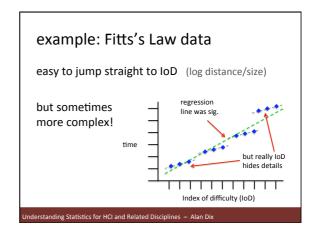


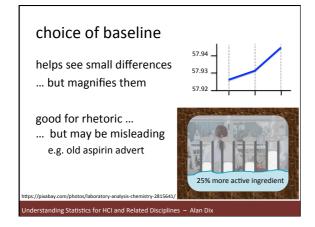
#### look at the data

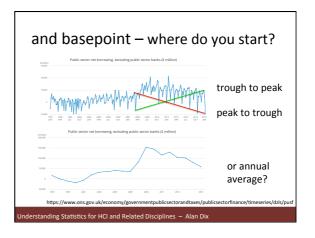
eyeball the raw data are there anomalies, extreme values? does it match your model?

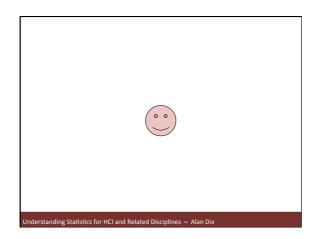
but remember randomness can be misleading data is not truth!

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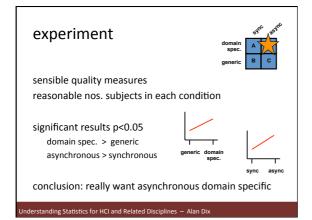


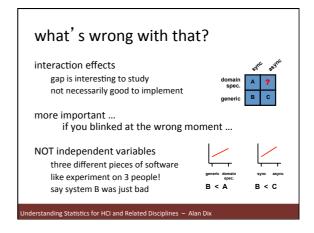
what have you really shown

- · think about the conditions
  - are there other explanations for data?
- · individual or population
  - small #of groups/individuals, many measurements
  - sig. statistics => effect reliable for each individual
  - but are individuals representative of all?
- · systems vs properties

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# a little story ... BIG ACM conference — 'good' empirical paper looking at collaborative support for a task X three pieces of software: A — domain specific software, synchronous B — generic software, synchronous C — generic software, asynchronous domain spec. generic B C Understanding Statistics for HCI and Related Disciplines — Alan Dix





### what went wrong?

borrowed psych method

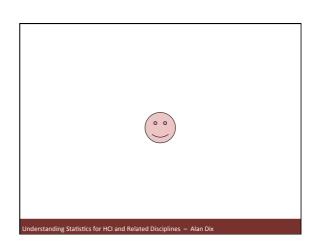
... but method embodies assumptions single simple cause, controlled environment

interaction needs ecologically valid experiment multiple causes, open situations

what to do?

understand assumptions and modify

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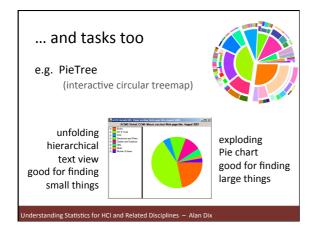


# diversity - individual/task

good for not just good

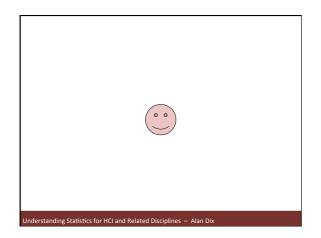
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#### don't just look at average! e.g. overall system A lower error rate than system $\ensuremath{\mathsf{B}}$ but ... system B better for experts error rate system B number system A 30 3.7% 7.4% novice 10 9.6% 2.7% expert 5.2% 6.2%



more important to know
who or what
something is good for

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#### mechanism

#### quantitative and statistical

what is true end to end phenomena

#### qualitative and theoretical

why and how mechanism

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## generalisation

empirical data at best interpolate

understanding mechanism allows: extrapolation application in new contexts

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# example: mobile font size

#### early paper on fonts in mobile menus:

well conducted experiment statistically significant results conclusion gives best font size

#### but ... a menu selection task includes:

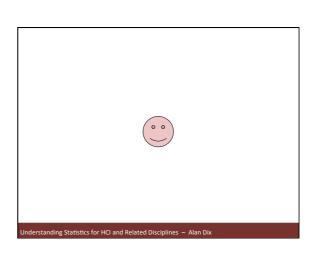
1. visual search (better big fonts)

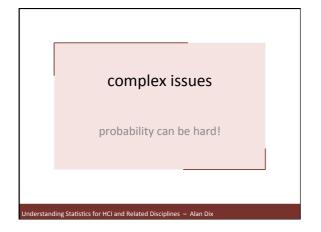
2. if not found scroll/page display (better small fonts)

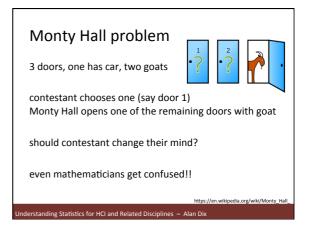
3. when found touch target (better big fonts)

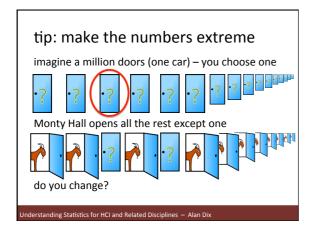
no single best size – the balance depends on menu length, etc.

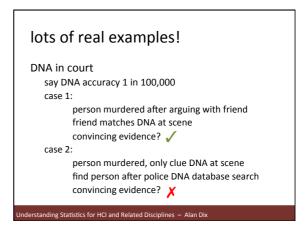
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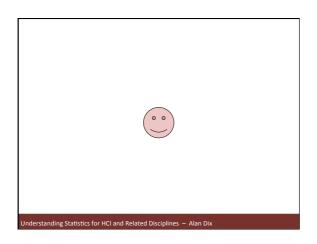














# building for the future

- repeatability
  - comparisons more robust than measures
  - RepliCHI
- meta analysis
- data publishing and open science

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