

Should behavioural and psychological factors be used to assess mortality risk?

Peter Banthorpe, Global Head of R&D, RGA Chris Falkous, Biometric Research Actuary, RGA

What we're not talking about today



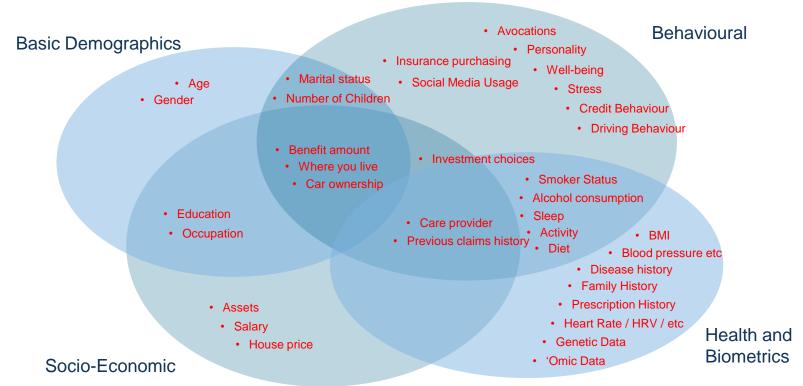




Behavioural Design



Many things influence or help predict longevity





Why talk about this now?

This is happening now



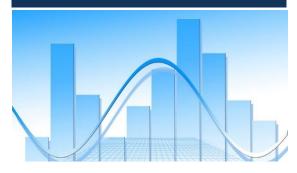
TransUnion / RGA Mortality Study

Credit-Based Solutions for Life Insurance

Scott Rushing FSA, MAAA Vice President & Actuary, Global Research and Development RGA Reinsurance Company

Vice President of Sales – Insurance TransUnion

Understand Causal Paths



Value from Social Media? Welcome to Facebook - L × Mattheward Media? Facebook https://www.facebook.com Facebook helps you connect and share with the people in your life.

Wearables



New Competitors



Should we do it?





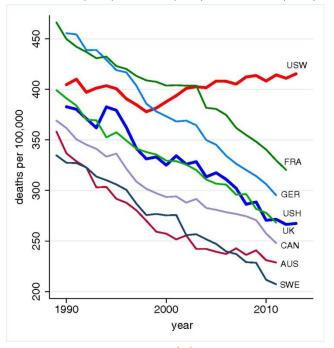
Reductions in Wellbeing associated with Adverse Population Mortality Trends

Rising morbidity and mortality in midlife among white non-Hispanic Americans in the 21st century

Anne Case¹ and Angus Deaton¹

Self-reported declines in health, mental health, and ability to conduct activities of daily living, and increases in chronic pain and inability to work, as well as clinically measured deteriorations in liver function, all point to growing distress in this population.

All-cause mortality, ages 45–54 for US White non-Hispanics (USW), US Hispanics (USH), and six comparison countries: France (FRA), Germany (GER), the United Kingdom (UK), Canada (CAN), Australia (AUS), and Sweden (SWE).







Personality



Personality

- What is personality?
- Differences in personality
 - Genetics
 - Environment (shared vs non-shared)
 - Interaction (gene expression affected by environment)
- Increasing stability over time
- It is possible to change personality
 - Trauma, parasitic infection, pharmaceutical effects
 - Life events (marriage, work, bereavement)



Assessing personality

What's Your Personality Type?

Use the questions on the outside of the chart to determine the four letters of your Myers-Briggs type. For each pair of letters, choose the side that seems most natural to you, even if you don't agree with every description.

1. Are you outwardly or inwardly focused? If you:

- Could be described as talkative, outgoing
- · Like to be in a fast-paced environment
- · Tend to work out ideas with others, think out loud
- · Enjoy being the center of attention

then you prefer

Extraversion

Could be described as reserved, private

- Prefer a slower pace with time for contemplation Tend to think things
- through inside your head Would rather observe than be the center of attention

then you prefer

Introversion

ISTP Action-oriented, logical, analytical, spontaneous, reserved, independent Enjoy adventure, skilled

at understanding how

mechanical things work

ESTP

Responsible, sincere, analytical, reserved,

ISFP Gentle, sensitive, nurturing, helpful, lexible, realistic. Seek to create a personal beautiful and practical.

spontaneous, optimistic, supportive, playful. Value inspiration, enjoy starting new projects

INFJ

Idealistic, organized, insightful, dependable

compassionate, gentle Seek harmony and

cooperation, enjoy intellectual stimulation

INFP

Sensitive, creative,

idealistic, perceptive, caring, loyal. Value inne

and possibilities.

