

GIRO Convention

23-26 September 2008
Hilton Sorrento Palace

Report of the Demand Modelling Working Party

Workshop A1

James Tanser, Watson Wyatt

Owen Morris, Norwich Union

Demand Modelling Working Party

- James Tanser (Chair)
- John Light
- Sophia Mealy
- Owen Morris

Special thanks to:

- Julie Fairbank
- Matthew Barnes

GIRO Working party

- Provide an introduction to the topic describing the terms used
- Summarise the current methodologies used in the market
- Summarise possible alternate methodologies identified by a search of available literature
- Investigate several methods using agreed methodology to determine the descriptive and predictive power of the methods when applied to actual insurance data
- Provide a brief conclusion and highlight areas for further work.

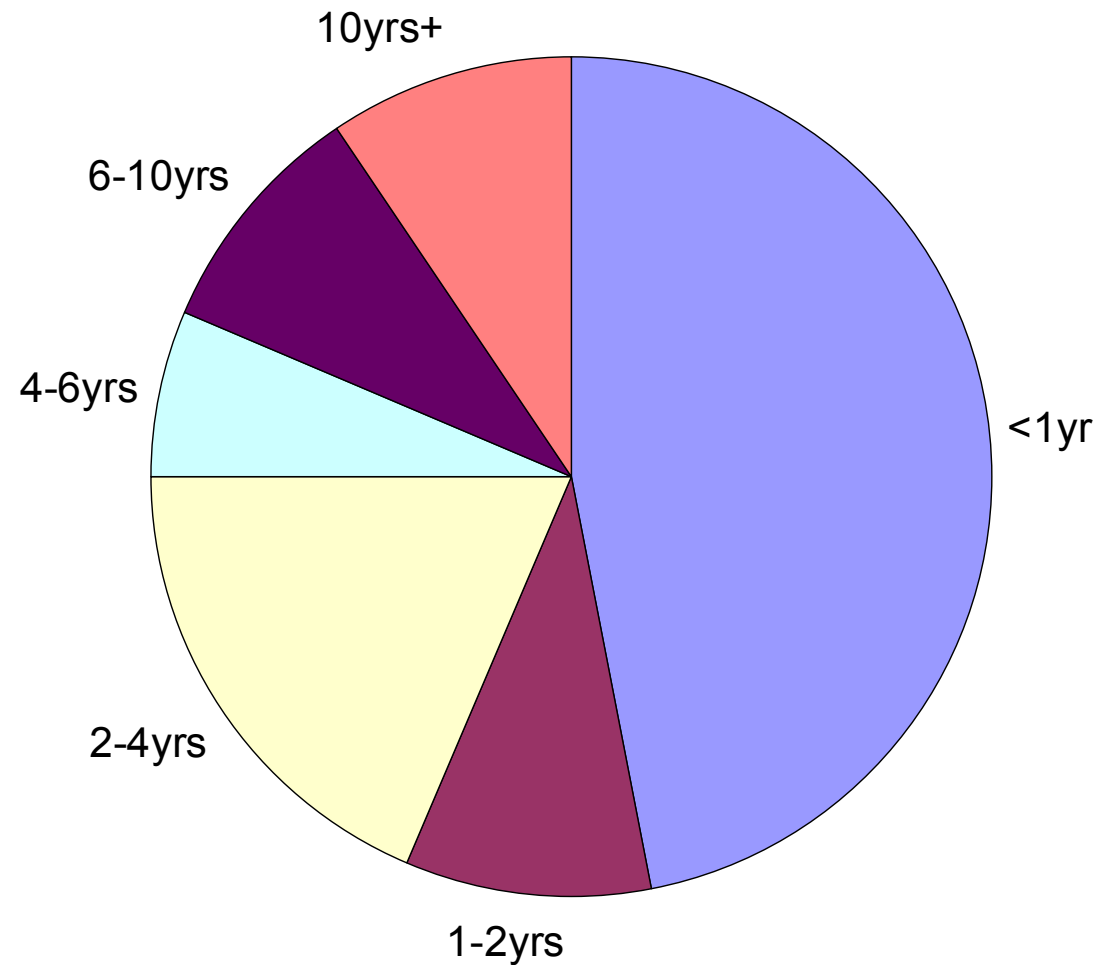
Agenda

- Survey
- Practical matters
- Comparative study
- Questions

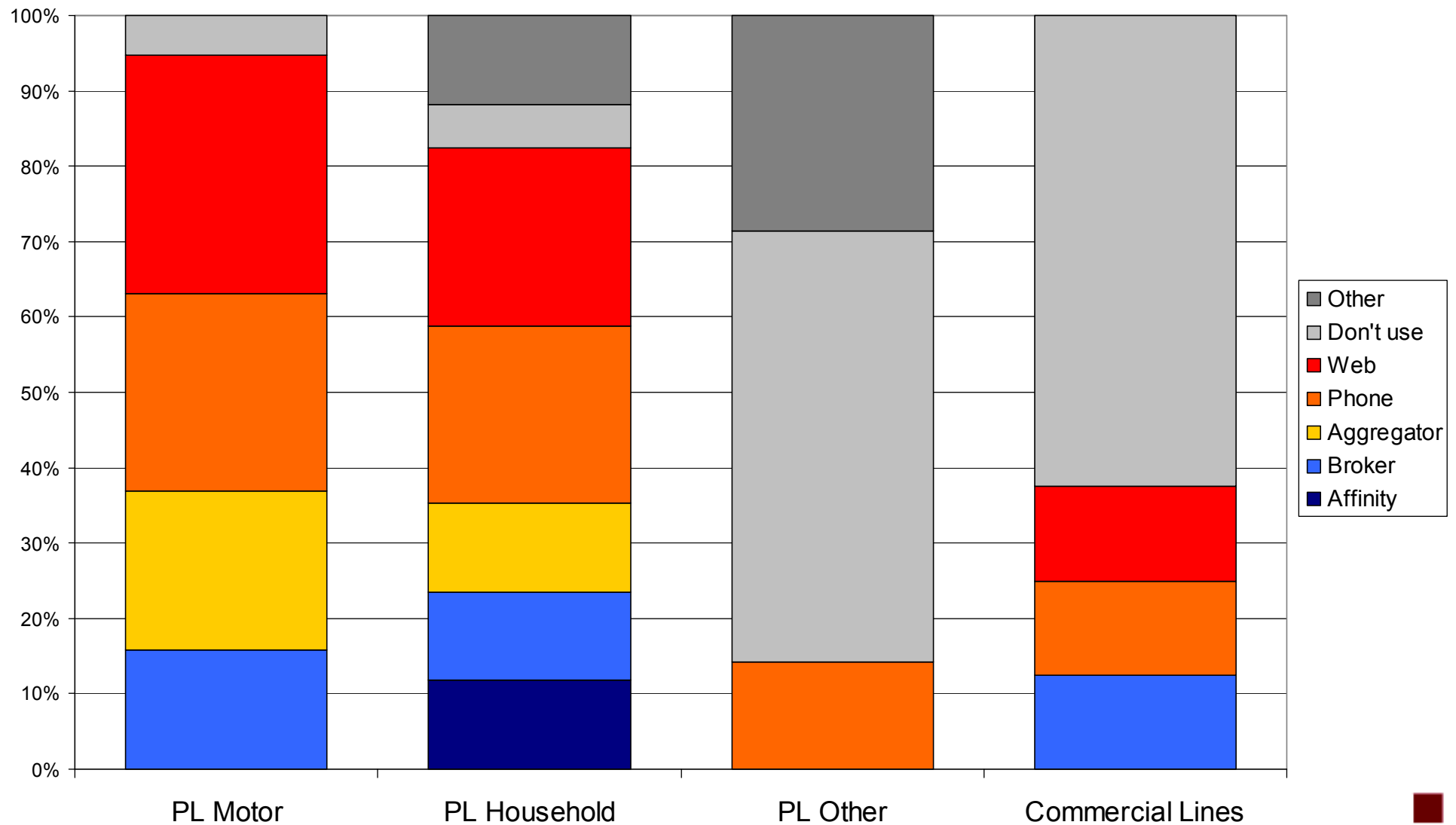
Survey

- Small response: 32 started and 11 finished
- Results interesting despite low response
- Champagne and Chocolates are equally popular!

