FLS 6441 - Methods III: Explanation and Causation

Week 4 - Survey and Lab Experiments

Jonathan Phillips

April 2019

► Why survey and lab experiments?

- ► Why survey and lab experiments?
 - 1. Treatments we cannot administer in reality

- Why survey and lab experiments?
 - Treatments we cannot administer in reality
 - 2. Random treatment assignment not permitted in reality

- Why survey and lab experiments?
 - Treatments we cannot administer in reality
 - Random treatment assignment not permitted in reality
 - 3. Outcome measurements that are hard to take in reality

- Why survey and lab experiments?
 - 1. Treatments we cannot administer in reality
 - 2. Random treatment assignment not permitted in reality
 - 3. Outcome measurements that are hard to take in reality
 - 4. Reduce variation in context and noise in data

- ▶ Why survey and lab experiments?
 - Treatments we cannot administer in reality
 - 2. Random treatment assignment not permitted in reality
 - 3. Outcome measurements that are hard to take in reality
 - 4. Reduce variation in context and noise in data
 - 5. To generalize beyond specific situations to abstract behaviour

Section 1 Lab Experiments

► Treatment Assignment: Same as a Field Experiment

- ► Treatment Assignment: Same as a Field Experiment
- ➤ **Treatment**: Not a manipulation of real world political or economic processes, but establishing controlled 'lab' conditions

- ► Treatment Assignment: Same as a Field Experiment
- ➤ **Treatment**: Not a manipulation of real world political or economic processes, but establishing controlled 'lab' conditions
 - The advantage: Control over context helps isolate mechanisms

- ► Treatment Assignment: Same as a Field Experiment
- Treatment: Not a manipulation of real world political or economic processes, but establishing controlled 'lab' conditions
 - The advantage: Control over context helps isolate mechanisms
 - ► The disadvantage: Can we generalize to the real world?

▶ Problems generalizing from the lab:

- ▶ Problems generalizing from the lab:
 - ▶ Hawthorne effect: Lab context influences behaviour, social desirability bias

- ▶ Problems generalizing from the lab:
 - ► **Hawthorne effect**: Lab context influences behaviour, social desirability bias
 - Context effects: The real-world always provides more information, more history

- ▶ Problems generalizing from the lab:
 - Hawthorne effect: Lab context influences behaviour, social desirability bias
 - Context effects: The real-world always provides more information, more history
 - ► **Process effects**: People care *how* decisions are made
 - Selection effects: Actors in specific roles are rarely representative samples, 'WEIRD' or pro-social lab subjects

► The lab differs from the field

- ► The lab differs from the field
 - ► The stakes

- ► The lab differs from the field
 - ► The stakes
 - ► The norms (specific norms of being an experimental subject)

- ► The lab differs from the field
 - ► The stakes
 - ► The norms (specific norms of being an experimental subject)
 - The degree of scrutiny

- ► The lab differs from the field
 - The stakes
 - ► The norms (specific norms of being an experimental subject)
 - The degree of scrutiny
 - ► The sample of individuals

- ► The lab differs from the field
 - The stakes
 - The norms (specific norms of being an experimental subject)
 - The degree of scrutiny
 - The sample of individuals
 - ► The degree of anonymity

► Lab experiments are *inherently* imperfect (Levitt and List 2006)

- ► Lab experiments are *inherently* imperfect (Levitt and List 2006)
- Decisions change depending on the degree of scrutiny

- ► Lab experiments are *inherently* imperfect (Levitt and List 2006)
- Decisions change depending on the degree of scrutiny
 - "You tip more when you're on a date"

- ► Lab experiments are *inherently* imperfect (Levitt and List 2006)
- Decisions change depending on the degree of scrutiny
 - "You tip more when you're on a date"
 - Social norms are activated, eg. treating one-shot games like repeated games

- ► Lab experiments are *inherently* imperfect (Levitt and List 2006)
- Decisions change depending on the degree of scrutiny
 - "You tip more when you're on a date"
 - Social norms are activated, eg. treating one-shot games like repeated games
 - Scrutiny alters who wants to make a decision as well as the decision they make