ENFJ

LTNI Make decisions in an ovative, independer strategic, logical, reserved, insightful.

· Value justice, fairness

- · Enjoy finding the flaws in an argument
- then you prefer

3. How do you prefer to make decisions? If you: Base your decisions on

- impersonal way, using logical reasoning
- Could be described as reasonable, level-headed

Thinking

- personal values and how your actions affect others
- Value harmony, forgiveness · Like to please others and
- point out the best in people Could be described as warm, empathetic

then you prefer

Feeling

2. How do you prefer to take in information? If you:

- · Focus on the reality of how things are
- · Pay attention to concrete facts and details
- · Prefer ideas that have practical applications
- · Like to describe things in a specific, literal way

then you prefer

Sensing

- · Imagine the possibilities of
- Notice the big picture, see how everything connects Enjoy ideas and concepts
- figurative, poetic way

Intuition

Outgoing, realistic, how things could be action-oriented, curious versatile, spontaneou: Pragmatic problem solvers and skillful negotiators.

for their own sake Like to describe things in a then you prefer

ESTJ Efficient, outgoing, analytical, systematic ke to run the show an

ESFP Playful, enthusiastic, friendly, spontaneous strong common sense enjoy helping people in

ESFJ Friendly, outgoing, to be helpful and please others, enjoy being active and productive

Caring, enthusiastic idealistic, organized, Skilled communicators who value connectior

inquisitive, versatile. Enjoy new ideas and challenges, value

Driven by their own

original ideas to achiev improvements.

INTP

Intellectual, logical,

precise, reserved,

flexible, imaginative.

Original thinkers who

enjoy speculation and creative problem solving.

Strategic, logical, Effective organizers o people and long-rang

4. How do you prefer to live your outer life? If you:

flexible

- · Prefer to have matters settled
- Think rules and deadlines should be respected
- Prefer to have detailed. step-by-step instructions
- · Make plans, want to know what you're getting into

then you prefer

Judging

- Prefer to leave your options
- See rules and deadlines as
- Like to improvise and make things up as you go
- Are spontaneous, enjoy surprises and new situations

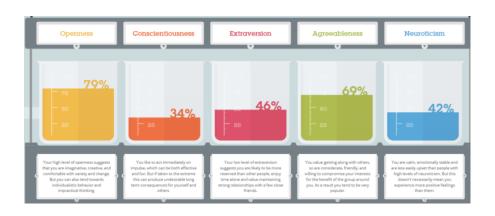
then you prefer

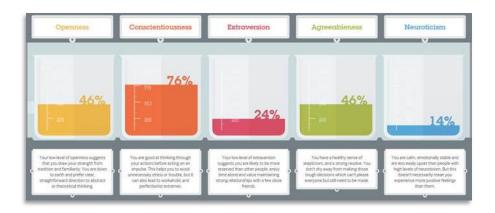
Perceiving



Source: Jake Beech - Own work, CC BY-SA 3.0, https://commons.wikimedia.org/w/index.php?curid=30859659

The Big 5 Personality Inventory



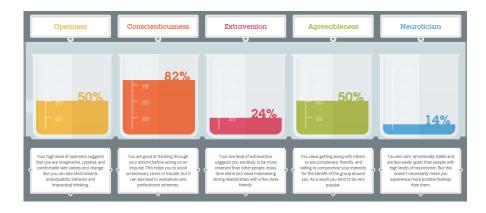


"MasterCard, in their 2014
Advisors report, found that
candidates accepted via
VisualDNA's personality data
were 23% less likely to default on
their borrowing..."