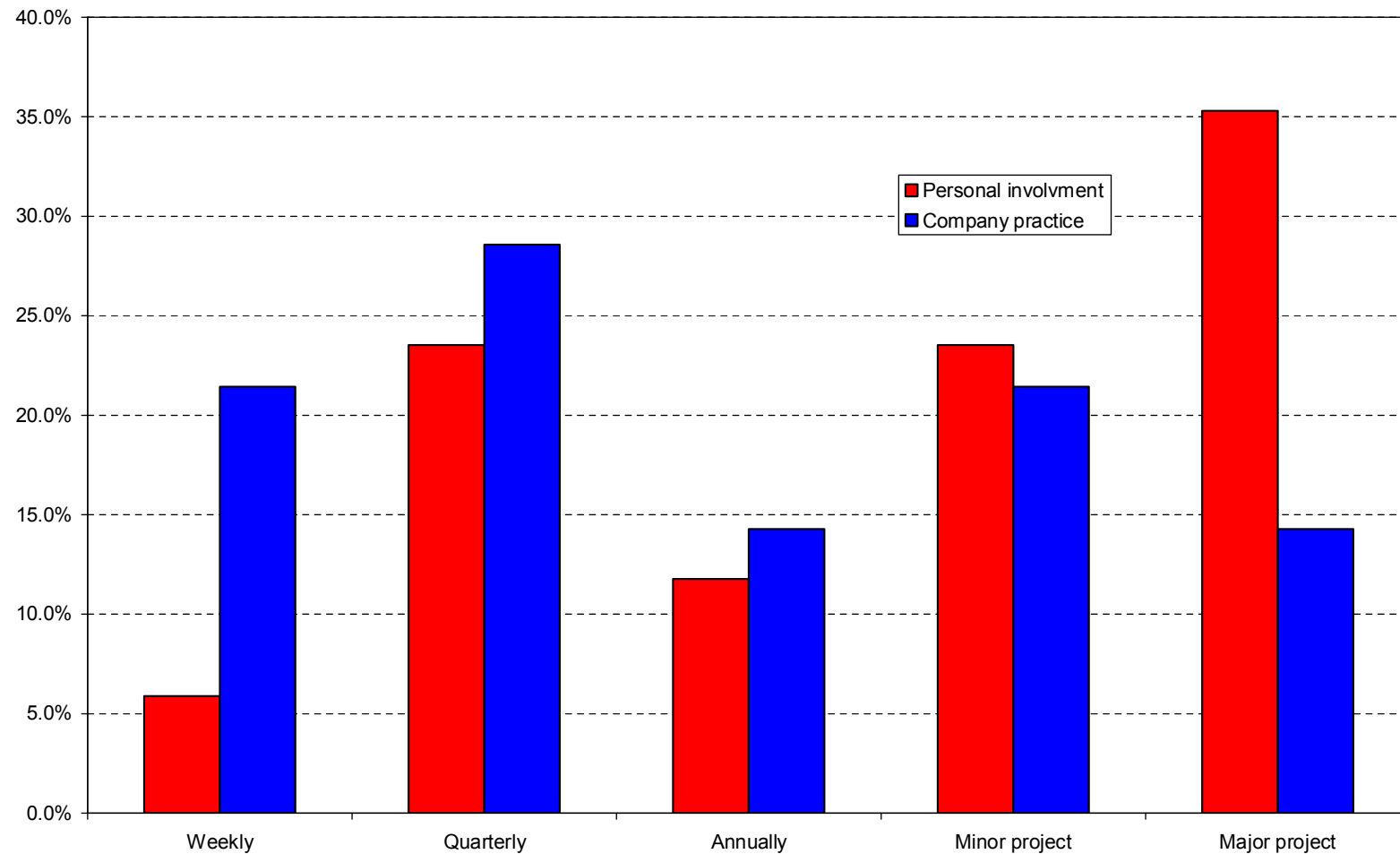
Number of years working with demand models



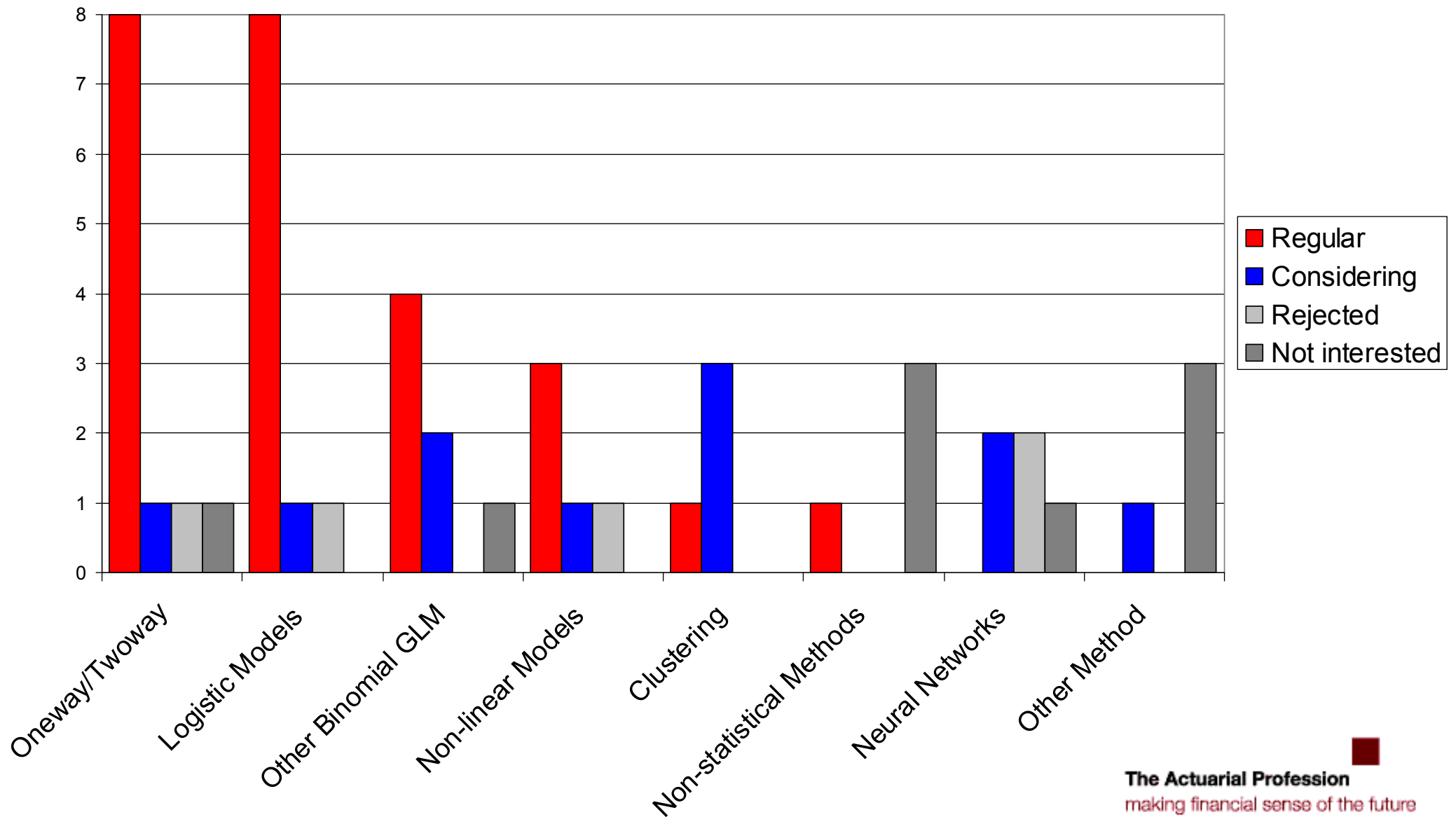
Where are demand models used?



