## **Statistical Theory and Practice**

Nancy Reid University of Toronto

June 8 2023



# 10th International Purdue Symposium on Statistics

Demystifying Data Science via Statistics: Theory and Applications
JUNE 6TH - 9TH, 2023





#### Statistical Decision Theory and Related Topics III, Volume 1



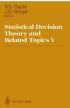
#### Statistical Decision Theory and Related Topics IV: volume 1



Springer New York, 1988 - Mathematics - 418 pages

The Fourth Purdue Symposium on Statistical Decision Theory and Related Topics was held at Purdue University dur period, June 15-20, 1986. The symposium brought together many prominent leaders and ununger researchers in stat decision theory and related seess. The 85 imited nevers and discussions presented at the summation are collected two-volume work. The papers are grouped into a total of seven parts. Volume I has three parts: Part 1 -Conditioning More v





#### Sixth Purdue International Symposium on Statistics

Statistics was held on June 17,23, 1998, at Punitue University in West Lafavette, Indiana, USA, The co- and contributed paper sessions. The workshop chairs were Shanti S. Gunta and Mary Ellen Book. "Statistical Genetics: An Intentisciplinary Future"

was held on June 21,23 consisting of invited and The symnosium beans with simultaneous workshops - contributed talks as well as poster sessions "Interfaces Between Major Statistical Paradiems"

and "Nutrition and Statistics" on June 17-19. The Statistics', page 4

Tonics" was held lone 19,21 and included invited See "Surb Purble International Symposium on

conference "Statistical Decision Theory and Related



Seventh Purdue International Symposium on Statistics

Information and Call for Panary for The Components of the Symposium

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Material Decision Theory and Related Space (T-17 Ann)

from and Winter Models for Large Data Sets (1) have

Statistical Consultancy (St. 64 hors)





### 60 years; 10 Symposia

- · Common themes:
  - · design of experiments
  - · selective inference
  - probability
  - · remarkable list of speakers
- · Theoretical frameworks
  - · statistical decision theory
  - Bayesian inference
  - asymptotic theory
- increasing emphasis on applications, increasing breadth

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#### • 2023

- probability
- · variable selection
- high-dimensional time series
- · dimension reduction
- multiple testing
- · federated learning
- data integration
- deep learning
- spatial statistics
- foundations
- Bayesian nonparametrics
- · actuarial science and risk
- ..

- · mobile health
- neuroscience
- 'omics
- Covid
- privacy
- · biomedicine
  - •••

Purdue Sy Chocolate 8 The story that never dies

Nudging A meta-analysis in behavioural science	e Mertens et al., PNAS 2021
<b>BWAS</b> Brain-wide association studies	Marek et al., Nature 2021
<b>Drought</b> Climate change attribution	World Weather Attribution report, April 2023
Women Co-authorship and gender	Ross et al., Nature 2022
Cash Effect of policy on mortality	Richterman et al., Nature 2023

Neuro Risk of Parkinson disease

Faces Human perception

Richterman et al., Nature 2023

Wardle et al., PNAS 2021

Globe & Mail. June 5

**Nudging** A meta-analysis in behavioural science many small effects, publication bias, Bayes/freq

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Faces Human perception

sign test, regression, computational modelling

data collection is complex, generally well-described, likely very impactful

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 e.g. Women: "the analytical linked dataset ... is constructed from three sources"

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- increasing emphasis on effect estimates and standard errors
   easier in some fields
   e.g. Drought, Cash, Nudging, Women

- Statistics needs a healthy interplay between theory and applications
  - theory meaning foundations, rather than theoretical analysis of specific techniques
- Foundations? "A solid base, on which rests a large structure"

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- must be continually tested against new applications
- "the practical application of general theorems is a different art from their establishment by mathematical proof"

(F 1958 SMRW 13th ed)

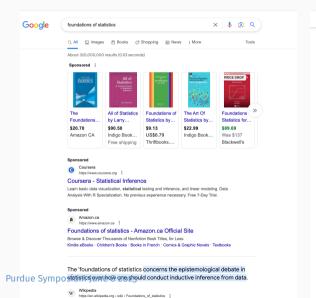
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@DanielBolnick