The Big 5 Personality Inventory







The Big 5 and mortality



- Earlier studies found it generally unrelated to mortality
- More recent investigations have found a protective effect



Few studies have found an association



- Tendency to be hostile and aggressive predicts mortality
- Otherwise, studies generally found no association



- · Mixed results
- Some studies have found a positive association, some negative



Conscientiousness and mortality

- Relevant to many Biopsychosocial processes
 - Engaging in healthy behaviour
 - Choose healthier environments
 - Select and maintain healthier friendships
 - Better educations, more successful careers, higher income
 - Moderates detrimental effect of unhealthy stressors
 - Gene x environment interactions
 - Compound effect of individual, tiny differences



Conscientiousness in childhood

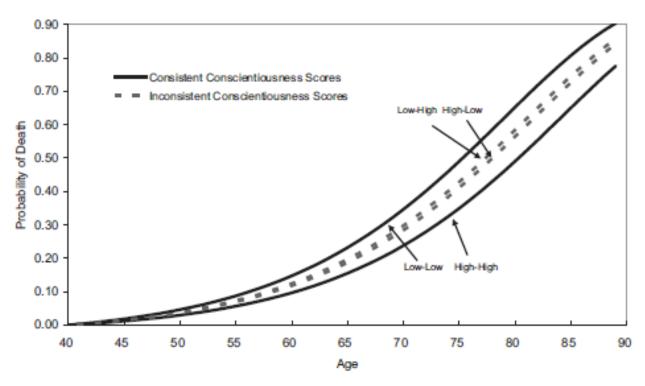


Figure 3. Estimated probability of a Terman participant's dying (1950-2000) by a given age, by conscientiousness at two time points.



Conscientiousness in adulthood

High conscientiousness related to lower risk of death

When controlling for	Mortality risk between highest and lowest levels of conscientiousness	Other significant indicators of lower risk of death (p<0.01)
Nothing	≈ 3.2x	High neuroticism Low agreeableness High extraversion
Age, sex, race, education, marital status	≈ 2.4x	
Age, sex, race, education, marital status, smoking, alcohol use, waist circumference	≈ 3.2x	



TRL Model Creation Process

Starting Data

- Built the model on 44 million lives and over 3 million deaths
- Started with over 800 variables offering features of individual's credit history

Variable Selection

- Selected variables that were:
 - Most predictive of the outcome
 - Stable over time
 - Non-gameable
 - Not too correlated with the other variables

Model Process

- Binary Logistic Regression
- Model validated internally using an additional 30 million lives

External Validation of Model

- Tested model using traditional mortality and lapse studies
- Used a random holdout dataset of another 18 million lives

TrueRisk[®] Life Score

TrueRisk[®] Life presented as a score from:



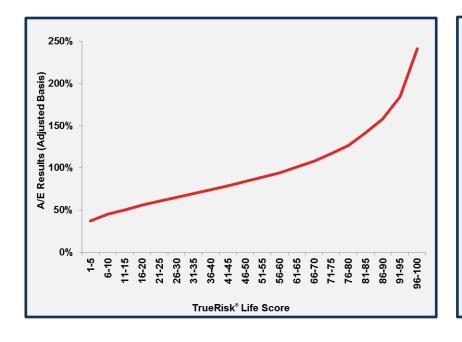
- Data comes from de-personalized 1998 credit archive (90% of the US population)
- Model calibrated to actual deaths occurring over a 12-year period

TransUnion & RGA built and tested TrueRisk® Life on 92 million individuals



Model Validation – Population Study

Overall Mortality



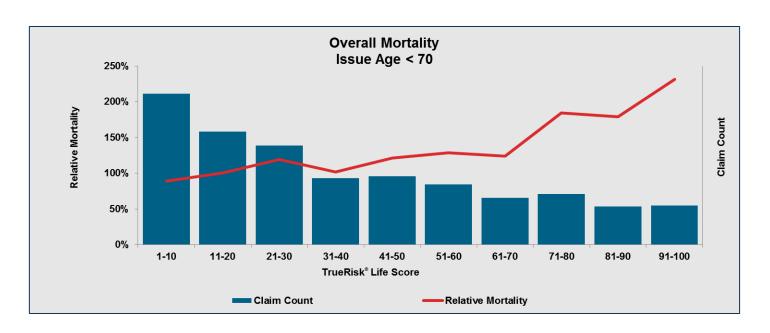
Details

- Mortality study performed on holdout sample of 18 million lives using a 1998 TransUnion archive and studying the lives during 1999-2010
- Score buckets are set to be uniform across the population
- Study shows 5 times segmentation (96-100 compared to 1-5)
- SSMDF used as source of deaths; used population mortality tables