- ► Lab experiments are *inherently* imperfect (Levitt and List 2006)
- Decisions change depending on the degree of scrutiny
 - "You tip more when you're on a date"
 - Social norms are activated, eg. treating one-shot games like repeated games
 - Scrutiny alters who wants to make a decision as well as the decision they make
 - Subjets use cues (heuristics) to draw on 'similar' situations from the real world

 Many studies find more cooperation in the lab than in the real world

- Many studies find more cooperation in the lab than in the real world
 - Scrutiny increases cooperation

- Many studies find more cooperation in the lab than in the real world
 - Scrutiny increases cooperation
 - Anonymity reduces cooperation

- Many studies find more cooperation in the lab than in the real world
 - Scrutiny increases cooperation
 - Anonymity reduces cooperation
 - ► That's interesting in itself! We can manipulate the degree of scrutiny/anonymity etc.

- Many studies find more cooperation in the lab than in the real world
 - Scrutiny increases cooperation
 - Anonymity reduces cooperation
 - That's interesting in itself! We can manipulate the degree of scrutiny/anonymity etc.
- ► Lab experiments may be generalizable where norms/morality is less important (???)

Lab-in-the-Field Experiments

► In a natural setting with the target population

Lab-in-the-Field Experiments

- ▶ In a natural setting with the target population
- Standardized, artificial treatment and measurement

Lab-in-the-Field Experiments

► Habyarimana et al (2007)

- ► Habyarimana et al (2007)
- ► Existing consensus: Ethnic diversity -> **Less** public goods provision

- ► Habyarimana et al (2007)
- Existing consensus: Ethnic diversity -> Less public goods provision
- ▶ But how? Theories:

- ► Habyarimana et al (2007)
- Existing consensus: Ethnic diversity -> Less public goods provision
- ▶ But how? Theories:
 - Preferences in-group fairness
 - Technology social networks permit identification and sanctioning
 - Strategy Selection choose to cooperate more often

- ► Lab-in-the-field
- ► Population: Ugandans
- ► **Sample**: 300 people in a diverse area with few public goods
- ► Treatment/Control: Various Games
- ► **Treatment assignment**: Random assignment to co-ethnic/non-co-ethnic

- ▶ **Preferences** dictator game between self and two others
 - ► No bias towards co-ethnics

- ▶ **Preferences** dictator game between self and two others
 - ► No bias towards co-ethnics
- ► **Technology 1, productivity** teamwork in a puzzle requiring communication
 - Co-ethnic teams don't perform any better

- ▶ **Preferences** dictator game between self and two others
 - No bias towards co-ethnics
- ► **Technology 1, productivity** teamwork in a puzzle requiring communication
 - Co-ethnic teams don't perform any better
- ► **Technology 2, social networks** Can you find a co-ethnic in the town faster than a non-co-ethnic?
 - ► Yes (43% vs 28% success)

- ▶ **Preferences** dictator game between self and two others
 - ► No bias towards co-ethnics
- ► **Technology 1, productivity** teamwork in a puzzle requiring communication
 - Co-ethnic teams don't perform any better
- ► **Technology 2, social networks** Can you find a co-ethnic in the town faster than a non-co-ethnic?
 - ► Yes (43% vs 28% success)
- ➤ **Strategy Selection** Does anonymity for the sender in the dictator game make a difference?
 - ► Yes offer more to co-ethnics when offerers believe they can be seen

- ▶ **Preferences** dictator game between self and two others
 - ► No bias towards co-ethnics
- ► **Technology 1, productivity** teamwork in a puzzle requiring communication
 - Co-ethnic teams don't perform any better
- ► **Technology 2, social networks** Can you find a co-ethnic in the town faster than a non-co-ethnic?
 - ► Yes (43% vs 28% success)
- ➤ **Strategy Selection** Does anonymity for the sender in the dictator game make a difference?
 - ► Yes offer more to co-ethnics when offerers believe they can be seen

► Conclusion: Norms and Networks allow co-ethnics to provide more public goods

- ► Conclusion: Norms and Networks allow co-ethnics to provide more public goods
 - ...But where are the public goods here?
 - Are public goods organized by voluntary contributions or coercive central authority?
 - ► Is this true of all parts of Kampala? Uganda? All ethnic groups?

