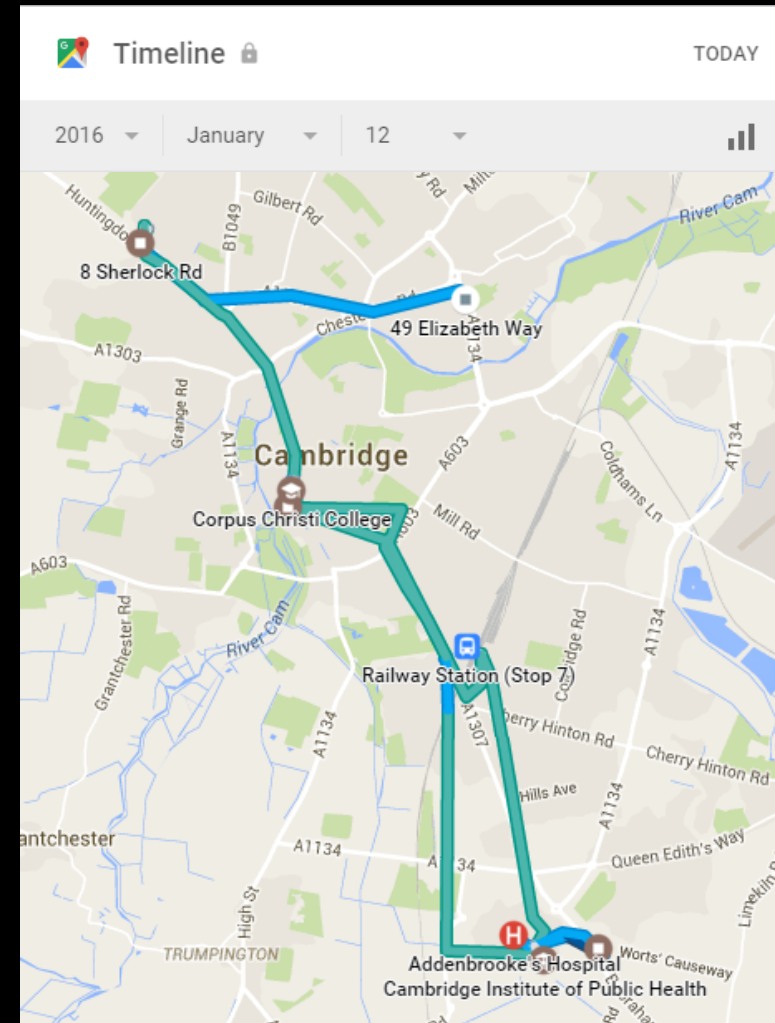
SF-36 & GOOGLE DATA?

The following questions are about activities you might do during a typical day. In the past 1-week does your health limit you in these activities? If so, how much?

(Please circle one number on each line)