Frequency of analysis



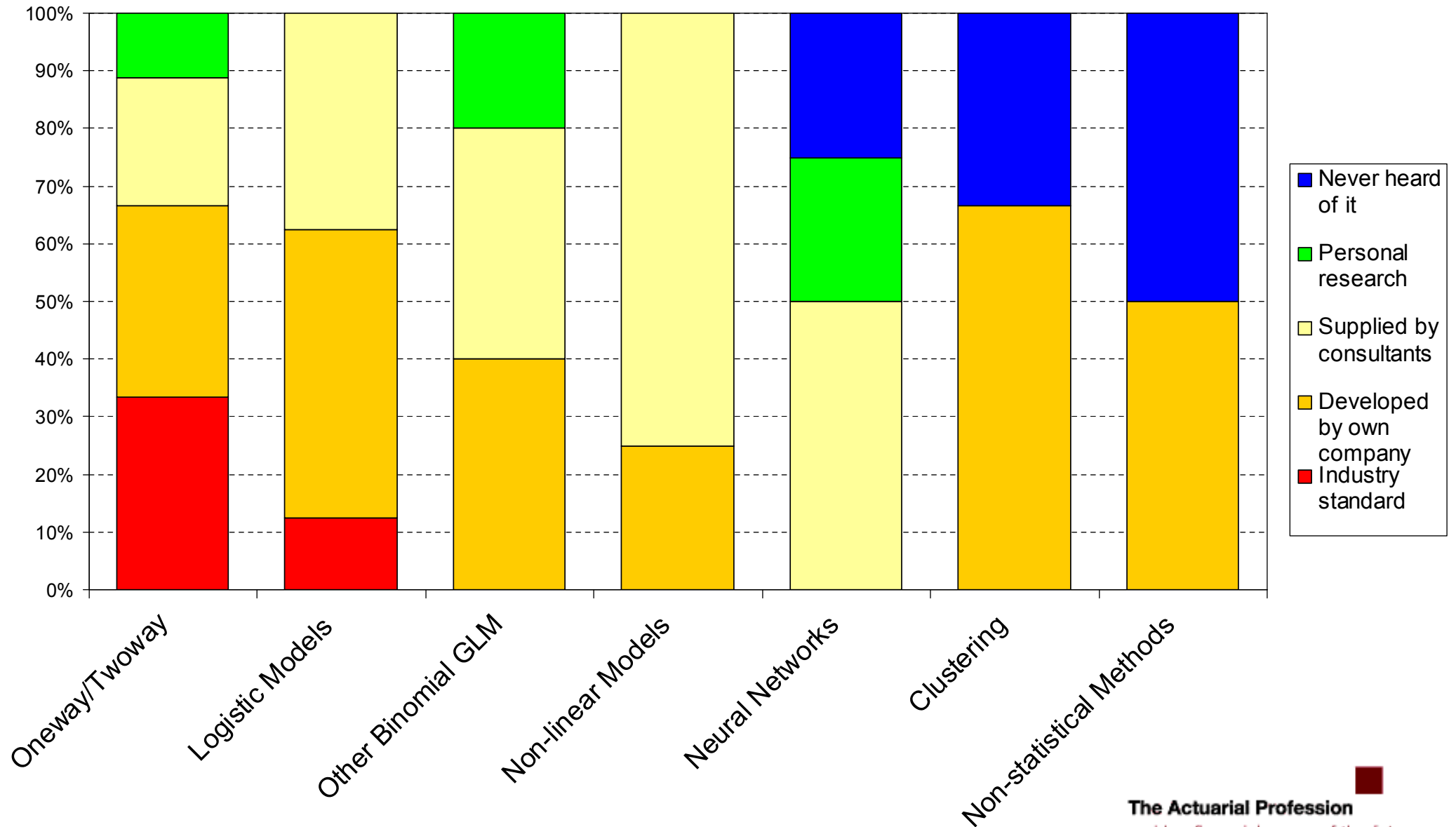
Methods in use



Who uses the methods?

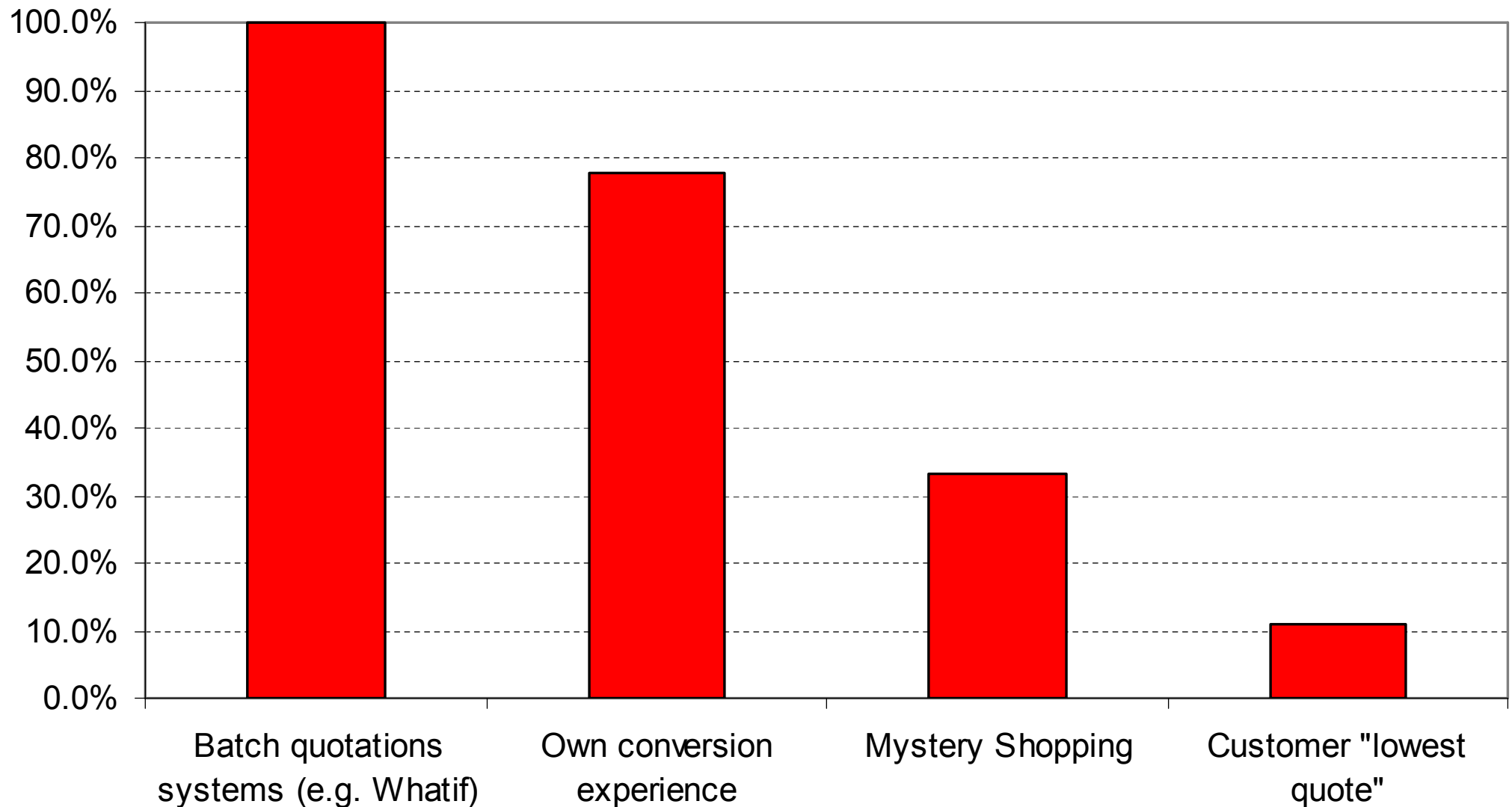
- One-ways and two-ways always used in combination with other technique
- 7/10 used one-way with Logistic (with one more just using logistic)
- 4/10 used other Binomial models, and of these 4, 2 looked at non-linear models and 1 at clustering