Section 2

► Treatment occurs within the survey questionnaire

- ► Treatment occurs within the survey questionnaire
 - Different versions of the questionnaire randomly applied

- ► Treatment occurs *within* the survey questionnaire
 - Different versions of the questionnaire randomly applied
 - Not a field experiment: Still an artificial context

- ► Treatment occurs *within* the survey questionnaire
 - Different versions of the questionnaire randomly applied
 - Not a field experiment: Still an artificial context
 - Not a lab experiment: People not brought to a single location or interacting

► Easy and cheap to implement

- ► Easy and cheap to implement
- ► Can be targeted to our real population of interest

- ► Easy and cheap to implement
- Can be targeted to our real population of interest
- ▶ But a limited range of 'weak' treatments possible

- ► Easy and cheap to implement
- Can be targeted to our real population of interest
- ▶ But a limited range of 'weak' treatments possible
- And outcome measurement normally takes place immediately

► Humans are subject to psychological and social influences

- Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects

- ► Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ▶ But we can also use these influences to our advantage:

- Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ▶ But we can also use these influences to our advantage:
- Framing Experiments how responses vary to question content

- Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ▶ But we can also use these influences to our advantage:
- Framing Experiments how responses vary to question content
- 2. Priming Experiments to measure the effect of a prime on a response

- ► Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ► But we can also use these influences to our advantage:
- Framing Experiments how responses vary to question content
- Priming Experiments to measure the effect of a prime on a response
- 3. Anchoring vignettes to increase reliability in measurement

- ► Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ▶ But we can also use these influences to our advantage:
- Framing Experiments how responses vary to question content
- 2. Priming Experiments to measure the effect of a prime on a response
- 3. Anchoring vignettes to increase reliability in measurement
- 4. List Experiments to reduce social desirability bias in measurement

- ► Humans are subject to psychological and social influences
- ► These create threats to estimating causal effects
- ► But we can also use these influences to our advantage:
- Framing Experiments how responses vary to question content
- Priming Experiments to measure the effect of a prime on a response
- 3. Anchoring vignettes to increase reliability in measurement
- 4. List Experiments to reduce social desirability bias in measurement
- 5. Conjoint Experiments to measure preferences

► How much do details in the question affect our responses?

- ► How much do details in the question affect our responses?
- ► Eg. A female citizen goes to her representative for help. How likely is she to receive help?

- ► How much do details in the question affect our responses?
- ► Eg. A female citizen goes to her representative for help. How likely is she to receive help?
- ► Eg. A male citizen goes to her representative for help. How likely is she to receive help?

► People responded differently to being told 'A' instead of 'B'. How do we interpret this?

- ► People responded differently to being told 'A' instead of 'B'. How do we interpret this?
 - 1. They were told 'A'/'B' by a survey enumerator do they trust them? What is the source?

- ► People responded differently to being told 'A' instead of 'B'. How do we interpret this?
 - 1. They were told 'A'/'B' by a survey enumerator do they trust them? What is the source?
 - 2. Are 'A'/'B' things that they would hear in the real world? In what context?

- ► People responded differently to being told 'A' instead of 'B'. How do we interpret this?
 - 1. They were told 'A'/'B' by a survey enumerator do they trust them? What is the source?
 - 2. Are 'A'/'B' things that they would hear in the real world? In what context?
 - 3. What are they communicating in their answer? To impress the surveyor? Who is listening to their answers?

- ► People responded differently to being told 'A' instead of 'B'. How do we interpret this?
 - 1. They were told 'A'/'B' by a survey enumerator do they trust them? What is the source?
 - 2. Are 'A'/'B' things that they would hear in the real world? In what context?
 - 3. What are they communicating in their answer? To impress the surveyor? Who is listening to their answers?
 - 4. What is at stake in the answer? Are there any actual consequences? Will they have to defend their answer in the community later? 'Cheap talk'

► A prior task that creates an unconscious bias in subsequent answers

- ► A prior task that creates an unconscious bias in subsequent answers
- Eg. We remind half of respondents about national Independence Day
- Then ask what they think about immigration
- Allowing us to measure the effect of 'nationalism' on migration attitudes

► Within/Between Survey Experiments

- ► Within/Between Survey Experiments
- ▶ Between: Treated and Control are different people

- ► Within/Between Survey Experiments
- ► Between: Treated and Control are different people
- ► Within: Treated and Control measures from the same person
 - But aren't these different 'units'??