"The data strongly support my hypothesis"



Google	foundations of statistics × 4	· • Q
	What are the basic foundations of statistics?	~
	What are the 4 fundamental elements of statistics?	~
	What are fundamentals of statistics?	~
	What are the 5 types of statistics?	- Feedback
	Amazon.ca https://www.amazon.ca > Foundations-Statistics-Leona	STATISTICS
	ScienceDirect https://www.sciencedirect.com > topics > mathematics > f ;  Foundations of Statistics - an overview The foundations of statistics have changed and evolved with time. The fairly use of probations of statistics analysis was closely tied to the development of	ability
	Goodreads https://www.goodreads.com i book i show i 163905   The Foundations of Statistics by Leonard J. Savage	
	Foundations of Statistics is a book that discusses concepts about statistics. Inductive Inference & Axiomatic Concept of Probability are some of the early  **** Rating: 3.9 - 7 reviews - US\$9.99	
	edX https://www.edx.org > course > fundamentals-of-statist :	
	Fundamentals of Statistics Fundamentals of Statistics Develop a deep understanding of the principles that under	



# Statistics, Foundations

D. A. S. Fraser
University of Toronto

- I. Background
- II. Overview
- III. Probability Model
- IV. Statistical Model
- V. Statistical Theory
- VI. Foundations
- VII. Principles
- VIII. Likelihood Asymptotics



By the mid 1950s there were substantial criticisms of the decision theory approach; in particular, there had been a major failure of the theory to produce reasonable statistical procedures for a broad range of problems.

In the mid-1950s publications by Fisher (1956) and Savage (1954) substantially altered the directions of statistics and opened wide areas for development. Fisher proposed insightful methods based on the earlier view of examining the model data combination  $(\mathcal{D}, \mathcal{M})$ . Savage favored the Bayesian approach emphasizing the use of personal priors to represent the latent views of the investigator concerning possible values for the parameters.

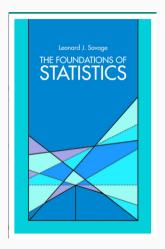
Both these directions opened new opportunities to a discipline that had become partially paralyzed by the decision-theoretic approach and by its inability to produce answers for wide-ranging problems.

#### VI. FOUNDATIONS

The foundations of statistics have changed and evolved with time. The early use of probability for statistical analysis was closely tied to the development of the least squares method, a widely used technique dating from Laplace and earlier. The Bayesian approach also comes from this same earlier period. Neither could be viewed at that time as an Purdue Symane in the grant from the purdue Symane in the grant from the purdue Symane in the grant from the

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... publications by Fisher (1956) and Savage (1954) substantially altered the directions of statistics and opened new opportunities to a discipline that had become partially paralyzed ...



It is unanimously agreed that statistics depends somehow on probability. But, as to what probability is and how it is connected with statistics, there has seldom been such complete disagreement and breakdown of communication since the Tower of Babel. Doubtless, much of the disagreement is merely terminological and would disappear under sufficiently sharp analysis.

Link

• probability, analysis, applied mathematics

modelling

• probability, analysis, applied mathematics

modelling

• Bayes, Neyman, Fisher

approaches to inference

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- $\bullet$  interpretation of p-values, confidence regions, credibility intervals, likelihood ratios
- role of sufficiency, ancillarity, conditioning, asymptotic theory
- sparsity, causality, assumption-free/lean inference, stability, prediction, decisions

## **Linking theory with practice**

**Nudging** A meta-analysis in behavioural science many small effects, publication bias, Bayes/freq **BWAS** Brain-wide association studies spatial correlation, many small effects, publication bias **Drought** Climate change attribution climate models, subgroup analyses, predictions Women Co-authorship and gender linear regression, binary outcome, confounding **Cash** Effect of policy on mortality Poisson regression, binary outcome, DiD **Neuro** Risk of Parkinson disease logistic regression, retrospective cohort study Faces Human perception sign test, regression, computational modelling **Nudging** 

### Statistics in the news

#### **Economist, July 29**





"plagued by publication bias"