TRL Model Validation – Insured Lives Study

Mortality of the 91-100 group is 2.6 times higher than the 1-10 group

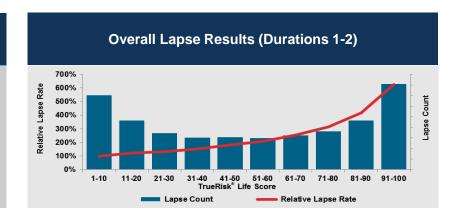




TRL Model Validation – Insured Lives Study

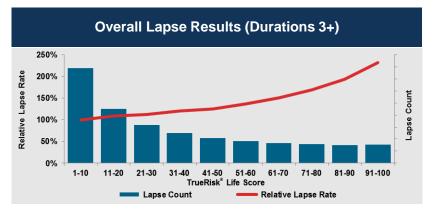
Details

- Term, UL and VUL
- Face amounts ≥ \$100k
- Issue Ages < 70



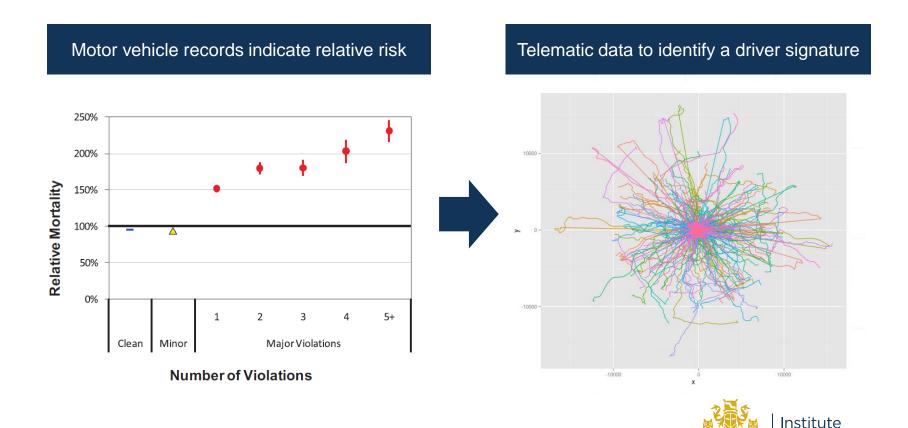
Results

- Lapse rate of 91-100 group is 6 times higher than 1-10 group in durations 1-2
- Continued segmentation seen in later durations, but less dramatic
- Similar results seen when looking at the curves by issue age band



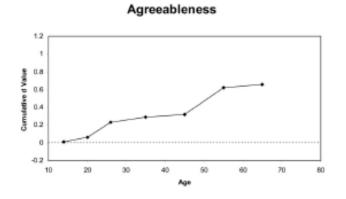


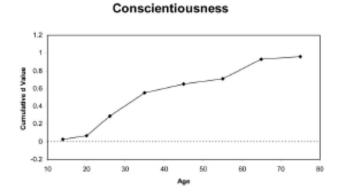
Driving habits – another behavioural risk classifier