Source of models

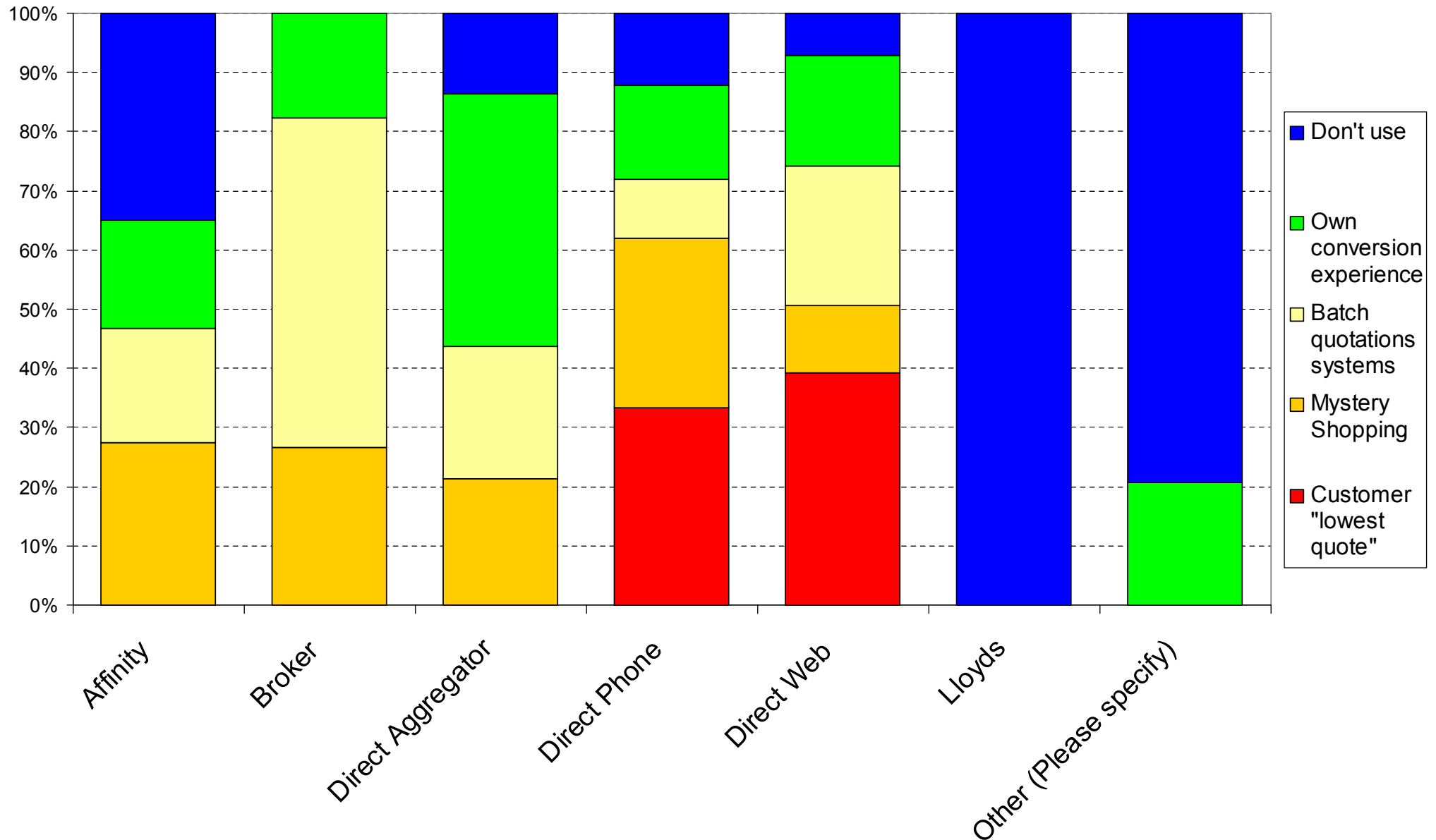


Sources of information

Competitor Data Sources - Overall Use



Sources of information



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Data

- Key to successful analysis
- Take care with missings
 - Systematic effects may distort analysis
- More data = More detail
- Beware trends and changes

New business versus Renewals

- Inertia key to renewals
 - Can get reasonable models without market information
- Market premium key to new business
- Price sensitivity best measured through trials

Competitor premiums

- Key to analysis, and hard to get!
- Relevance of information varies by channel:
 - Quotation systems
 - Screen scraping
 - Customer self reporting
- Aggregators own key data

Use of models

- Price optimisation
- Scenario testing
- Marketing

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Generalised linear models

$$E[\underline{Y}] = \underline{\mu} = g^{-1}(\underline{X} \cdot \underline{\beta} + \underline{\xi})$$

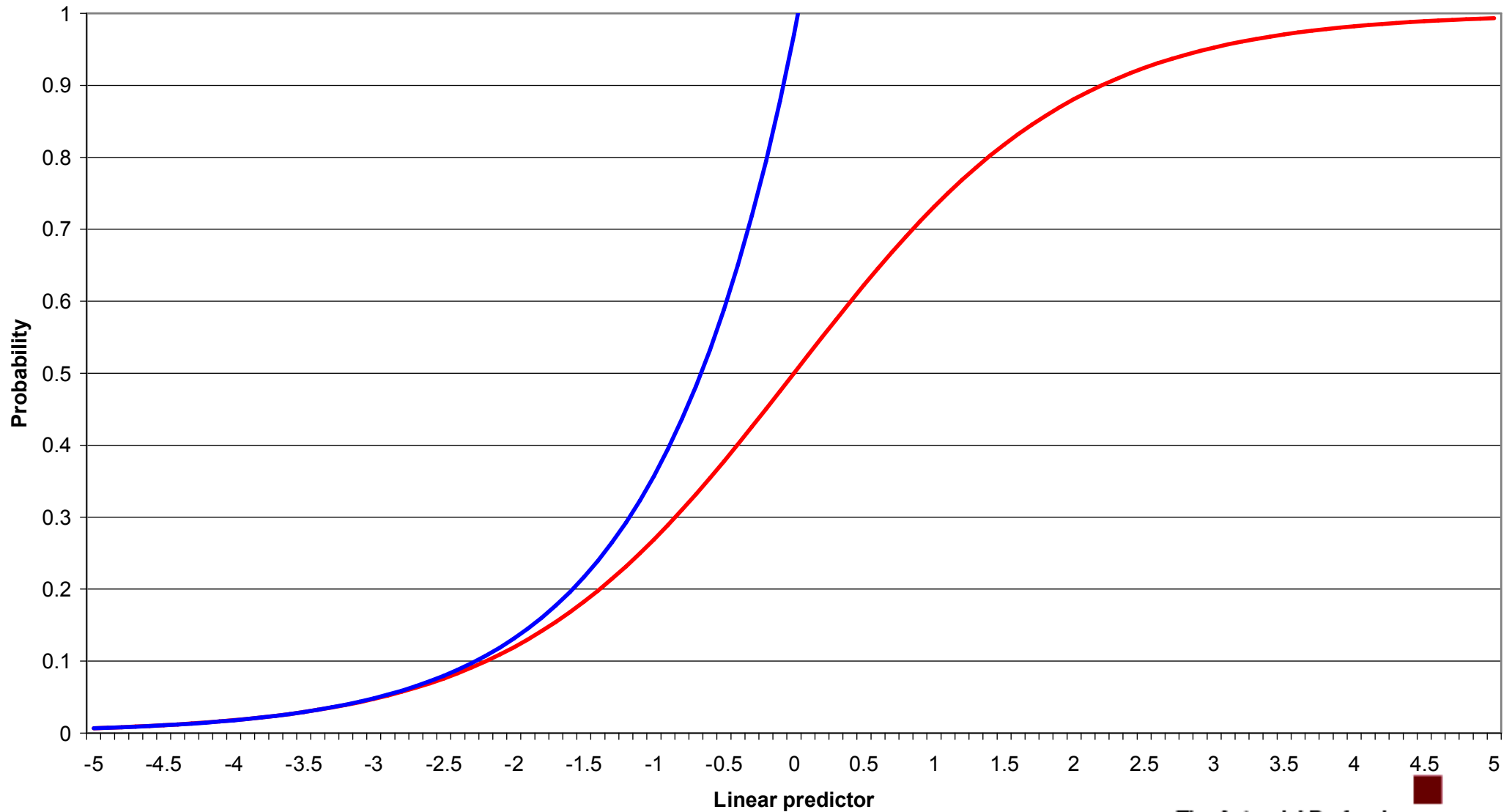
$$\text{Var}[\underline{Y}] = \phi \cdot V(\underline{\mu}) / \underline{\omega}$$

Models: traditional view

- A logistic model is most appropriate
 - considers $\log(p / [1-p])$ with binomial error
 - maps $[0,1]$ to $[-\infty, \infty]$
 - invariant to whether you model success or failure
- If lapses are low and results not to be used directly, a Poisson multiplicative model can help
 - theoretically wrong (can predict multiple lapses), but easier to communicate

