

Technical writing is different from expository or literary writing in a number of ways. I will briefly discuss

Structure

Figures, tables, and equations

Writing about data

Writing about statistical calculations

More stuff to read (chapters on github repo):

- Appendix to “Stat Labs” by Deb Nolan and Terry Speed
- Book “Eloquent Science” by David M. Schultz - see especially the chapter for non-native speakers of English
- Article “The Science of Scientific Writing” by Gopen and Swan - introduces idea of reader expectations

Examples for this lecture are taken from

Beaudoin and Swartz (2010) Strategies for pulling the goalie in hockey. *The American Statistician*, 64: 197-204.

Berrocal et al. (2010) Probabilistic weather forecasting for winter road maintenance. *Journal of the American Statistical Association*, 105: 522-537.

McShane and Wyner (2011) A statistical analysis of multiple temperature proxies: Are reconstructions of surface temperatures over the last 1000 years reliable? *The Annals of Applied Statistics*, 5: 5-44.

Wolkewitz et al. (2010) Two pitfalls in survival analyses of time-dependent exposure: A case study in a cohort of Oscar nominees. *The American Statistician*, 64: 205-211.

Typical sections in a statistics report or paper:

Title

Author(s)

Abstract / executive summary

Keywords

Introduction

Background / literature review

Data description

Methods

Results

Discussion / conclusion

Appendices

References

However, this varies *widely*....

Examples:

Beaudoin & Swartz

Title

Authors

Abstract

Key words

1. Introduction

2. Data analysis

3. Simulation model

4. Bayesian parameter estimation

5. Simulation results

6. Concluding remarks

References

Berrocal et al.

Title

Authors

Abstract

Key words

1. Introduction

2. Data and methods

2.1 Road maintenance problem

2.2 Statistical model

2.3 Model fitting

2.4 Choice of training period

2.5 Generating forecasts

3. Results

4. Discussion

References

McShane & Wyner

Title

Authors

Abstract

1. Introduction

2. Controversy

3. Model evaluation

3.1 Introduction

3.2 Preliminary evaluation

3.3 Validation against pseudo-proxies

3.4 Interpolation versus extrapolation

3.5 Variable selection: True proxies versus pseudo-proxies

3.6 Proxies and local temperatures

3.7 Discussion of model evaluation

4. Testing other predictive methods

4.1 Cross-validated RMSE

4.2 Temperature reconstructions

5. Bayesian reconstruction and validation

5.1 Model specification

5.2 Comparison to other models

5.3 Model reconstruction

5.4 Comparison to other reconstructions & posterior calculations

5.5 Model validation

6. Conclusion

Acknowledgements

References

Wolkewitz et al.

Title

Authors

Abstract

Key words

Abbreviations

1. Motivation

The Cohort of Oscar nominees

2. Multiscale modeling

3. Illustration of length and time-dependent bias

Lexis diagram

Risk sets

Length bias

Time-dependent bias

4. Statistical methods

Multistate models

Death hazard ratios

Software

5. Results

Length bias in the Oscar study

Time-dependent bias in the Oscar study

6. Discussion

Appendix

Basic concepts of survival theory

Right-censoring and left-truncation

Hazard function

Estimating the (cumulative) hazard

Hazard-based analyses in multistate models

References

Some observations:

The number of sections and levels in the hierarchy is related to the length of the paper; readers need more “signposts” in a longer paper. But it’s rare to have a nested structure with more than two levels.

The material often determines its own logical structure, although some topics must come before others (e.g. methods before results).

It helps to highlight difficult and/or important topics in their own section or sub-section. Informative names make it easy for readers to find what they need.

Important: It is very rare that technical reports or journal articles are read through from start to finish, in order. Most readers employ non-linear reading and skimming.

A more typical ordering for reading would be

Title, Abstract

Skim Figures, Introduction, Results, Conclusions

Skim Methods

Perhaps decide to read in detail

A study of managers at Westinghouse showed 100% read the abstract, 60% read the introduction, 50% read the conclusion, 15% read the body of the report, and 10% read the appendix.

You can use non-linear reading to your advantage by

1. Writing the abstract carefully to reflect the main findings and highlight what is interesting.
2. Choosing your figures to reflect your main points, making sure they are clear, and including important conclusions drawn from the figure in a caption as well as in the text.
3. Anticipating what the reader will be looking for, and making these things into specific section headings.