Mertens et al. 2021



RESEARCH ARTICLE | PSYCHOLOGICAL AND COGNITIVE SCIENCES | 8



# The effectiveness of nudging: A metaanalysis of choice architecture interventions across behavioral domains

THIS ARTICLE HAS BEEN UPDATED

# The response

LETTER | JULY 19, 2022 | 🔮

No reason to expect large and consistent effects of nudge interventions

Barnabas Szaszi, Anthony Higney, [...] Elizabeth Tipton

VIEW THE ORIGINAL ARTICLE:

The effectiveness of nudging: A meta-analysis of choice architecture interventions across behavioral domains

THIS ARTICLE HAS A REPLY:

Reply to Maier et al., Szaszi et al., and Bakdash and Marusich: The present and future of choice

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Left-truncated effects and overestimated meta-analytic means

Jonathan Z. Bakdash and Laura R. Marusich

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No evidence for nudging after adjusting for publication bias

Maximilian Maier, František Bartoš, [...] Eric-Jan Wagenmakers

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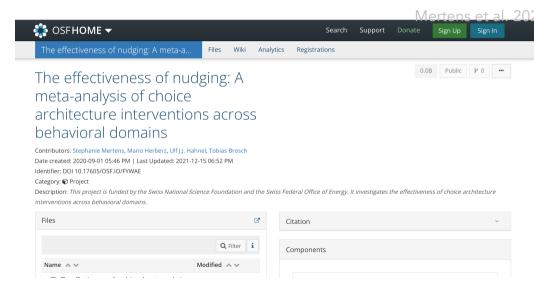
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## In fairness



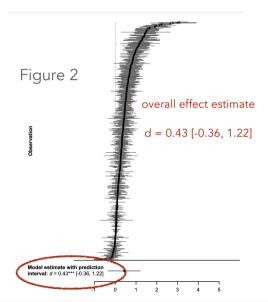
## Some details

#### "Materials and methods"

Mertens et al. 2021

- 440 estimates of effect size: (treatment control mean)/(estimated std error)
- 212 unique publications; sometimes several tmts with the same control
- Random effects to accommodate this
- Additional fixed effects (moderators) for secondary analysis —
   types of interventions; behavioural domain; study characteristics
- Publication bias assessed by plotting standard error vs effect size

Egger's tes



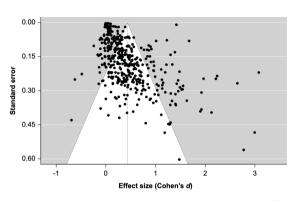


Fig. 3. Funnel plot displaying each observation as a function of its effect size and SE. In the absence of publication bias, observations should scatter

Standard error increases with effect size

## The letters

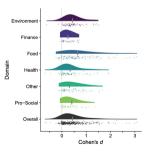
1. Maier et al. — publication bias not correctly taken into consideration

"A newly-proposed bias-correction technique — robust Bayesian meta-analysis avoids an 'all-or-none' debate over whether or not publication bias is 'severe' "

2. Szaszi et al. — the average effect size is not very informative, given the variation between studies

"Even after adjusting for publication bias, the effects  $\dots$  vary considerak

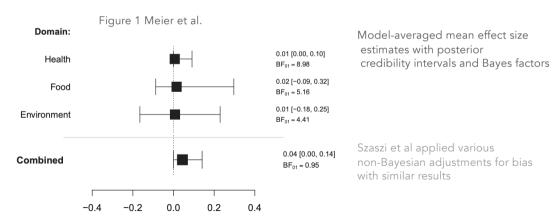
3. Bakdash & Maurisch — estimated effects in studies are right-skewed



## **Foundations**

#### 1. Maier et al. — publication bias not correctly taken into consideration

"A newly-proposed bias-correction technique — robust Bayesian meta-analysis avoids an 'all-or-none' debate over whether or not publication bias is 'severe' "



**Drought** 



Science

# Deadly African drought not possible without climate change, study finds



Warming climate made long rains twice as likely to underdeliver, World Weather Attribution calculates

Thomson Reuters · Posted: Apr 27, 2023 8:43 AM EDT | Last Updated: April 27



News

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Opinion

Video

News | Drought

# Global warming made Horn of Africa drought possible: WWA study

World Weather Attribution scientists say rising greenhouse gas emissions made the nearly 3-year drought at least 100 times more Puttled Symposium June 8 2023

#### Climate Change Made East Africa's Drought 100 Times as Likely, Study Says

The findings starkly show the misery that the burning of fossil fuels, mostly by rich countries, inflicts on societies that emit almost nothing by comparison.





