▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

## (Super small) Data or Super (Small Data)? Using (Inter)Individual Estimates to Get the Most from the Least

#### a talk about data with surprisingly little data

#### Phillip M. Alday Matthias Schlesewsky Ina Bornkessel-Schlesewsky

EXAL+, 16 January 2016

◆□▶ ◆□▶ ◆ 臣▶ ◆ 臣▶ ○ 臣 ○ の Q @

## Introduction

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## A few small assumptions (making my bias explicit)

0. Languages differ, but there is one human brain, i.e. core computational mechanism, processing them all.

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

## A few small assumptions (making my bias explicit)

- 0. Languages differ, but there is one human brain, i.e. core computational mechanism, processing them all.
- 1. The goal of psycho- and neurolinguistic research is to develop models of human language processing, both in perception and production.
- 2. Quantitative models are generally more specific and precise and thus more desireable than qualitative models.

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

## A few small assumptions (making my bias explicit)

- 0. Languages differ, but there is one human brain, i.e. core computational mechanism, processing them all.
- 1. The goal of psycho- and neurolinguistic research is to develop models of human language processing, both in perception and production.
- 2. Quantitative models are generally more specific and precise and thus more desireable than qualitative models.

Research in practice entails developing and testing models both in their general shape and in their specific parameter values.

Introduction

Prior information

Adaptive testing

Principles and Parameters

◆□▶ ◆□▶ ◆ 臣▶ ◆ 臣▶ ○ 臣 ○ の Q @

## Why are we here?

Introduction

Prior information

Principles and Parameters

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

## Why are we here? Because of (Indo-)Germanic hegemony

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## Why are some languages less studied?

- lack of interest
- lack of money
- lack of appropriate research infrastructure
- lack of "accessibility"
- dominance of Western European (esp. Germanic) and Chinese institutions

#### Arabic and other understudied languages ...

Why is Arabic understudied?



▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## Arabic and other understudied languages ...

#### Why is Arabic understudied?

- widely spoken
- geopolitically important
- extensive written tradition
- interesting for a number of reasons

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

## Arabic and other understudied languages ...

#### Why is Arabic understudied?

- widely spoken
- geopolitically important
- extensive written tradition
- interesting for a number of reasons

We don't have much neurocognitive data on Arabic, but we do have a lot of other data.

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

I can't believe I'm saying this but maybe more traditional linguistics has something to teach us ...

Introduction

Prior information

Adaptive testing

Principles and Parameters

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三三 - のへぐ

## Prior information

## Traditional linguistics

#### Typology and "grammars"

- understand, acknowledge and model language variation and diversity (that's why we're here today)
- deep relationship between processing strategies and structures (*neurotypology*)
- genetic relationships and language change provide insights into local optima (Zipf's Law – coming to Evolang!)

Adaptive testing

Principles and Parameters

Unless you're using evidence-based procedures, I can't hear a word you're saying.



## (Slightly less) Traditional linguistics

#### Offline ratings

- acceptability judgments
  - use subject's own behavioral response in predicting neuro-response, not just as trial-rejection criterium
  - norming and pre-test data
- interpretations
- used e.g. in research on Competition Model from MacWhinney and Bates
- qualitative rankings can be made quantitative

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## **Corpus linguistics**

#### Frequency

- frequency plays a crucial role in many models of language processing and change
- accurate estimate and not just subjective impression allows for testing the desired hypothesis
- important whether
  - frequency influences processing strategy
  - processing constraints drive frequency
  - all of the above

## **Corpus linguistics**

Existence

- null evidence always a problem
- as well as existence of speaker-produced errors



http://www.tofugu.com/wp-content/uploads/2015/05/cat-cartoon-grammar.jpg

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

#### **Computer linguistics**

# Asking whether a machine can think is like asking whether a submarine can swim – E. Dijkstra

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

### **Computer linguistics**

# Asking whether a machine can think is like asking whether a submarine can swim – E. Dijkstra

Nonetheless, there are some common principles at work (some form of computation; buoyancy).

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

## Computer linguistics

#### Stochastic incrementality

- modern research parsers have a stochastic component trained on large corpora
- this provides an ideal way to combine models and corpus data
- incremental parsers (e.g. shift-reduce parsers) additionally provide a model of the effects of incrementality

#### Natural language processing

- focus on extraction and expression of meaning
- instead of mapping to a platonic abstraction

## Computer linguistics

We gave the monkeys, the banancis2 ... because they, were hungry. ... because they were ripe.

http://specgram.com/CLIII.4/08.phlogiston.cartoon.zhe.html

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ●の00

#### Iterative updates

- incorporating prior information as model parameters/covariates, in structure of the model, or even as a prior in the Bayesian sense – is an iterative process.
- the basic logic of scientific research is based on successive approximations to reality.
- approximations are improved by removing error, and so we should focus on the things we can't yet explain.