You might also consider writing in a non-linear order.

For example, my typical order of writing a paper is something like

Figures & tables

(Outlining)

Methods & results

Intro & discussion

(More analysis)

Revisions to above

Abstract

It really helps to start writing **BEFORE** your analysis is completely done. This helps highlight exactly what is still needed.

Figures and tables should be numbered and referred to by number in the text. The caption should include any details needed to interpret the plot.

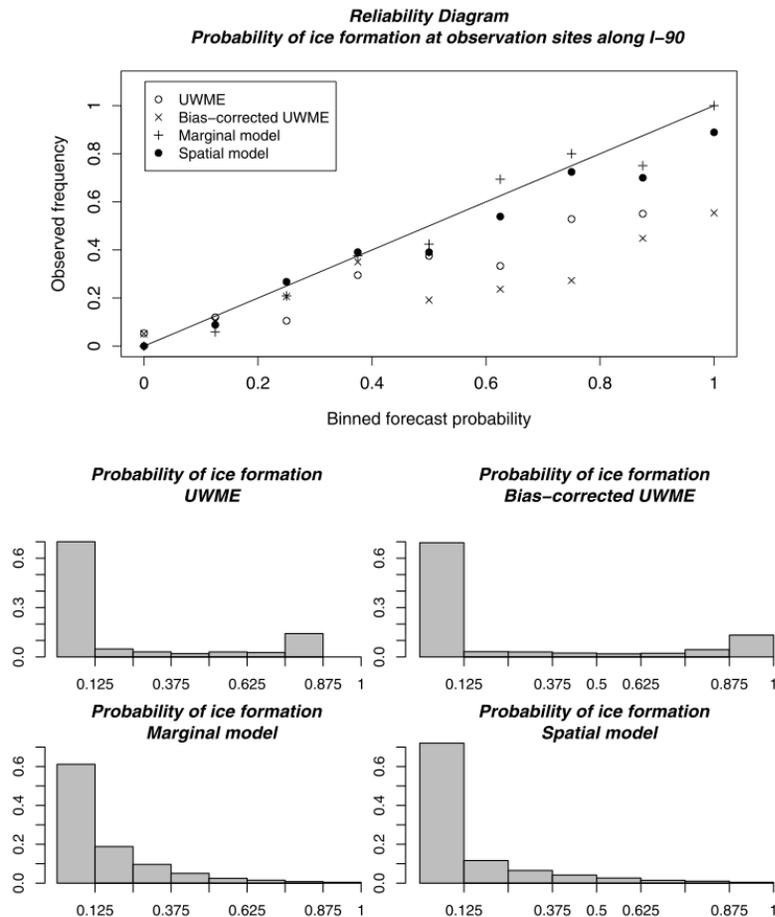


Table 2 shows that for the probability that ice forms at individual locations, both of our probabilistic forecasting methods substantially outperformed the raw and the ensemble forecasts. Figures 7 and 8 show reliability diagrams for the probabilistic forecasts of ice formation provided by the marginal model, the spatial model, the UWME, and the bias-corrected UWME. Our methods were superior to both ensembles, which were very underdispersed and particularly unreliable for high predicted probabilities. At individual locations along I-90, the marginal and spatial methods performed similarly, as expected. However, for probability forecasts of the spatial aggregate “ice formation along \mathcal{I} ,” the spatial model was more reliable: the marginal model tended to overestimate the probability of ice formation.

Figure 7. Reliability diagram for probability forecasts of ice formation at observation sites by the UWME, the bias-corrected UWME, the marginal model, and the spatial model, for the 2003–2004 and 2004–2005 winter seasons. Histograms of the forecast probabilities are also shown.

Note that plots in journal articles or reports often write in the caption how different lines or symbols are to be interpreted. This convention is not universal (see e.g. the figure on the previous slide, which used a legend).