A water well near the town of Kelafo in Ethiopia, one of the nations hit hardest by the drought. Eduardo Soteras/Agence France-Presse — Getty Images

# Human-induced climate change increased drought severity in Horn of Africa

Link

- 1. Joyce Kimutai, Kenya Meteorological Department, Nairobi, Kenya
- 2. Clair Barnes, Grantham Institute, Imperial College London, UK
- 3. Mariam Zachariah, Grantham Institute, Imperial College, London, UK
- 4. Sjoukje Philip, Royal Netherlands Meteorological Institute (KNMI), De Bilt, The Netherlands
- 5. Sarah Kew, Royal Netherlands Meteorological Institute (KNMI), De Bilt, The Netherlands
- 6. Izidine Pinto, Royal Netherlands Meteorological Institute (KNMI), De Bilt, The Netherlands
- Piotr Wolski, Climate System Analysis Group, University of Cape Town, Cape Town, South Africa
- Gerbrand Koren, Copernicus Institute of Sustainable Development, Utrecht University, Utrecht, the Netherlands
- Gabriel Vecchi, Department of Geosciences, Princeton University, Princeton, NJ 08544,
   USA, High Meadows Environmental Institute, Princeton University, Princeton, NJ 08544,

The data Kimutai et al. 2023

- observational data, 3 sources
  - 1. global daily rainfall & temperature
  - 2. daily rainfall
  - 3. monthly rainfall

 $0.5^{\circ} \times 0.5^{\circ}$ , 1979 – infra-red, "SoA", 1981 –

1981-2014

• 4-year smoothed mean surface temperature

proxy for anthropogenic climate change

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· climate modelling data, 4 sources

1. combine 12 global and 8 climate models: resolution  $0.44^{\circ}$ 

2. combine 5 global and 4 climate models: resolution 0.22°

3. atmosphere-ocean coupled GCMs (two)

4. sea-surface temperature forced ensemble, high resolution

29 sims

10 sims

10/3 simulations

11 simulations

response is log<sub>10</sub> (monthly rainfall) in 2021 and 2022

and  $\log_{10}(PET)$  — potential evapotranspiration

covariates are global temperature anomaly, and ENSO index

El Nino-Southern Oscillation

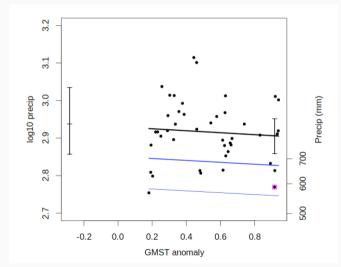
• "As a measure of anthropogenic climate change we use smoothed GMST"

Global Mean Surface Temperature

- "Methods for observational and model analysis ... and synthesis are used according to the World Weather Attribution Protocol"

  Philip et al. 2020
  - 1. trend using observational data
  - 2. find climate models consistent with 1.
  - 3. compare predictions from 1. and 2.
  - 4. synthesize results in 3. to provide conclusions

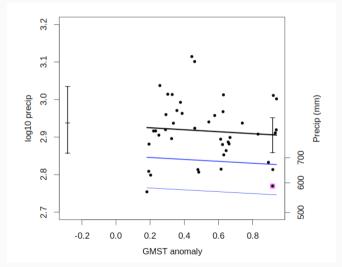
The results



 rainfall decreasing with increasing temperature

but not much

The results Kimutai et al. 2023

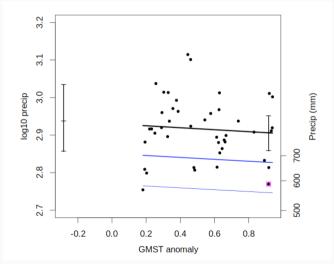


 rainfall decreasing with increasing temperature

but not much

2022 rainfall is about a
 1 in 20 year event

The results



 rainfall decreasing with increasing temperature

but not much

- 2022 rainfall is about a 1 in 20 year event
- 2022 drought about
   2 times more likely under climate change
- uncertainty 0.1 to 360