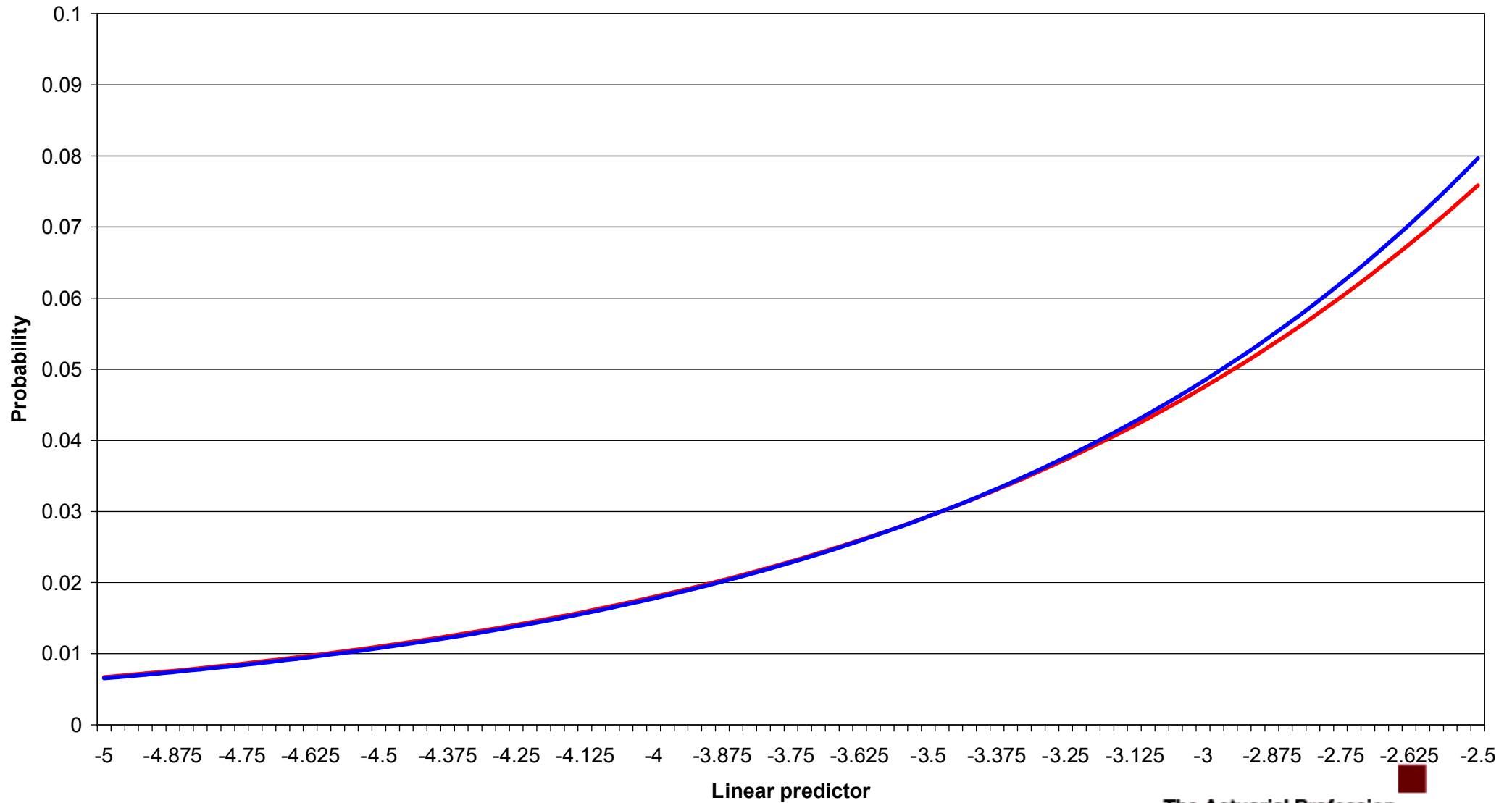
Log versus Logit

Logit link function



Log versus Logit

Logit link function

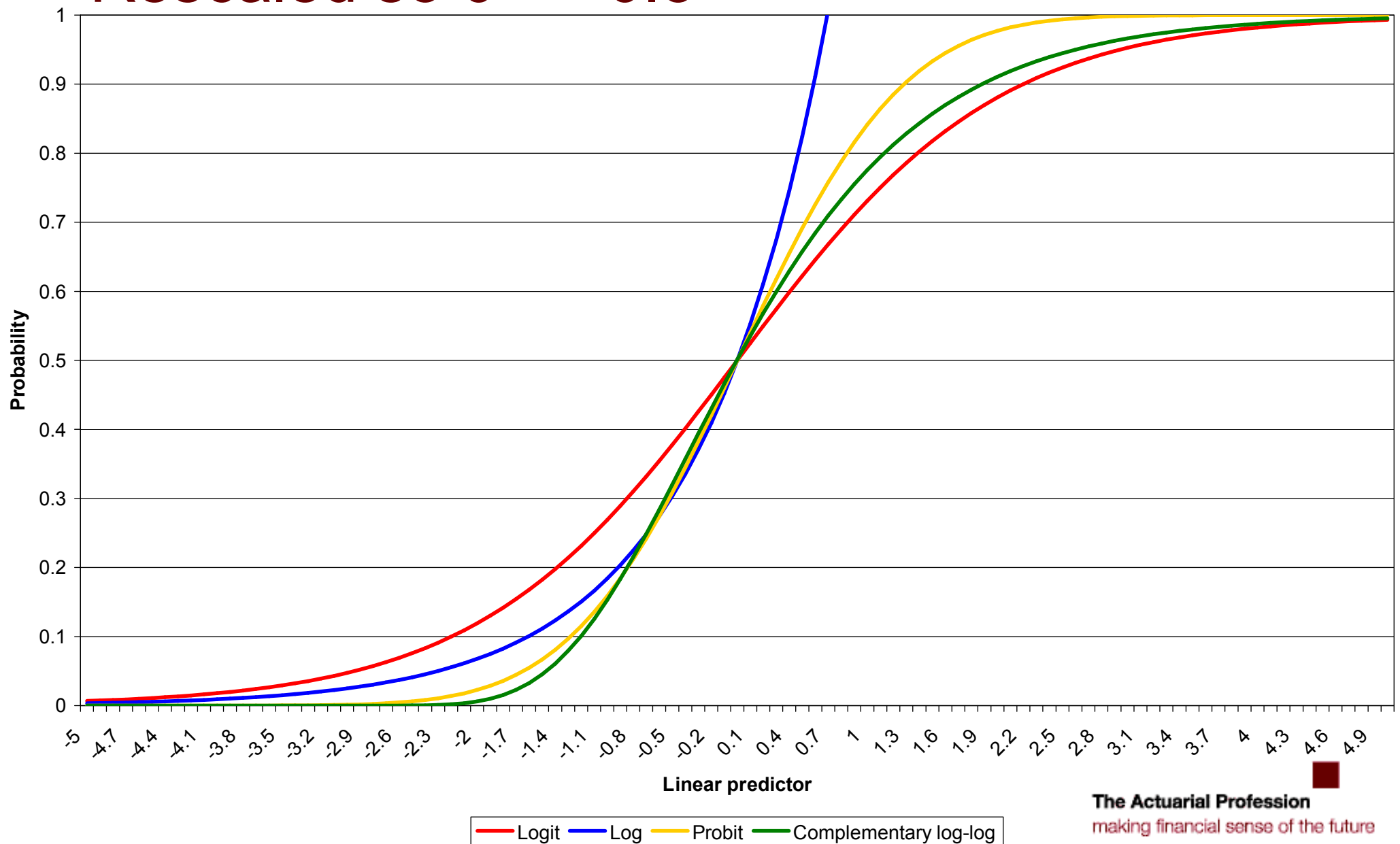


Other models

- Could try:
 - log link
 - probit link
 - complementary log-log link
- Transform the data
 - Sampling

