What can we learn from quantitative teaching assistant evaluations?

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Why study TAs?

- They improve student success in CS1 (Wilson, 2001), especially for minorities (Roberts, 1995)
- TAs give 46% of our contact hours in 1st & 2nd-year CS
- Issues of TA quality
Why TA evaluations?

- Evaluations are an imperfect measure of teaching

- Nevertheless, good TA evaluations encourage TAs (Bomotti, 1994)

- TAs are new teachers, flexible in their approach to teaching (Muzaka, 2009)
  - They are also hungry for feedback (Patitsas, 2012)
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Introduction
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Work distribution
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Grades
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Implications

Studying TAs

- TAs are linked to:
  - student retention (O’Neal, 2007)
  - student grades (Paul, 2010)
  - and provide role models to our students (Patitsas, 2012)

- In this talk, we will be examining a number of such matters; however, we’ll be leaving more questions than answers.
We ask...

What would you like to know about your TAs?

What do you think your TAs should know about their teaching?
The five criteria are evaluated on a 5-point Likert scale, and are:

- Well prepared (W.P.)
- Helpful (Help.)
- Considerate of students (Consid.)
- Easily understood (E.U.)
- An effective instructor (E.I.)
Delving into findings...

- We acquired 231 anonymized TA evaluations
- And we looked at the correlations between those criteria
- Pop quiz: what is the correlation coefficient needed to be a “relationship” in the social sciences?
### Evaluation Criteria, contd.

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<tr>
<td>Well prepared</td>
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<tr>
<td>Helpful</td>
<td>0.8</td>
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<tr>
<td>Considerate</td>
<td>0.7</td>
<td>0.8</td>
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<tr>
<td>Easily understood</td>
<td>0.7</td>
<td>0.8</td>
<td>0.6</td>
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<tr>
<td>Effective instr.</td>
<td>0.8</td>
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- All are $p < 0.001$
- Coarse-grained evaluations!
Methods

1. Literature review: what variables are worth testing?
2. Look at those variables
3. Threshold for significance: $p < 0.001$
4. $n=231$, so large sample size

So what did we learn?
First and second year courses at UBC have:
- 3 hours of lecture / week
- One 2 or 3 hour lab session / week
- In some cases, a 1 hour tutorial

Lab sections:
- Have been 20 and 25 students

While all TAs work roughly the same number of hours, work distribution varies
Negative correlation: between how many lab sections a TA taught and their TA evaluations ($r=-0.4$).

- TAs with 1-2 lab sections had better evaluations than TAs with 3
- TAs with 4 lab sections a week performed worst of all

Open question: why? Boredom effect? Contact hours are more tiring?
We found that paired TAs received similar evaluations.

**Open question:** *Do better TAs bring their partners up? Or do students just rank pairs as a unit? Is the lab only as good as how well the TAs work together?*
Context: Types of TAs

- Graduate Teaching Assistants (GTAs):
  - Monthly GTAs
    - The TA work is part of their promised funding.
  - Hourly GTAs
    - They apply for the position, we select the hourly GTAs.
- Undergraduate Teaching Assistants (UTAs):
  - They apply for the position, select the UTAs we hire.
That UTAs performed better in teaching first-year has been found at other institutions (Mendenhall, 1983)

Open question: Why? Course-specific knowledge vs. conceptual knowledge?
Previous work found biases against female instructors in evaluations (Wachtel, 1998)

We found no overall differences between how female and male TAs were ranked by their students.
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Experience

No statistically significant link between course-specific experience and TA evaluations.

Threats to validity: we could only match TAs within a course, and only within the time period we sampled.
While there is evidence that TAs improve as teachers over time, the effect of their experience may be counterbalanced by more “sterner” teaching (Patitsas, 2012).

Previous work on teaching evaluations has found no link between experience and evaluations (Wachtel, 1998).
The literature finds that TAs affect students’ performance (Paul, 2010).

We found no link between students’ final grades and their TAs’ evaluations.

This does not discount the likelihood that a TA has an effect on their students’ performance.

Whatever effect a TA has on their students is not captured by these quantitative evaluations.
Previous literature has found that a negative lab atmosphere will contribute to students’ decision to take fewer science courses (O’Neal, 2007).

We found a weak, correlation between low-ranked TAs’ evaluations and how many more computer science classes their students took.

Open question: *are TAs scaring away students, or are unengaged students are rating TAs down?*
Open questions

**How** can we provide TAs with effective formative assessment?

To what extent are quantitative student evaluations useful? What questions would be better?

**How** could qualitative evaluations be better harnessed?

**Why** are UTAs ranked better than GTAs?

Causality: student retention, TA evals

Causality: pairing
Implications and Suggestions

- TA evals are coarse-grained, need more info to make award/hiring decisions
- Probably not very useful to the TAs; part of the TA quality issue?
- Put UTAs on first-year
- Keep TA workloads reasonable
- Pair your TAs
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