## Climate Change Made East Africa's Drought 100 Times as Likely, Study Says

The findings starkly show the misery that the burning of fossil fuels, mostly by rich countries, inflicts on societies that emit almost nothing by comparison.

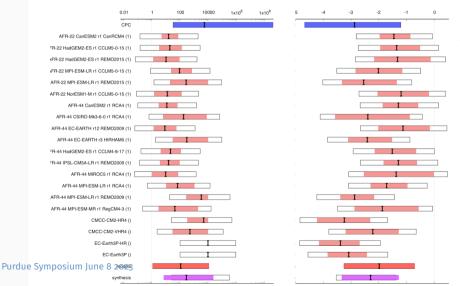




- change the response to SPEI rainfall adjusted for evaporation
- consider 'long rains' and 'short rains' separately
   MAM, OND
- combine model simulation results with observational data
- the first is most important; 2022 drought now **5500** times more likely uncertainty 32 to  $4 \times 10^8$

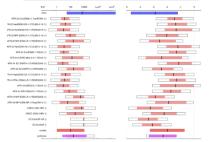
#### Combining climate simulations and data

#### (a) Probability Ratio (left) and Intensity change (right) for current vs. 1.2degC cooler climates

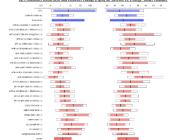


precipitation adjusted for PET

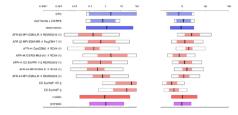
#### (a) Probability Ratio (left) and Intensity change (right) for current vs. 1.2degC cooler climates



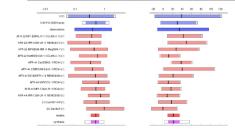
#### (a) Probability Ratio (Ioff) and Intensity change (right) for current vs. 1.2deef; cooler climates



#### (a) Probability Ratio (left) and Intensity change (right) for current vs. 1.2degC cooler climates



#### (a) Probability Ratio (left) and Intensity change (right) for current vs. 1.2degC cooler climates



#### The theory

- extrapolation beyond observations
   extreme value modelling
- assigning uncertainty to combined results components of variance
- ratios of estimated probabilities unbounded confidence intervals
- joint modelling of precipitation and evapotranspiration copula modelling

weighted average?

#### Behrens-Fisher

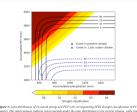


Figure 1. And chardware of 4.4 mark proxy and 1921 ords corresponding SPEI drops in complication (CPC decount). Two city converse belowers and proxy and and the paint decounter indicate mere character effect and decounter indicate the same return periods of a 1.2°C confer character. The header converse represent of affirmed toward of decounter or proxy and approxy point indicate in 1.2°C confer character. The header converse represent affirmed toward of decounter converse and approxy point indicate in 2.25 and a 2.25 and a

## nature

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Explore content > About the journal > Publish with us >
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nature > articles > article
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Article Open Access | Published: 22 June 2022

#### Women are credited less in science than men

Matthew B. Ross, Britta M. Glennon, Raviv Murciano-Goroff, Enrico G. Berkes, Bruce A. Weinberg & Julia I. Lane ☑

```
Nature 608, 135–145 (2022) | Cite this article
```