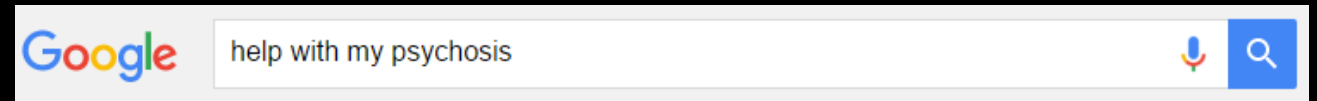
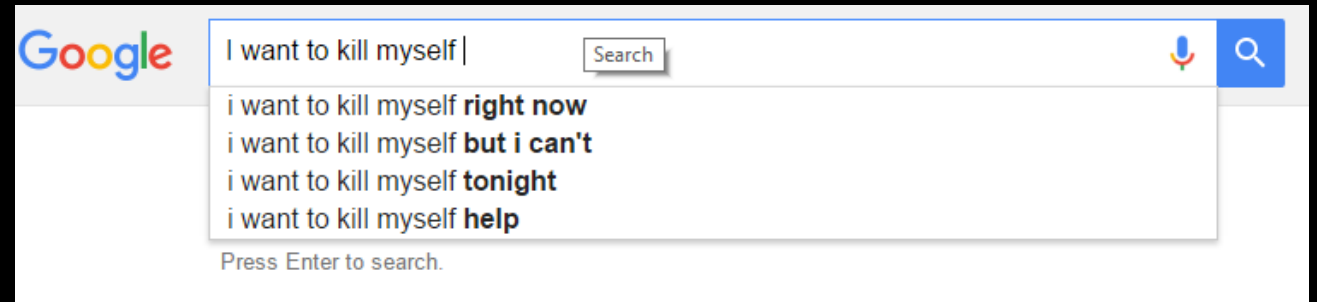
ACTIVITIES	Yes Limited A lot	Yes Limited A little	No, Not Limited At All
3a: Vigorous activities, such as running, lifting heavy Objects, participating in strenuous sports	1	2	3
3b: Moderate activities, such as moving a table, pushing a vacuum cleaner, bowling, or playing golf	1	2	3
3c: Lifting or carrying groceries	1	2	3
3d: Climbing several flights of stairs	1	2	3
3e: Climbing one flight of stairs	1	2	3
3f: Bending, kneeling, or stooping	1	2	3
3g: Walking more than one kilometre	1	2	3
3h: Walking half a kilometre	1	2	3
3i: Walking 100 metres	1	2	3
3g ww: Wheeling more than one kilometre	1	2	3
3h ww: Wheeling half a kilometre	1	2	3
3i ww: Wheeling 100 metres	1	2	3
3j: Bathing or dressing yourself	1	2	3

^aModified from SF-36¹: Items 3 (a to j) are the original SF-36 questions, while 3g ww to 3i ww (shaded area) comprise the supplementary SF-36ww modification.



MENTAL HEALTH AND GOOGLE DATA?

- Do you have thoughts of suicide?
- Have you been experiencing delusions?



GOOGLE AND THOMAS INSEL

Former head of NIMH in the USA

Moved to Google in late 2015.

“Technology can have greater impact on mental healthcare than on the care for heart disease, diabetes, cancer or other diseases...

It could transform this area in the next five years.” (Insel, 2015)



IS USING DATA LIKE THIS
A BIT CREEPY*?

*OFF-PUTTING TO PATIENTS

PredictiveDataProject.com

Assessment

- 27 Yes/No Questions on aspects of Big Data
- Measuring attitudes towards Personalisation, Ethics, Wearable Tech, Internet of Things, Finance, Policy-Making, Cloud Storage, etc.
- 20-item BIG5 Personality test at the end

Feedback

- Participants given **real-time** feedback on the similarity of their answers to all previous participants



Thank you!

Thank you for participating in the Predictive Big Data test.

Your similarity to the general population:

60%

Calculated from your answers to all previous questions.

Your personality

Openness:

50%

Conscientiousness:

85%

Extraversion:

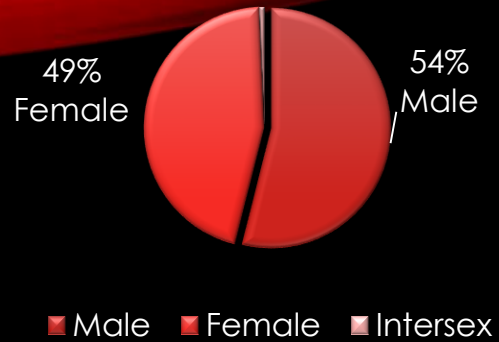
88%

Agreeableness:

16%

Neuroticism:

48%



Average age

30

Sample (by Oct '15)

No. of test sessions in total (Sept)