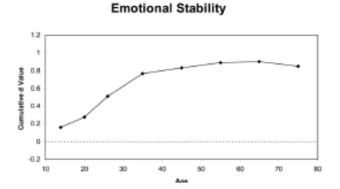


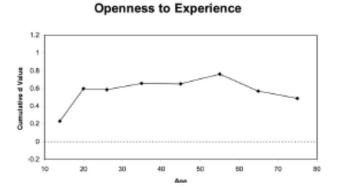
and Faculty of Actuaries

Personality changes over time





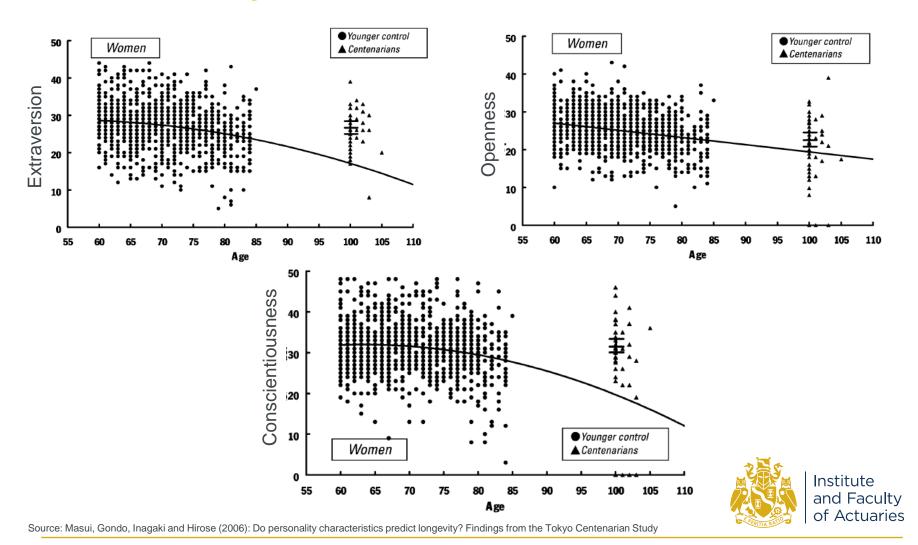




Source: Roberts, B. W., Walton, K. E., & Viechtbauer, W. (2006). Patterns of mean-level change in personality traits across the life course: A meta-analysis of longitudinal studies



Personality and the oldest old





Subjective well-being and stress

Subjective Well-Being

Evaluative (life satisfaction)

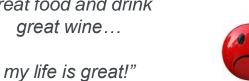
Hedonic (moods/emotions)

Eudemonic (purpose/meaning)

"I've got a great job, great family, great friends, I live in a great home, I eat great food and drink great wine...