97k Accesses | 57 Citations | 3111 Altmetric | Metrics

The Data

- "finding 'what isn't there' from 'what is there' is a fundamental problem in statistics"
- · analytic data
  - 118 campuses send deidentified data to U Michigan
  - tracks spending on personnel for each research project
  - 57 campuses with complete data 2013-2016
  - identify teams: PI, faculty, PDF, PhD, UGrad, Research Staff
  - identify publications
  - identify gender, job titles, scientific fields, patents, ...
- 9800 teams with 129,000 team members
- 39,000 articles; 18m 'potential authorships', 367,000 actual authorships

scientific articles

+ survey + qualitative analysis payroll, all funding sources

weighting of 1 for each person

Web of Science

Purdue Symposium June 8 2023

- response attribution rate  $=\frac{\# \text{ actual authorships}}{\# \text{ potential authorships}} = \operatorname{pr}(\text{ attribution })$
- covariates date of publication, number of days worked in the team, calendar time, position in the team, team's PI
- model

$$P(\text{ named }) = \beta_0 + \beta_1 woman + \beta^T \text{ covariates } + \text{ error}$$

#### **Empirical strategy**

The empirical approach was to estimate linear regressions using a model of the form

$$P[\text{named}_{i,t,e,l}|... = \beta_0 + \beta_1 \text{woman}_{i,e} + X_{i,e} + M_{i,t} + O_{i,e} + \text{Team}_{i,l} + \mu_{i,t,e,l}$$
(1)

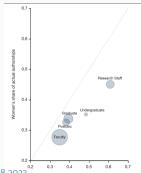
overall attribution rate 3.1%;

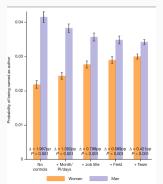
attribution rate for men 4.23%; attribution rate for women 2.12%

includes patents

difference smaller when covariates included

but still statistically significant





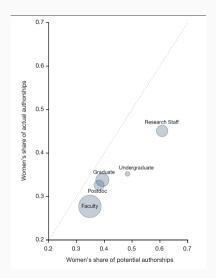


Figure 1

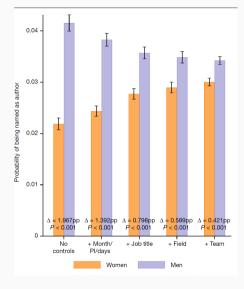


Figure 2

- · linear regression but the response is a proportion
- logistic function is pretty linear for  $p \in (0.2, 0.8)$
- but these p's  $\in$  (0.01, 0.04)

also many t-tests comparing ps

- there's a paper for that!
  - On the linear in probability model for binary data

    Battey, Cox & Jackson 2019

    least squares estimate is more robust, coefficient directly interpretable

    less efficient, incorrect for observations out of range
- possibly more concerning: what is the unit of observation? potential authorship? article? team? are the standard errors correct?

**Faces** 

NEUROSCIENCE

## Americans tend to assume imaginary faces are male

Why people perceive faces in inanimate objects as male by default is still unclear



**Faces** 



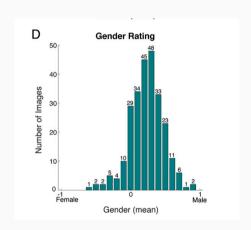
... Faces Wardle et al. PNAS 2021













# Will AI really change everything? Not likely

Although tools like ChatGPT can astonish us with their output. they are not operating anywhere near human intelligence

JOSEPH WILSON

OPINION

PhD candidate in linguistic anthropology at the University of Toronto

told, "Are you ready?"

o you have AI fatigue yet? Not a day goes by without hreathless commentary on the increasing power of artificial-intelligence models A del-

blindly optimistic, claiming that AI will magically solve everything from climate change to the opioid crisis, or they are darkly dystopian, warning us that AI and destroy humanity.

themselves "warn" people of the existential threats AI could pose. as they did in an open letter recently calling for a pause in development, it functions as a marketing campaign. The tech companies are essentially congratulating each other for creating something too good. Google's chief executive. Sundar Pichai. has called AL without irony, a technology "more profound than

fire or electricity." The public doesn't know what to believe and they're worried. newly released poll conducted by uge of new apps and services Innovative Research Group for promises to disrupt everything the 2023 Provocation Ideas Fes-Purduefrond Realth Safet follow to leduca 2 Owal shows that 47 per cent of tion. "The future is here." we are Canadians are more concerned

than excited about the increased

"future-proof your career" or "hecome AI literate"

what we read about AI is hype. In could escape its silicon chains. AI tools will probably give us slightly better-written spam in Even when AI developers our inboxes and reams of crappy. machine-generated websites. Real, life-saying applications are indeed possible in fields such as health care and agriculture, but they'll be hard to spot amid all the junk. Although tools such as ChatGPT and Midjourney are fun to play with and can astonish us with their output, they are not operating anywhere near human intelligence. They are essentially performing a clever parlour trick.

