Latent Trait Analysis for Risk Management of Complex Information Technology Projects

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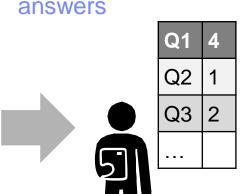


- □ Input data, x : questionnaire answers
 - Surveyor asks about the project status
 - Project manager answers to the questions
- □ Predicted value, y: failure or success (after contract signing)

IT system development project



x: questionnaire





failure or success?











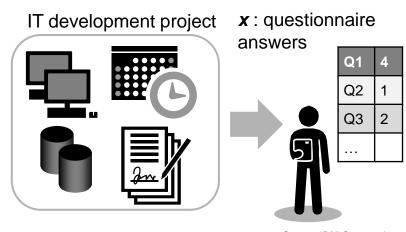




(For ref.) What the questionnaire looks like

□Major topics covered

- Communication issues with the client
- Well-definedness of the project scope
- Issues related to subcontractors and internal teams
- -Project management issues











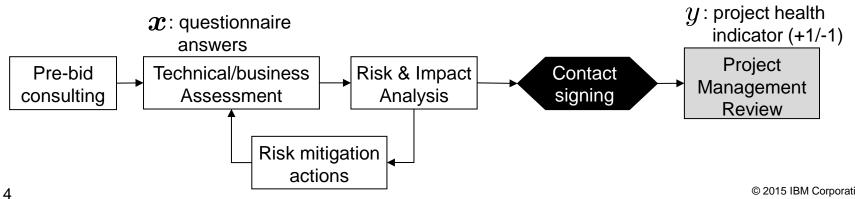




Challenge (1): No evidently bad answers. Need to discover indications of failures from apparently good answers

- □lterative review process allows removing all evident risk factors
 - -This is actually a prerequisite to get into the final review right before contract signing

- □However, some of them might be "pretending" as good
- □Wish to discover such indications









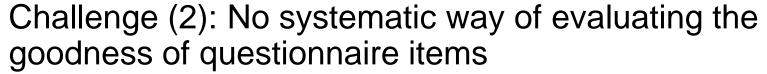




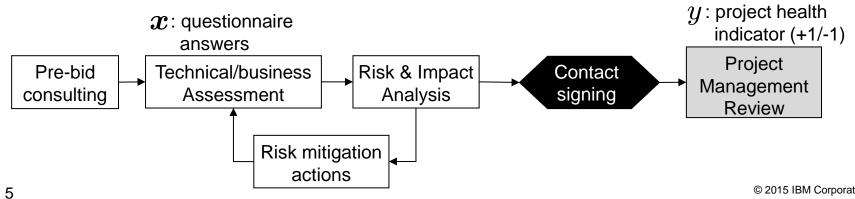








- □Individual questions are defined by human experts based on their experience
- **□Some questions may be** unnecessary and/or incur too much cost to answer
- **□Wish to have a** quantitative approach to evaluate the informativeness of each questions.













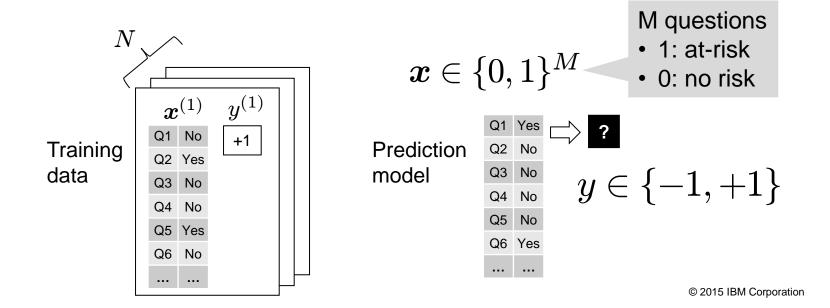






Problem summary

- □1. Compute the informativeness of the question items
- □2. Build a predictive model for project failure/success (y) given a new set of questionnaire answers (x)











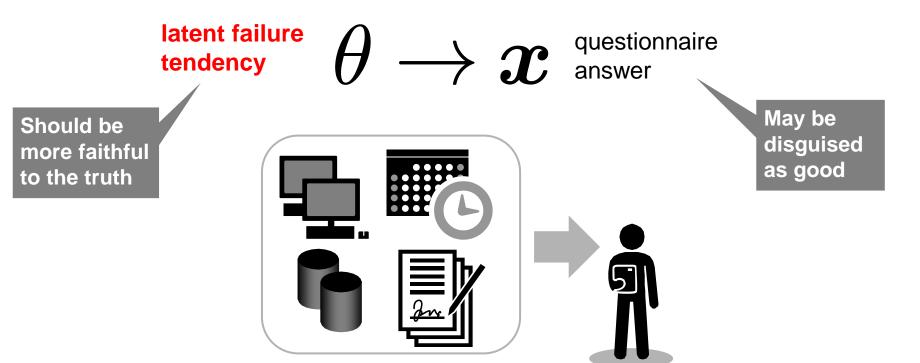








Key idea: Assume **x** is stochastically generated by a latent variable that is more faithful to the truth















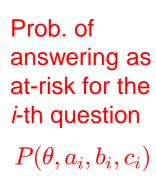


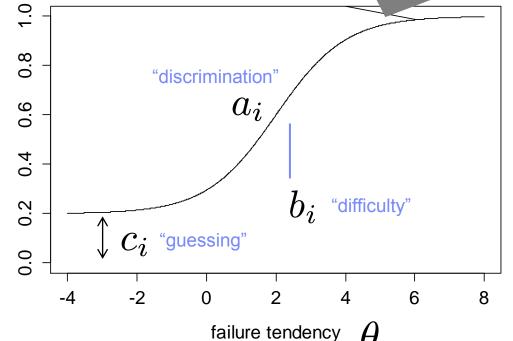


Using a "shifted S-curve" as a natural model of cognitive biases

□Represents nonlinear relationship of $\; heta o oldsymbol{x} \;$

- Overly optimistic for smaller risks
- Overly pessimistic for larger risks
- Sometimes use a guess