33,937

Answered At Least One Question

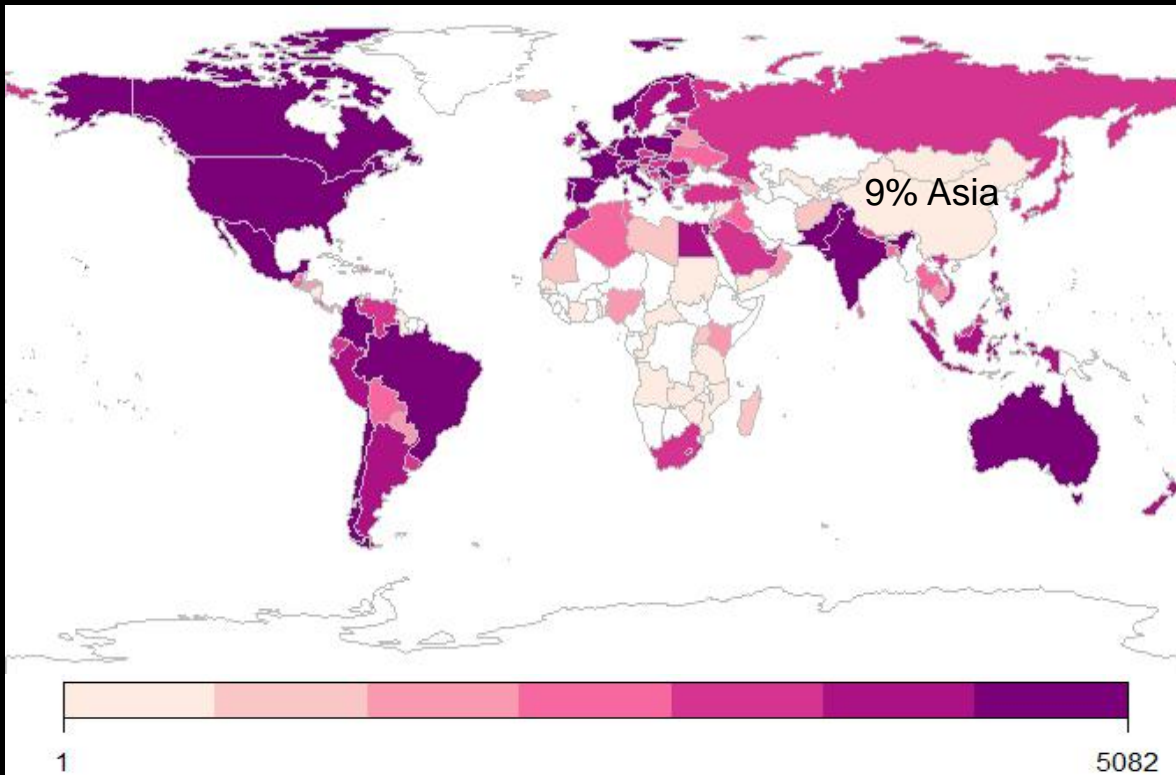
19,126

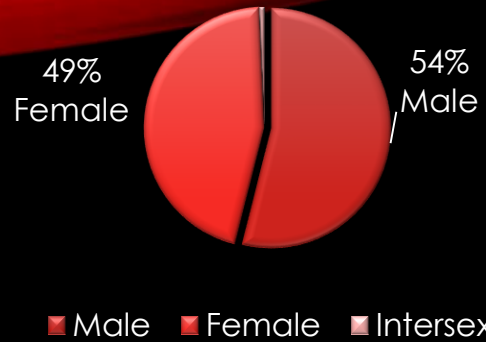
All Big Data Questions

10,411

All Personality Questions

8,871





Average age

30

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Answered At Least One Question