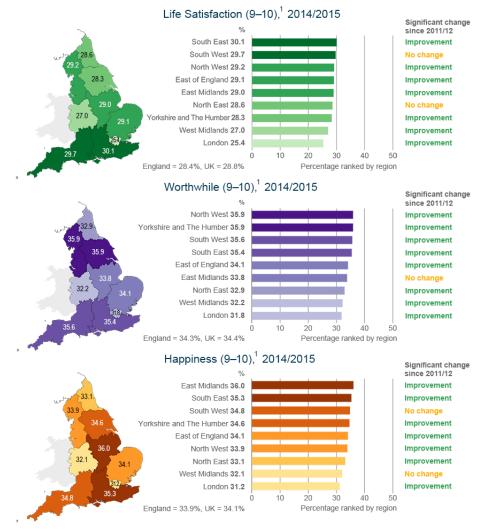








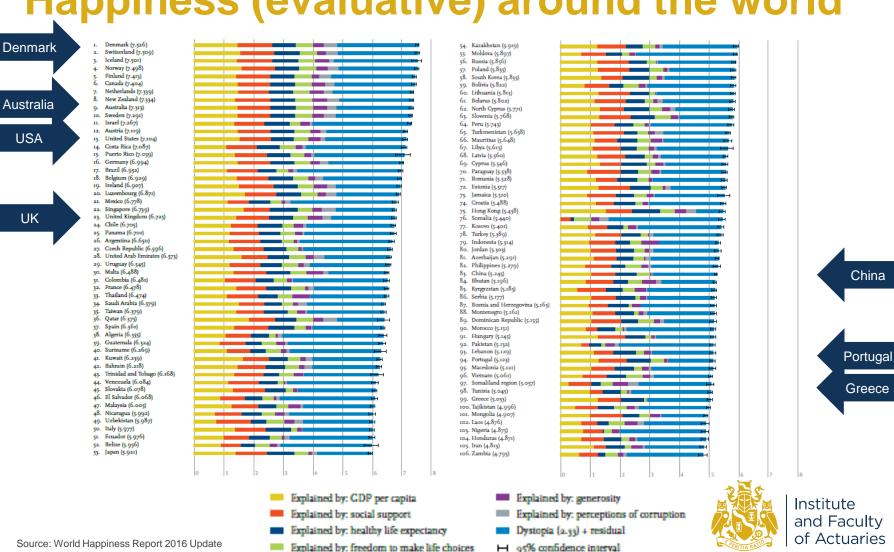
Subjective Well-Being in England





Source: ONS

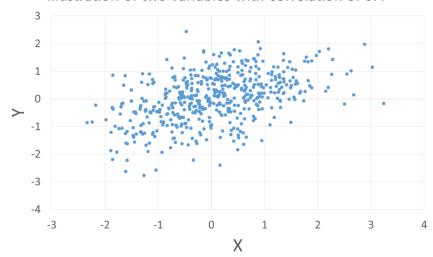
Happiness (evaluative) around the world



Personality and Subjective Well-Being

Personality trait	Correlation with					
	Positive affect	Negative affect	Life satisfaction			
Neuroticism	-0.41	0.66	-0.39			
Extraversion	0.45	-0.31	0.27			
Conscientiousness	0.42	-0.31	0.28			

Illustration of two variables with correlation of 0.4





Subjective Well-Being and mortality

- Diener and Chan (2011)
- "... the case that subjective well-being influences health and longevity in healthy populations is compelling..."





Subjective Well-Being and mortality

The impact of eudemonic wellbeing

When controlling for	Mortality risk of lowest quartile of eudemonic wellbeing to highest		
Age, gender	≈ 0.42		
Age, gender, demographic and health indicators	≈ 0.62		
Age, gender, demographic and health indicators, depression and health behaviours	≈ 0.70		



Source: Steptoe, Deaton and Stone (2015): Subjective wellbeing, health, and ageing.

What things can we measure?





Social media data

Well-being and problematic Facebook use

Seydi Ahmet Satici a,*, Recep Uysal b

^a Anadolu University, Educational Faculty, Department of Psychological Counseling and Guidance, Eskisehir, Turkey ^b Yildiz Technical University, Educational Faculty, Department of Psychological Counseling and Guidance, Istanbul, Turkey

Mining Facebook Data for Predictive Personality Modeling

Dejan Markovikj

Faculty of Computer Science and Engineering Ss. Cyril and Methodius University Rugier Boshkovikj 16 Skopje, Macedonia dejan@markovikj.com

Sonja Gievska

Faculty of Computer Science and Engineering Ss. Cyril and Methodius University Rugier Boshkovikj 16 Skopje, Macedonia sonja.gievska@finki.ukim.mk

Michal Kosinski

The Psychometrics Centre University of Cambridge Psychology, Free School Lane Cambridge CB2 3RQ United Kingdom mk583@cam.ac.uk

sinski David Stillwell

The Psychometrics Centre University of Cambridge Psychology, Free School Lane Cambridge CB2 3RQ United Kingdom ds617@cam.ac.uk

Predicting Personality from Twitter

Jennifer Golbeck*, Cristina Robles*, Michon Edmondson*, and Karen Turner*

*University of Maryland;{jgolbeck,crobles,michonk8,kturner}@umd.edu

What does your selfie say about you?

Lin Qiu a,*, Jiahui Lu a, Shanshan Yang a, Weina Qu b, Tingshao Zhu b

^a Division of Psychology, Nanyang Technological University, Singapore ^b Institute of Psychology, Chinese Academy of Sciences, China

Online networks and subjective well-being 1

Fabio Sabatini 2 *

Francesco Sarracino

Our Twitter Profiles, Our Selves: Predicting Personality with Twitter

Daniele Quercia§ Michal Kosinski† David Stillwell* Jon Crowcroft§

§The Computer Laboratory, University of Cambridge, UK
†The Psychometrics Centre, University of Cambridge, UK
*School of Psychology, University of Nottingham, UK