The reason we are astonished by their output is because, as a species, we're gullible. We tend to read human characteristics into any pattern that even mildly resembles a human. We see faces in electrical sockets and spot human silhouettes in evening shadows. We feel bad for a discarded

heightened empathy is one of the ways technology companies The reality is that most of have captured the public's attention in recent months. OpenAI the near term, this new crop of launched ChatGPT (which generates text) and DALL-E (which generates images) online and for free so the public could play around with them. It let the public work itself into a frenzy as they identified characteristics in the programs that were previously thought to be exclusively human: reason, humour, emotion, creativity. But generative AI can do none of these things. It has the form of human expression but no content.

> The technology that runs under the hood of these tools is not fundamentally new. The mathematical models have changed in recent years, and new chips are been slow to act. making computation cheaper and more efficient, but ChatGPT only functions like a powerful autocomplete feature. Trained on an enormous amount of data. the model predicts which words

capital and technological knowhow in the hands of very few billionaires. As such, the field of Al is desperately in need of regulation. This is necessary not because tech companies might unleash a mathematical model that will suddenly become conscious and take over the world, but for the very real, boring reasons that have always existed: so they don't take advantage of poorly paid temp workers, or refuse calls to be transparent with their algorithms, or flood social media with misinformation, or violate copyright laws by scraping the web for data without the permission of its owners. Sadly, these are things that Big Tech is already doing, and governments have

sky-high, further concentrating

Fear, as populist politicians and headline writers know well. is best evoked by appealing to the unknown. Whether it's the fear of AI-gone-rogue or the fear of falling behind in the race to "The reason we are astonished by their output is because, as a species, we're gullible. We tend to read human characteristics into any pattern that even mildly resembles a human"

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## **Linking theory with practice**

**Nudging** A meta-analysis in behavioural science

many small effects, publication bias, Bayes/freq

**Drought** Climate change attribution

climate models, subgroup analyses, predictions

Women Co-authorship and gender

linear regression, binary outcome, confounding

Faces Human perception

sign test, regression, computational modelling

#### ... What are the foundations of statistics?

• probability, analysis, applied mathematics

modelling

· Bayes, Neyman, Fisher

approaches to inference

nature of uncertainty

epistemic, empirical

· nature of induction

belief functions, inferential models

- $\bullet$  interpretation of p-values, confidence regions, credibility intervals, likelihood ratios
- role of sufficiency, ancillarity, conditioning, asymptotic theory
- sparsity, causality, assumption-free/lean inference, stability, prediction

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- $\bullet$  interpretation of p-values, confidence regions, credibility intervals, likelihood ratios
- realistic estimates of precision, complex dependencies, subgroup analyses
- out-of-sample predictions, sensitivity to assumptions

**Observations** mine

- statistical "workflows" seem to be emerging in different disciplines
  - e.g. Drought "A Protocol for probabilistic extreme event attribution analysis"

Philip et al 2020, Adv. Stat. Clim. Met. Ocean

· e.g "Writing statistical methods for ecologists"

Davis & Kay 2023, Ecosphere

- tutorial-type articles in scientific journals
  - Annals of Thoracic Surgery the statistician's page
  - J Am Medical Association Guide to Statistics and Methods
  - Nature Methods Points of Significance
- "open data" observed in the breach
  - Drought "Almost all the data are available via the KNMI Climate Explorer"
  - Women "datasets generated ... are available at the Virtual Data Enclave Repository"
- sleuthing is hard

## **Congratulations to the Department**

## THANK YOU



#### **Communication!** — This just in



#### **Communication!** — This just in





Purdue Symposium June 8 2023

Conclusions. The US life expectancy disadvantage began in the 1950s and has steadily worsened over