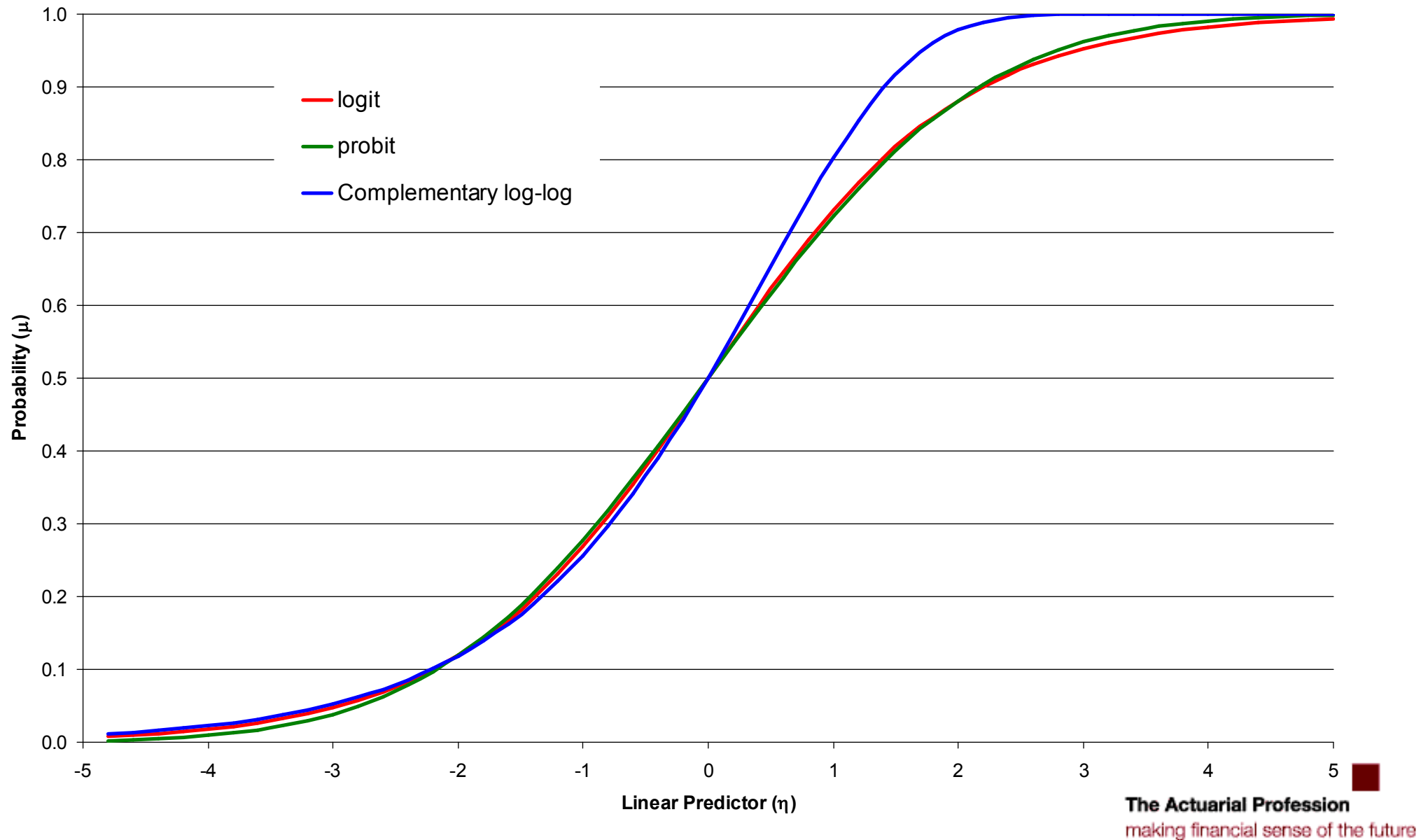
Link functions

Rescaled so $0 \Rightarrow 0.5$



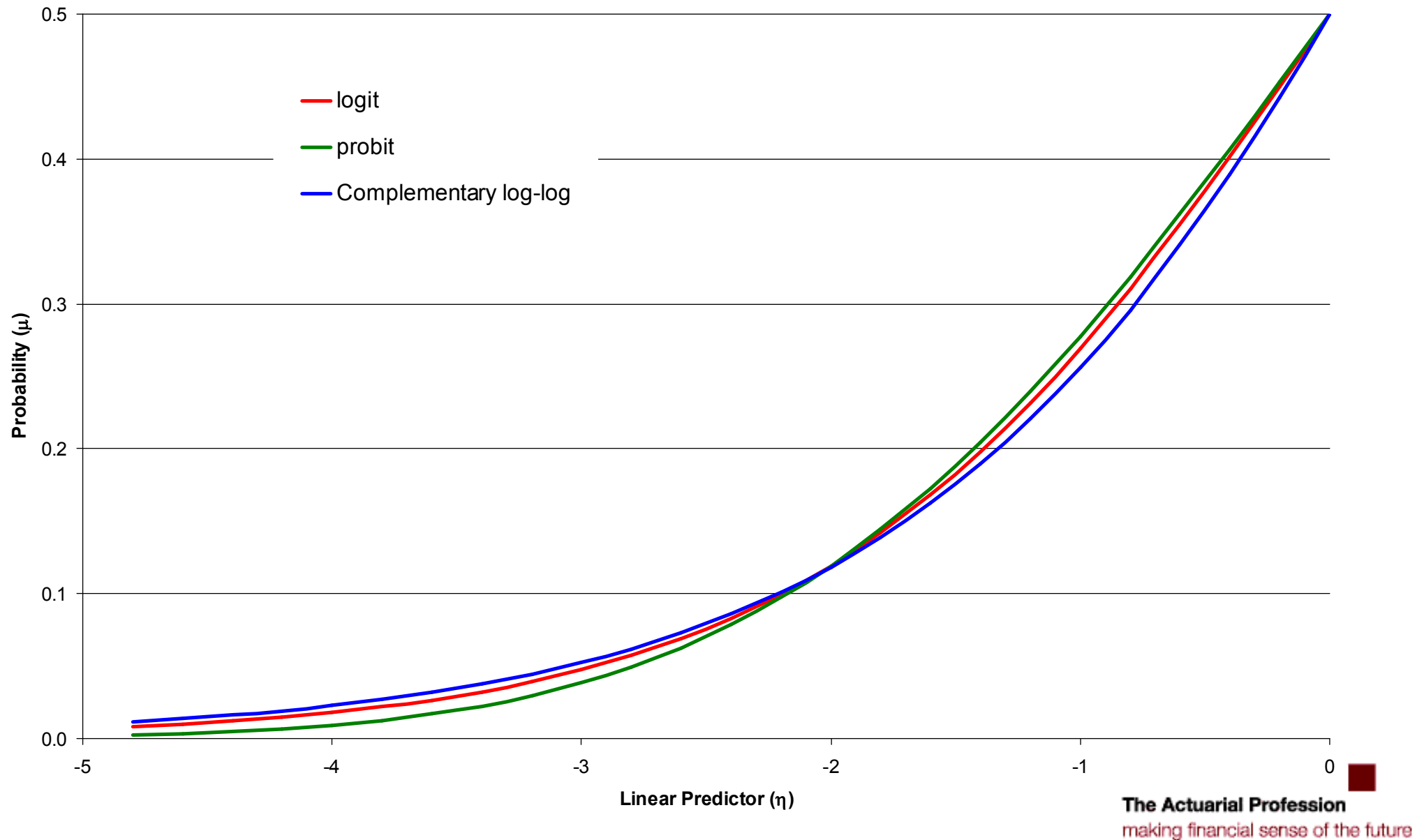
Link functions

Rescaled to be the same at 0 and -2



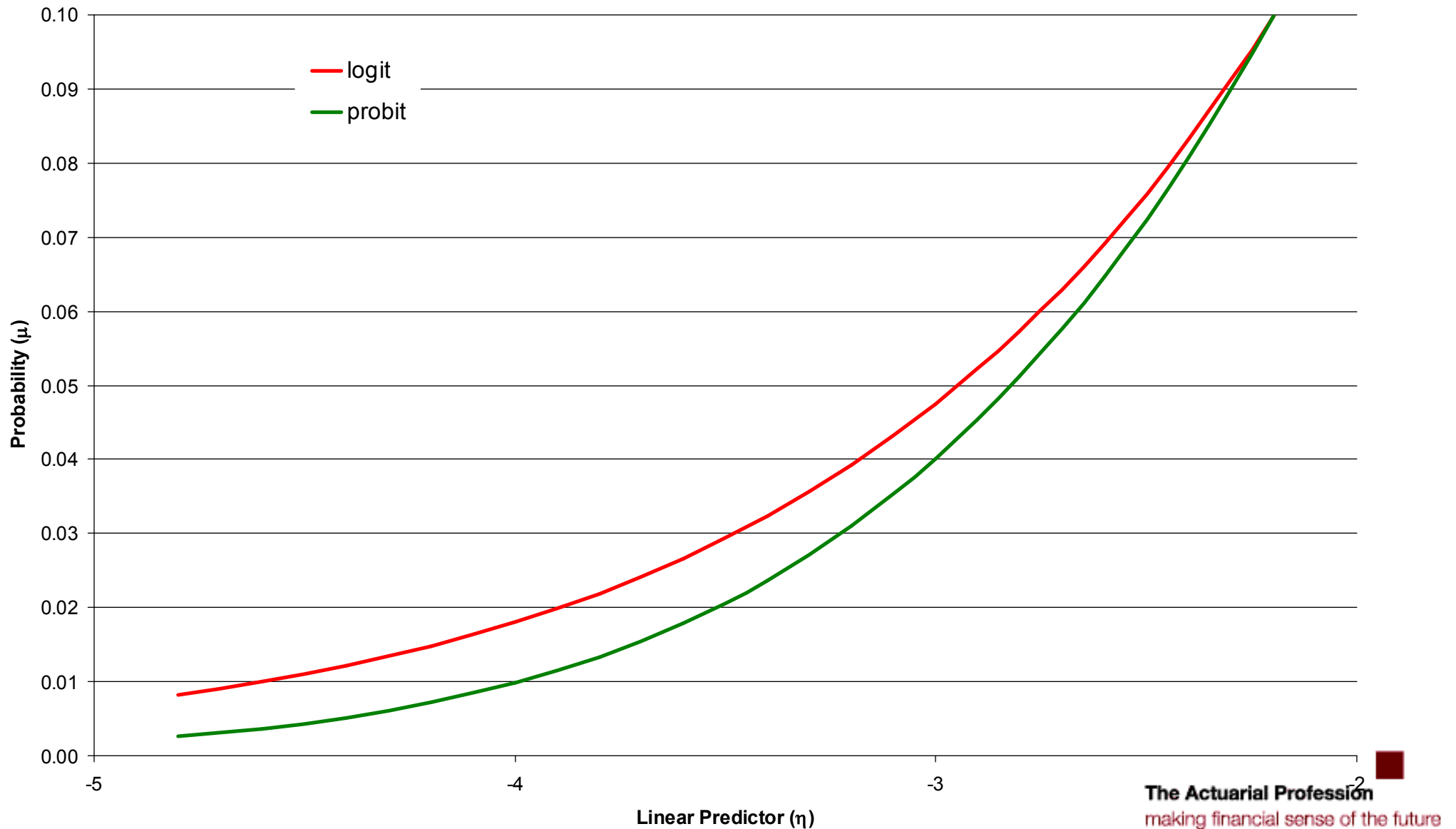
Link functions

Rescaled to be the same at 0 and -2



Link functions

Rescaled to be the same at $p=0.1$



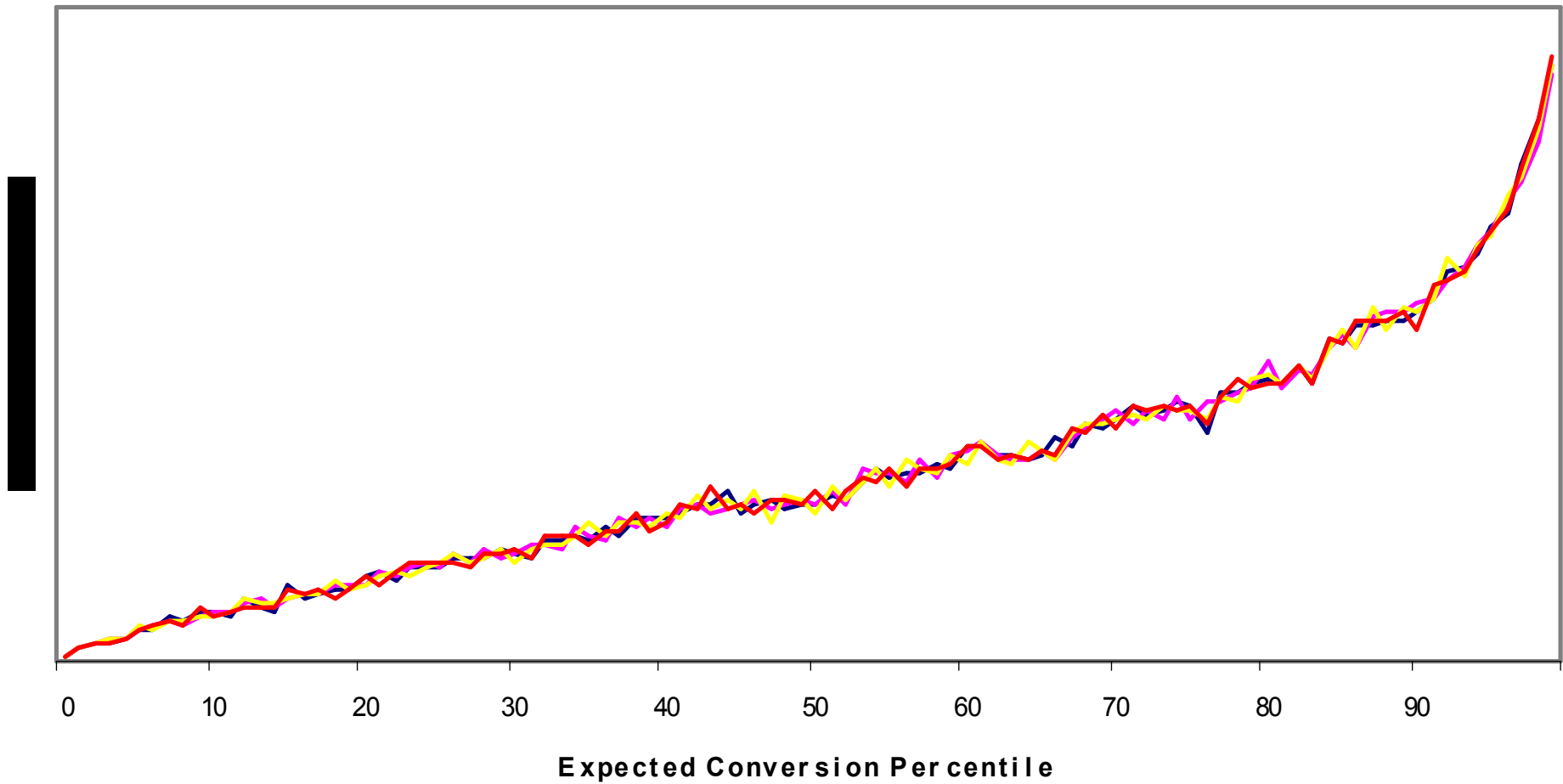
Data analysis

- Two datasets examined:
 - “High” typical of traditional channels
 - “Low” typical of new channels
- Data split into Train and Test using time split
- Base model was Logistic
 - Briefly iterated to get reasonable model
 - Same variable selection applied everywhere

Lift curves: Definition

- Take out of sample data and add fitted values
- Sort data according to expected value
- Create N pots of equal exposure
 - N typically 10, here 100
- Calculate actual in each pot and plot on graph
- Key points:
 - Test of order only, not goodness of fit
 - Can compare very different models easily