Characterizing Geographic Variation in Well-Being Using Tweets

H. Andrew Schwartz,* Johannes C. Eichstaedt,* Margaret L. Kern, Lukasz Dziurzynski, Megha Agrawal Gregory J. Park, Shrinidhi K. Lakshmikanth, Sneha Jha, Martin E. P. Seligman, and Lyle Ungar University of Pennsylvania

hansens@seas.upenn.edu, jeich@sas.upenn.edu

Richard E. Lucas Michigan State University

Personality and Patterns of Facebook Usage

Yoram Bachrach Microsoft Research yobach@microsoft.com Michal Kosinski University of Cambridge mk583@cam.ac.uk Thore Graepel Microsoft Research thoreg@microsoft.com

Pushmeet Kohli Microsoft Research pkohli@microsoft.com David Stillwell University of Cambridge ds617@cam.ac.uk







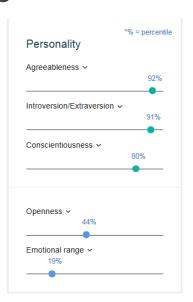




Try it yourself...

- mypersonality.org
 - University of Cambridge Psychometrics Centre
 - Data on 6m personality test results and 4m Facebook profiles
- IBM Watson Personality Insights







Smartphone apps









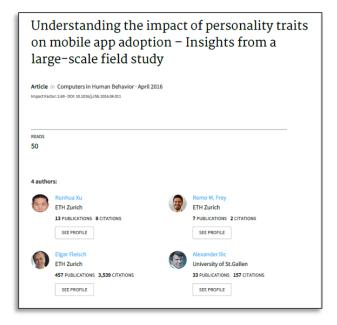




















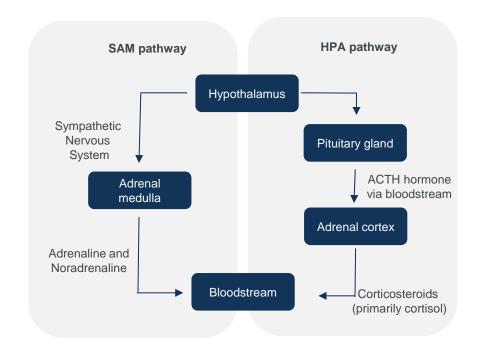
Institute and Faculty of Actuaries

Stress

$$\frac{{}_{t}p_{x}q_{x+t}p_{y}q_{y+t} + {}_{t}p_{x}q_{x+t}(1 - {}_{t}p_{y}) + {}_{t}p_{y}q_{y+t}(1 - {}_{t}p_{x})}{1 - (1 - {}_{t}p_{x})(1 - {}_{t}p_{y})} + X$$

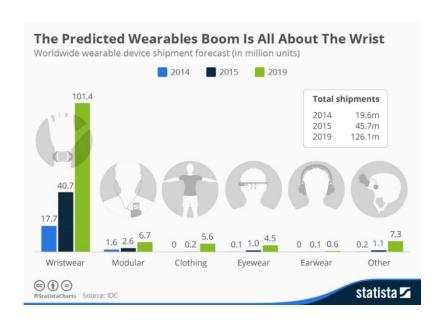
A Complex Process

- Biological aspects
- Psychological aspects
- Social aspects
- Acute stress
- Chronic stress
- Resilience
- Linked to:
 - Higher CVD Mortality and Morbidity
 - AIDS/HIV