This type of model is known as latent trait **model** in psychometrics



















(For ref.) Latent trait model

□The method to normalize the SAT test

- -The total score is NOT simply the total number of accurately answered items
 - Scores are normalized to be in the range of [200-800].
 - Even random guess to all of the items gives you at least 200 points to each section.
- -Examinee's ability is treated as a latent variable to be estimated
- □In our case, examinees' ability corresponds to latent failure tendency





















Bayesian framework to estimate latent failure tendency

Historical record

$$(\boldsymbol{x}^{(n)}, y^{(n)})$$

$$n = 1, \dots, N$$

training phase

New question answer

prediction phase

Generative model for x based on the shifted S-curve

$$p(\boldsymbol{x} \mid \theta, \boldsymbol{a}, \boldsymbol{b}, \boldsymbol{c})$$

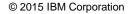
Prior distribution for theta

$$f(\theta) = \sqrt{\frac{1}{2}} \exp\left(-\frac{1}{2}\theta^2\right)$$

model parameters

$$\hat{m{a}},\hat{m{b}},\hat{m{c}}$$

























Making prediction using estimated latent failure tendency

New question answer

prediction phase

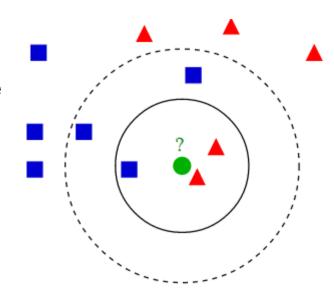
estimated failure tendency



Simply use k-NN classification based on the distance in the latent space

$$d(\boldsymbol{x}, \boldsymbol{x}^{(n)}) = \sum_{g=1}^{G} w_g \left(\hat{\theta}_g - \hat{\theta}_g^{(n)}\right)^2$$

Extend the original LTA to include multiple latent variables (→ see the paper)





















Experiment: Using service provider's real quality assurance data

- Questionnaire called CRA (contract risk assessment)
- □M = 22 rather qualitative questions
- □N = several hundred
- □Each question is yes (at-risk) or no (no-risk)
- □Final project evaluation is failure (y=+1) or non-failure (y = -1)





















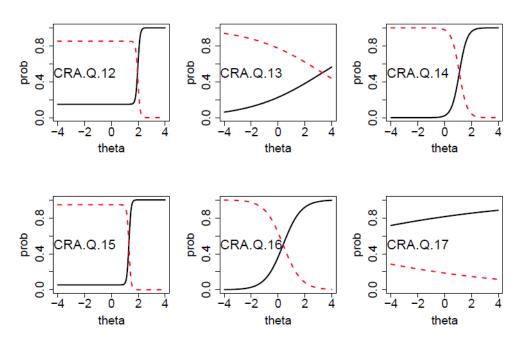
Result (1): Estimated S-curves clearly provide practical information on the usefulness of each question

□Q17 is hardly useful

- -Formal question on service pricing
- –Expected to be less influential

□Q14 and Q16 are useful

 Ask straight about the clarity and feasibility of the development plan























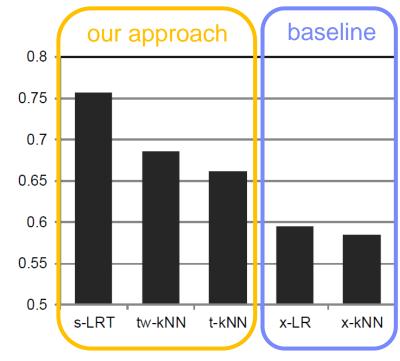
Result (2): Achieved better accuracy

□Compares F-value

-harmonic mean between troubled project accuracy and non-troubled project accuracy

□Clearly outperform the baseline

- -Baseline is based only on x
 - Logistic regression
 - Simple k-NN
- Our approach uses theta instead of x



Comparison of failure prediction accuracies.





















Conclusion:

□Proposed a new approach to project risk management inspired by a psychometric theory

- **□By correcting cognitive biases, we confirmed that the** proposed method
 - -provides practically useful information on the usefulness of each question
 - -clearly outperforms known alternatives



Thank you!

















Our session TS9: Big Data for Management, is scheduled for Wed, May 13, 2015, 14:00 - 15:30

The order of presentation is:

- Early Network Failure Detection System by Analyzing Twitter Data
 - Kei Takeshita, Masahiro Yokota, Ken Nishimatsu, NTT,
- Resolution Recommendation for Event Tickets in Service Management
 - Wubai Zhou, Liang Tang, Tao Li, Florida International University, USA,
 - · Larisa Shwartz, IBM T.J. Watson Research Center, USA,
 - Genady Grabarnik, St. John's University, USA.
- Dude, Ask The Experts!: Android Resource Access Permission Recommendation with RecDroid
 - Bahman Rashidi, Carol Fung, Virginia Commonwealth University, Canada,
 - Tam Vu, University of Colorado Denver, USA.
- Latent Trait Analysis for Risk Management of Complex Information Technology Projects
 - Tsuyoshi Ide, Sinem Guven, Ea-Ee Jan, IBM T. J. Watson Research Center, USA,
 - Sergey Makogon, Alejandro Venegas, IBM Global Technology Services, Algeria.

Chair: Prof Hanan Lutfiyya

- ハナン・ルトフィヤ教授
- presentation should last around 17 minutes within the 22 minutes allocated slot time to leave at least (!) 5 minutes for questions.

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