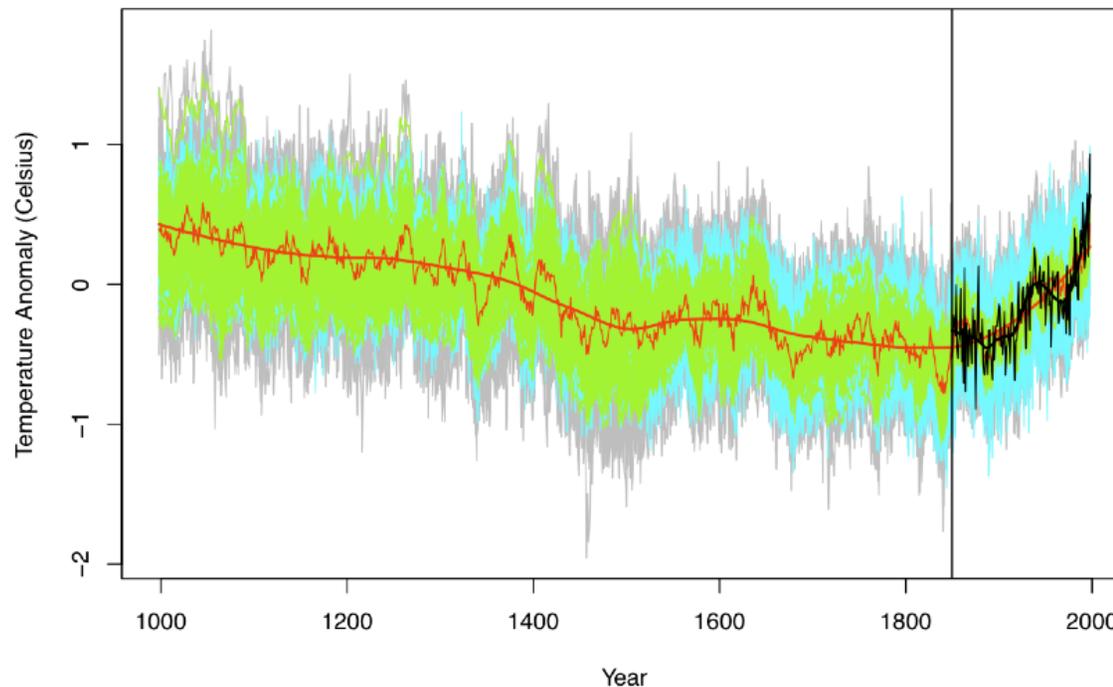


FIG 16. Backcast from Bayesian Model of Section 5. CRU Northern Hemisphere annual mean land temperature is given by the thin black line and a smoothed version is given by the thick black line. The forecast is given by the thin red line and a smoothed version is given by the thick red line. The model is fit on 1850-1998 AD and backcasts 998-1849 AD. The cyan region indicates uncertainty due to ϵ_t , the green region indicates uncertainty due to $\bar{\beta}$, and the gray region indicates total uncertainty.

Mathematical expressions are sometimes separated from the rest of the text for emphasis, particularly if they are long. They are numbered only if they are referred to at some later point in the text.

Our model assumes that individual team scoring rates arise from a population of league-wide scoring rates

$$\theta_{is} \sim \text{Gamma}(a_s, b_s),$$

where the parameters a_s and b_s have independent prior distributions

$$\begin{aligned} a_s &\sim \text{Gamma}(\alpha_{as}, \beta_{as}) && \text{and} \\ b_s &\sim \text{Gamma}(\alpha_{bs}, \beta_{bs}). \end{aligned} \tag{3}$$

The hyperparameters α_{as} , β_{as} , α_{bs} , and β_{bs} , $s = 1, \dots, \tilde{m}$, are set in an empirical Bayes fashion by considering the sample scoring rates. The Gamma hyperparameters are chosen such that $\alpha_{as} > 1$ and $\alpha_{bs} > 1$. We impose a Uniform(0, 1) prior for f in (2) according to the widely held belief that home-ice confers an advantage. The primary parameter of interest in our

When **presenting the data**, ask yourself:

Who collected the data and is it publicly available? Often a URL is included. If the data were collected as part of an experiment, describe how this was conducted.

What are the variables (only those included in the report) and their units? Summary statistics and plots may be used to highlight important features.

Is there missing data? If so, how much, and how will this be treated?

Is there a particular date range or location(s) associated with the data?

When **writing about calculations**, you should describe them (giving mathematical expressions where appropriate), but generally do **NOT** include code or names of specific variables or functions in the report.

Examples:

“We calculated MLEs for each of $B = 1000$ simulated datasets according to (5). Histograms of their distributions are shown in Figure 3.”

“All calculations were carried out in the R programming language, with additional functions from the fields package. Annotated code is contained in the Appendix.”