Wearables, health scores and behavioural change

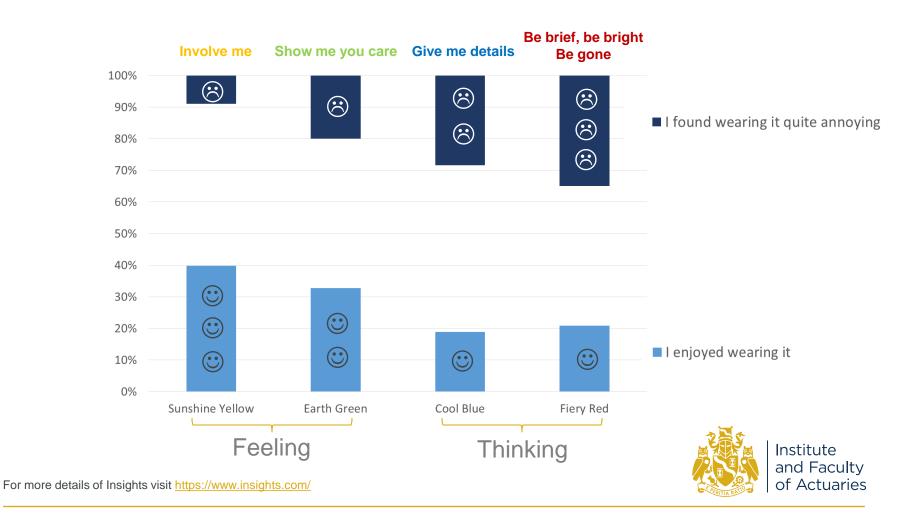








Enjoyment of wearing the device varies by psychometric profile





Should insurers use this data?

Holding and using (Big) Data – Peter's 4 R's

Regulations

- Increasing focus
- Data Protection Regulations
- Anti-discrimination legislation
- Profiling versus Decisions



How Data Makes Insurance Work Better for You

I. THE DIGITAL REVOLUTION AND BIG DATA pq6 The world is changing at an unprecedented pace - so just what is so different about data these days, and why does that affect the insurance industry? 2. UNDERSTANDING YOU pg I 0 Insurers need to understand you in order to provide you with products and services that really work for you. 3. GETTING THE RIGHT PRICE pg 14 Getting the right insurance isn't simply about getting the right product, it's also about getting it at the right price, which fairly reflects your circumstances. 4. ENABLING YOU TO MAKE THE RIGHT **DECISIONS TO MINIMISE YOUR RISK** pq18 Insurers take on risk on behalf of their customers, and are keen to use the changing digital world to help customers understand and control the risks they face. 5. IMPROVING THE CLAIMS PROCESS pq22 Insurers want to use data and technology to make the claims process as easy and flexible as possible, and to combat fraud more effectively. 6. HOW WE TREAT YOUR DATA pg26 Using data innovatively means that insurers have to continue to build a trusting relationship with their customers, which involves being transparent about how we use your data.

Source: How Data Makes Insurance Work Better For You, Association of British Insurers. Accessed February 2016



Holding and using (Big) Data – Peter's 4 R's

Regulations

- Increasing focus
- Data Protection Regulations
- Anti-discrimination legislation
- Profiling versus Decisions

Reasonable Expectations

- Fairness
- Country Specific
- Industry and data specific
- What have you told the client
- Does it benefit the consumer?



Approximate amount people would pay to protect each data type (US\$)

	United States	China	India	UK	German
Government Identification	\$11 2		•		
Credit Card Information		•	\$22		
Digital Communication	s	\$4	•		
Web Browsing		•			
Health History	•	•		\$5 9	\$184

Source: Customer Data: designing for Transparency and Trust, May 2015. Morey, Forbath and Schoop.

Original source included other types of data, all with low amounts (e.g. search history, location, purchase history, contact information

Researchers used purchasing parity rather than exchange rates to convert all amounts to U.S. dollars



Holding and using (Big) Data – Peter's 4 R's

Regulations

- Increasing focus
- Data Protection Regulations
- Anti-discrimination legislation
- Profiling versus Decisions

Risks

- Reputational damage
- Financial Penalties
- Theft / Mis-use / Loss
- Data governance procedures and culture

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Holding and using (Big) Data – Peter's 4 R's

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Reasonable Expectations

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- What have you told the client
- Does it benefit the consumer?

Rewards

- Consumer
- Company (Red vs Blue Ocean)
- Employee
- Regulator





Conclusions



Conclusions

- Looked at 5 factors of personality and 4 aspects of SWB
- Self-reported information can be manipulated
- Hard to validate and challenge responses to personality tests or SWB surveys
- Look instead to objective measures linked to personality or SWB
- Credit behaviour is cutting edge
- Availability, access, understanding and manipulation of data
- "Do the right thing"
- Regulations, Risks, Reasonable expectations, Rewards

Be responsible custodians of data and give your customers what they want



Questions

Comments

Expressions of individual views by members of the Institute and Faculty of Actuaries and its staff are encouraged.

The views expressed in this presentation are those of the presenter.



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