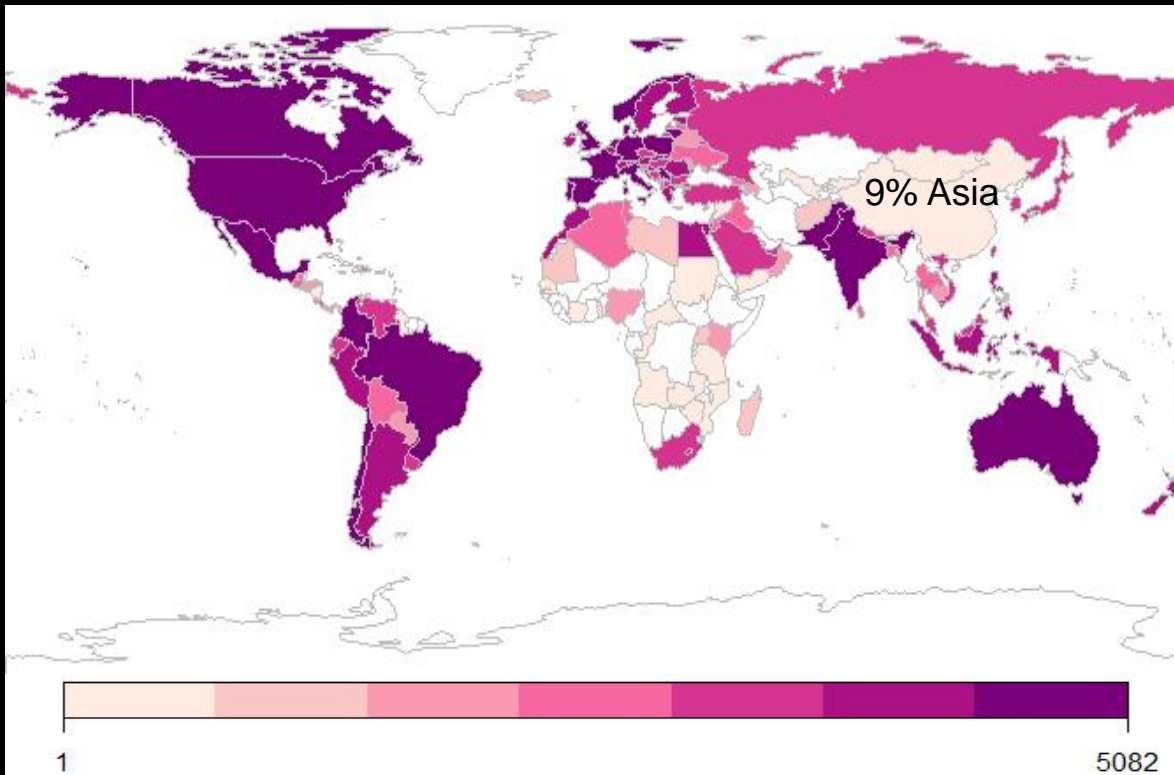
19,126

All Big Data Questions

10,411

All Personality Questions

8,871



Results: Audit of Global Public Opinion

Are those with access to your personal data able to accurately predict your future behaviour? (n=19,100)

62% NO

38% YES



Should predictive technologies be used to improve the quality of healthcare, for example by helping doctors to recommend personalised nutrition and exercise plans? (n=14,187)

16% NO

84% YES



Should predictive technologies be used assess your eligibility for a mortgage? (n=13,725)

62% NO

38% YES



Do you think your organisation ought to invest in predictive technologies? (n=2,830 CMOs)

62% NO

73% YES



Is it important for you to understand the psychological attributes of your customers ? (n=2,489 CMOs)

8% NO

92% YES



PATIENTS AND SOCIAL MEDIA

71% of patients were happy to share social media information with their doctor

Downloaded from <http://qualitysafety.bmj.com/> on February 9, 2016 - Published by group.bmj.com
BMJ Quality & Safety Online First, published on 13 October 2015 as 10.1136/bmjqs-2015-004489

ORIGINAL RESEARCH

Linking social media and medical record data: a study of adults presenting to an academic, urban emergency department

Kevin A Padrez,^{1,2} Lyle Ungar,^{2,3,4} Hansen Andrew Schwartz,^{2,3,4}
Robert J Smith,^{2,4,5} Shawndra Hill,^{2,6} Tadas Antanavicius,^{2,4}
Dana M Brown,^{2,5} Patrick Crutchley,^{2,4} David A Asch,^{2,7}
Raina M Merchant^{2,5}

CONCLUSION

- Modern psychometrics offers many opportunities for improved research and clinical practice
- Computer adaptive tests work well in the 'real world'
- Predictions are engaging
- myPersonality showed what will/could be possible for 'inferential psychometrics' in health research and the power of feedback
- The future is exceptionally exciting in this field!

THANK YOU!

cg598@cam.ac.uk

concertoplatform.com
drchrisgibbons.com

@DrCGibbons
@concerto_cam
@campsych