## Iterative updates in practice

- using a priori estimates from previous research can provide good model fit
- rapid quantitative empirical estimates of model free parameters possible
- both methods yield similar fit
- but the latter is preferred (less sensitive to researcher whim and post-hoc massaging)

	Df	AIC	logLik
quantized from previous research	11	395190	-197584
estimates from a single-subject	11	395223	-197600
pooled estimates from four subjects	11	395252	-197615

Alday et al (2014, Neuroinformatics), Alday et al (2015, Ling. Vanguard)

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## Rapid acquisition of parameter values

- simple behavorial design
- verb-final constructions in German: dass NP1 NP2 V
- NPs: random selection of fully crossed design for CASE x ANIMACY x NUMBER
- V: randomly assigned number, all transitive
- canonical, non-canonical and ungrammatical sentences (doubel accusatives, no agreement)
- task: pick the actor

#### Rapid acquisition of parameter values



Alday et al (2015), Linguistic Vanguard イロト イヨト イヨト イヨト

æ

#### Rapid acquisition of parameter values



Alday et al (2015), Linguistic Vanguard イロト イヨト イヨト イヨト

æ

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

## Rapid acquisition of parameter values

- We managed to estimate three parameters in a model in less than 30 minutes.
- But for those parameters, we reached a point of diminishing returns well before the end of the experiment.
- We could have used our time better.

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

## I can't believe I'm saying this but maybe more spam has something to teach us ...

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三三 - のへぐ

## Adaptive testing

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ □ のへぐ

## "Unbiased" Sampling

purely random sampling is the foundation of experimentation

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

## "Unbiased" Sampling

- purely random sampling is the foundation of experimentation
- except when it's not:
  - inclusion and exclusion criteria for participants and trials
  - constructed, controlled stimuli
  - etc.
- all of this done either before or after the experiment, but not during
- (ideally the criteria are determined before, even if applied after!)
- we choose our stimuli, participants and trials so that we can focus on hypotheses of interest

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

## "Unbiased" Sampling

- purely random sampling is the foundation of experimentation
- except when it's not:
  - inclusion and exclusion criteria for participants and trials
  - constructed, controlled stimuli
  - etc.
- all of this done either before or after the experiment, but not during
- (ideally the criteria are determined before, even if applied after!)
- we choose our stimuli, participants and trials so that we can focus on hypotheses of interest
- with modern computing, we can focus in real time!

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

### Bias isn't always a bad thing in statistics

- we can exclude conditions dynamically in an experiment
- we prioritize conditions based on variance and convergence rate
- priorities may change during an experiment
  - sudden improvement in one parameter
  - complex interaction leads to a previous estimate becoming worse
- priority has an upper bound to prevent "false" dominance via high variance / low convergence
  - poorly chosen model
  - trivial or non-existent effect
  - free variation (both inter- and intra-individual)

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

### Bias isn't always a bad thing in statistics

- we can exclude conditions dynamically in an experiment
- we prioritize conditions based on variance and convergence rate
- priorities may change during an experiment
  - sudden improvement in one parameter
  - complex interaction leads to a previous estimate becoming worse
- priority has an upper bound to prevent "false" dominance via high variance / low convergence
  - poorly chosen model
  - trivial or non-existent effect
  - free variation (both inter- and intra-individual)
- biased estimators can outpeform unbiased estimators!

▲□▶ ▲□▶ ▲□▶ ▲□▶ ▲□ ● ● ●

## Precedent and prior formulation

- such adaptive techniques are widely used in spam filtering
- medical research has special rules for clinical trials with strong effects apparant early on
- philosophical relationship to MCMC, especially considerations of ideal chain length
- relationship to filtering remove distracting, "less interesting" portions of the signal
- indeed this can be viewed as an example of Kalman filtering or recursive Bayesian estimation

#### Rapid acquisition, again



Alday et al (2015), Linguistic Vanguard  $\langle \Box \rangle$   $\langle \Box$ 

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三三 - のへぐ

## Principles and Parameters

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三三 - のへぐ

#### (No, I'm not going to stop the bad word play any time soon.)

▲□▶ ▲□▶ ▲ □▶ ▲ □▶ □ のへぐ

## Principles I

Use the data we already have available

- previous research
- the Internet

▲ロ ▶ ▲周 ▶ ▲ 国 ▶ ▲ 国 ▶ ● の Q @

## Principles I

Use the data we already have available

- previous research
- the Internet

#### Focus on model specification and comparison

- reduce free parameters
- maximize predictive power
- strive for parsimony
- use modern statistical tools (mixed-effects models, Bayesian methods)

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

## Principles II

#### Maximize entropy and optimize data collection

- between and within participants
- between and within experiments

▲□▶ ▲□▶ ▲ 三▶ ▲ 三▶ 三 のへぐ

## Principles II

#### Maximize entropy and optimize data collection

- between and within participants
- between and within experiments

#### Hire a programmer and statistician

• (I have reasonable rates, but a very long wait list ...)

▲□▶ ▲□▶ ▲□▶ ▲□▶ ■ ● ●

## From (super small) data to super (small data)

#### Stand on the shoulders of giants giant data sets

Big data provides an empirical, quantitative starting point.

#### Nucleation point for explosive, iterative growth

Recursion is not just something to study in language, but something for studying language! Research, participants and trials are not context-free.

Introduction

Prior information

Adaptive testing

Principles and Parameters

◆□▶ ◆□▶ ◆三▶ ◆三▶ 三三 のへぐ

#### In other words...

## Use your computer and your statistics like it's 2016!

▲□▶ ▲□▶ ▲□▶ ▲□▶ □ のQで

### Questions?



http://www.economist.com/blogs/prospero/2015/05/johnson-polyglots

Slides, all published code, and some code in progress available at: palday.bitbucket.org