Lift Curve - High Conversion Data Set



LOGIT



POISSON

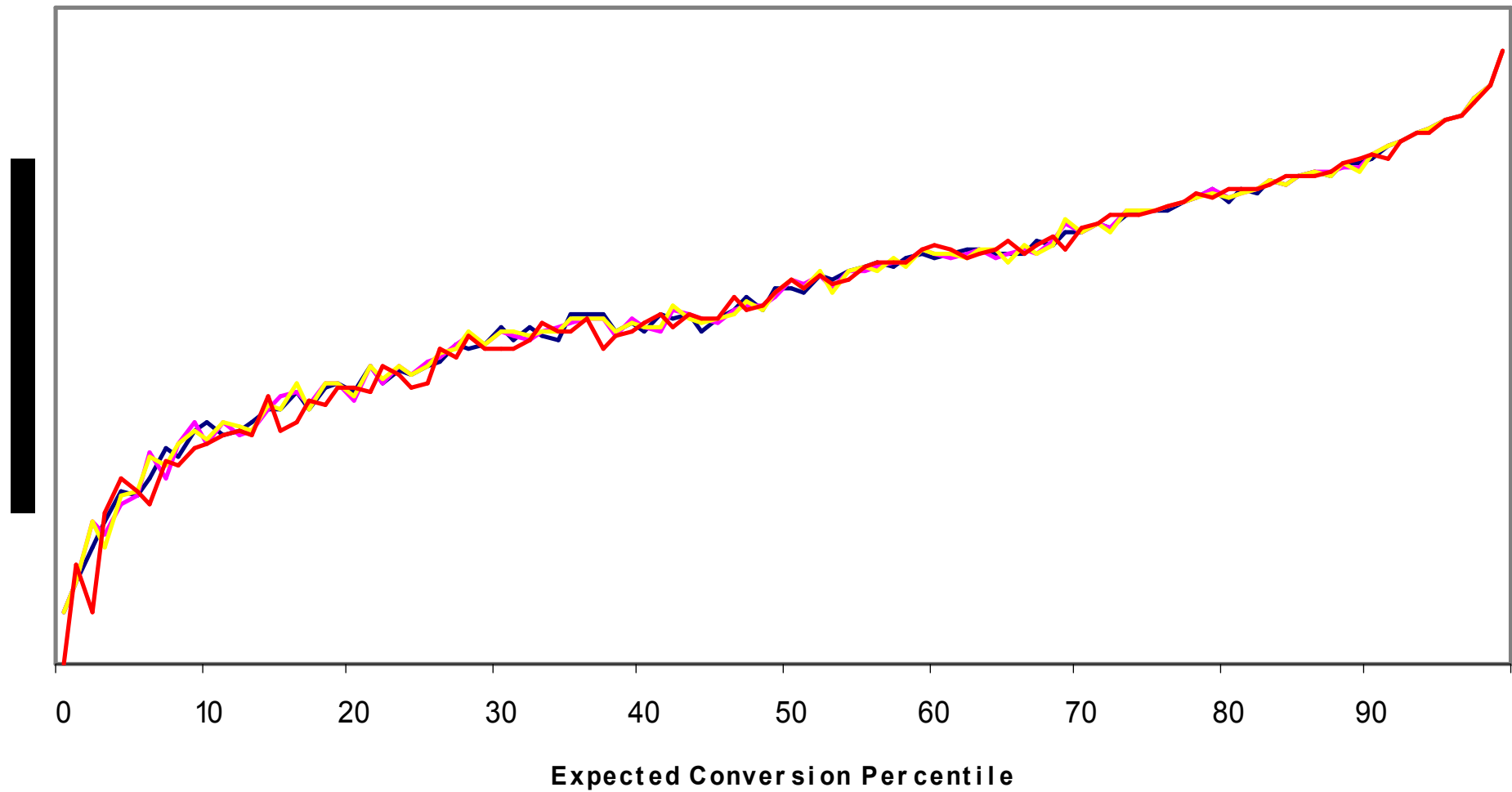


COMPLL



PROBIT

Lift Curve - Low Conversion Data Set



Note: Y-axis on log scale

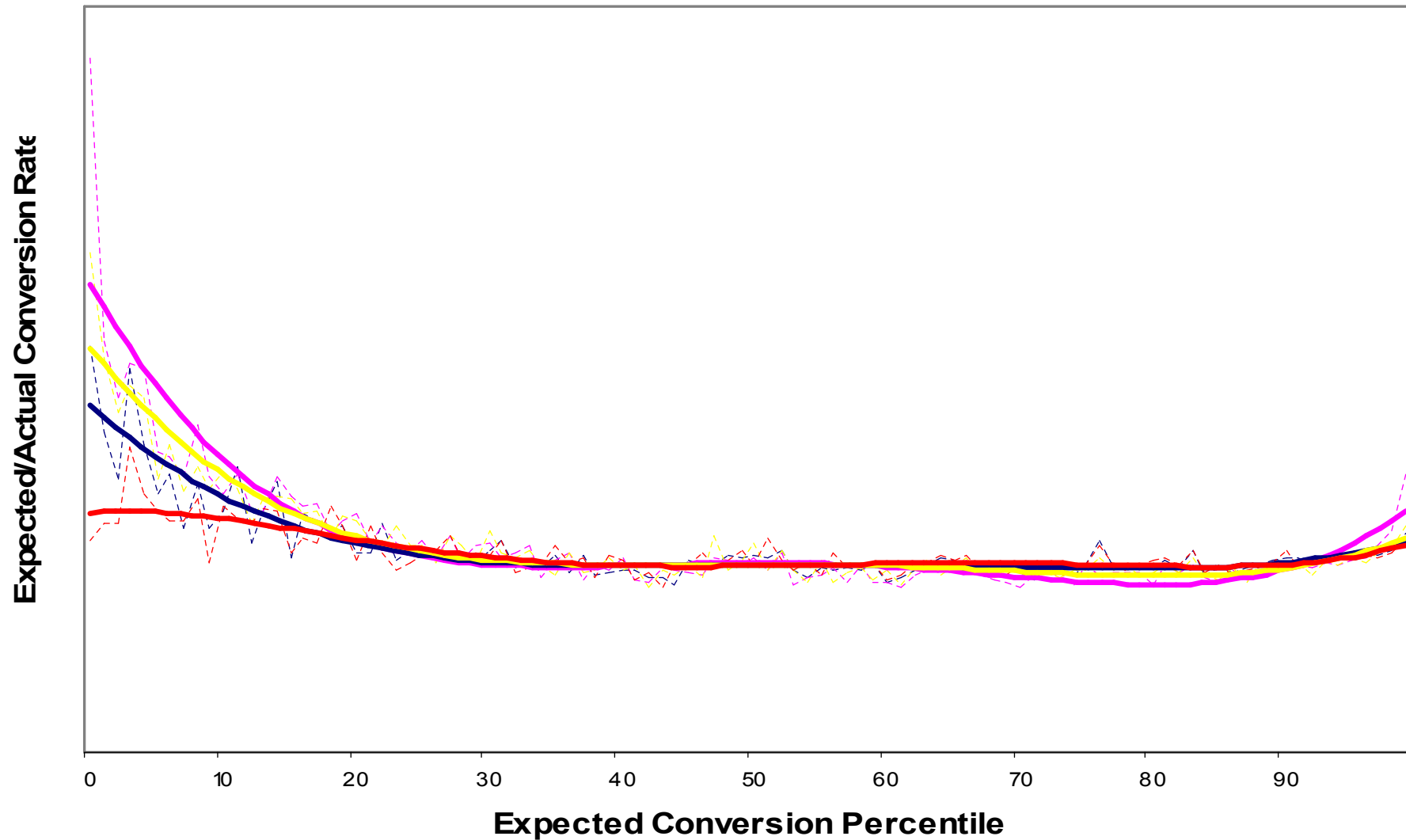
Lift curves: Conclusions

- All models appeared to do equally well in separating high and low conversion segments

Actual versus Expected: Definition

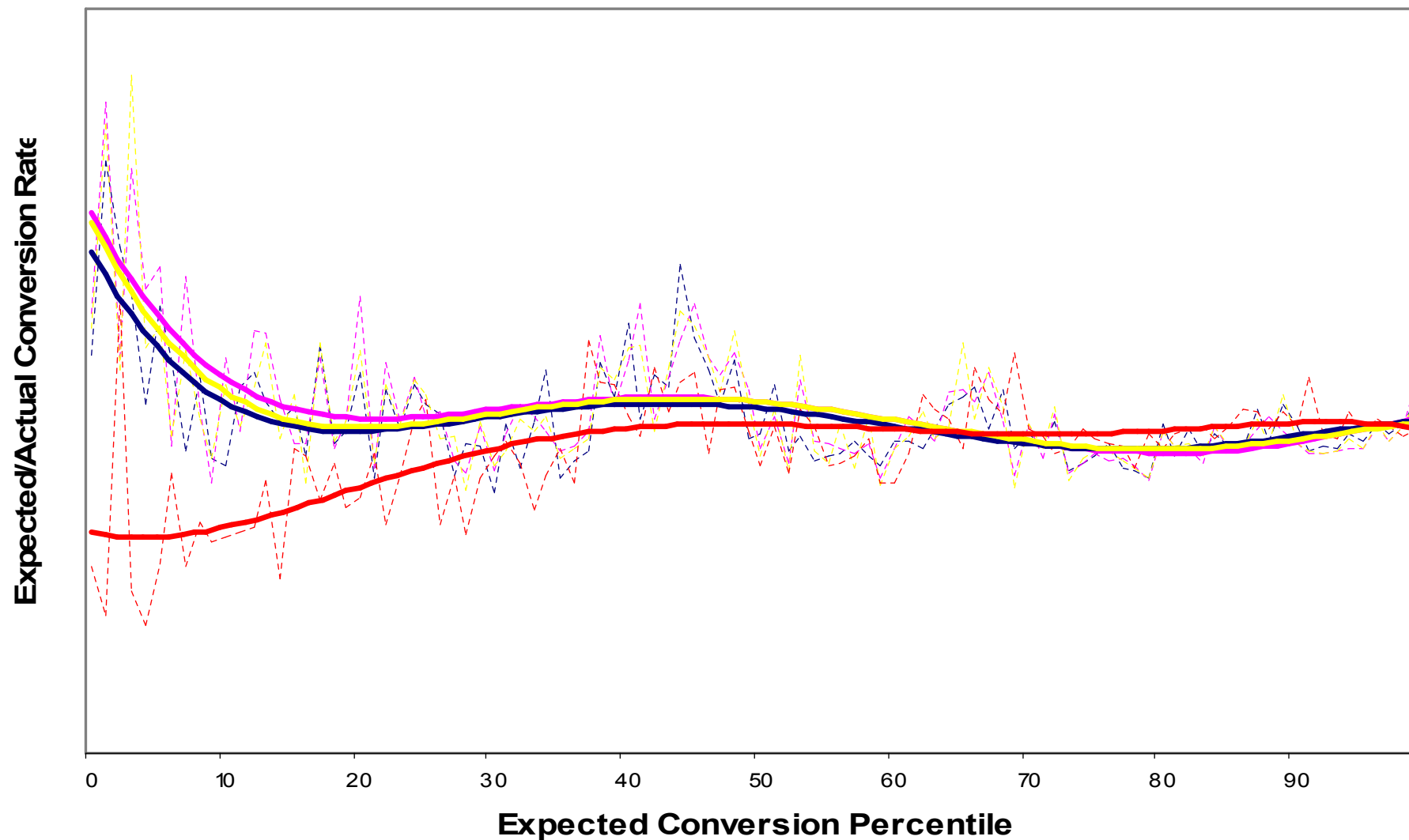
- Take out of sample data and add fitted values
- Sort data according to expected value
- Create 100 pots of equal exposure
- Calculate Expected / Actual in each pot and plot on graph
- Key points:
 - A flat line is equally good (or bad) everywhere
 - Systematic over or under estimation revealed by departure from $y=1$ line (not shown)

Actual Vs Expected - High Conversion Data Set



--- LOGIT	--- POISSON	--- COMPLL	--- PROBIT
— Poly. (POISSON)	— Poly. (LOGIT)	— Poly. (COMPLL)	— Poly. (PROBIT)

Actual Vs Expected - Low Conversion Data Set



Actual versus Expected: Conclusions

- Within any given model, there appears to be a systematic overestimation of low conversion segments
- Poisson/Log link is worst option
 - Poor at both ends for high conversion!
- Binomial/Probit appears flatter
 - Not clear what is happening for low conversion
- Binomial/Logit next best shape (but marginally better predictor)
 - Methodology may bias analysis to prefer this method

Future research

- More datasets would help firm up conclusions
- More exotic methods not examined:
 - Sampling
 - Non-linear models
- Some topics remain unpublishable:
 - Best interactions
 - Best data sources
 - Best way to treat competitor information

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