- ► Within/Between Survey Experiments
- ► Between: Treated and Control are different people
- ► Within: Treated and Control measures from the same person
 - ▶ But aren't these different 'units'?? Yes!

- ► Within/Between Survey Experiments
- ► Between: Treated and Control are different people
- ► Within: Treated and Control measures from the same person
 - But aren't these different 'units'?? Yes!
 - But the time difference is usually just a few minutes, so maybe more plausible

- ► Within/Between Survey Experiments
- ► Between: Treated and Control are different people
- ► Within: Treated and Control measures from the same person
 - But aren't these different 'units'?? Yes!
 - But the time difference is usually just a few minutes, so maybe more plausible
 - More problematic is contamination

► The entire point of survey experiments is that the questions we ask change the answers we get

- ► The entire point of survey experiments is that the questions we ask change the answers we get
- ► So the answers to every question depend on the previous questions

- ► The entire point of survey experiments is that the questions we ask change the answers we get
- ► So the answers to every question depend on the previous questions
- ► Usually affects all respondents equally

- ► The entire point of survey experiments is that the questions we ask change the answers we get
- ► So the answers to every question depend on the previous questions
- Usually affects all respondents equally
- But survey experiments that vary across respondents might change ALL subsequent responses

3. Anchoring Vignettes

3. Anchoring Vignettes

► Survey experiments are valuable for measurement

- ► Survey experiments are valuable for measurement
 - ▶ Most survey responses are biased to impress the researcher

- Survey experiments are valuable for measurement
 - Most survey responses are biased to impress the researcher
 - Social desirability bias has differential effects across respondents and topics
 - Most people say they recycle, even though they do not

- Survey experiments are valuable for measurement
 - Most survey responses are biased to impress the researcher
 - Social desirability bias has differential effects across respondents and topics
 - Most people say they recycle, even though they do not
 - ► Rich people lie more than poor people

- Survey experiments are valuable for measurement
 - Most survey responses are biased to impress the researcher
 - Social desirability bias has differential effects across respondents and topics
 - ► Most people say they recycle, even though they do not
 - ► Rich people lie more than poor people
- ► List experiments make individual responses *invisible* to the researcher

- Survey experiments are valuable for measurement
 - Most survey responses are biased to impress the researcher
 - Social desirability bias has differential effects across respondents and topics
 - ► Most people say they recycle, even though they do not
 - ► Rich people lie more than poor people
- ► List experiments make individual responses *invisible* to the researcher
- Knowing this, hopefully the respondent answers more accurately

- Survey experiments are valuable for measurement
 - Most survey responses are biased to impress the researcher
 - Social desirability bias has differential effects across respondents and topics
 - Most people say they recycle, even though they do not
 - ► Rich people lie more than poor people
- ► List experiments make individual responses *invisible* to the researcher
- Knowing this, hopefully the respondent answers more accurately
- ► Gonzalez-Ocantos et al (2010) list experiment on vote-buying

I am now going to read out a list of activities. Please count the number of these activities that you have done in the past one year. Please do not tell me WHICH activities you have done, only the TOTAL NUMBER of them:

- ▶ Voted
- ► Attended a Town Hall Meeting
- Travelled to the State Capital

I am now going to read out a list of activities. Please count the number of these activities that you have done in the past one year. Please do not tell me WHICH activities you have done, only the TOTAL NUMBER of them:

- Voted
- Attended a Town Hall Meeting
- Been offered a gift, some food or money in exchange for your vote;
- Travelled to the State Capital

- ► Nicaragua 2008 municipal elections (Gonzalez-Ocantos 2012)
- ► **Direct Question**: Have you received a gift or favour in exchange for your vote?
 - ▶ 3%

- ► Nicaragua 2008 municipal elections (Gonzalez-Ocantos 2012)
- ► **Direct Question**: Have you received a gift or favour in exchange for your vote?
 - **▶** 3%
- List experiment:
 - Just the difference in mean responses between treatment and control lists
 - \triangleright 24% = 2.31 2.06

Assumptions:

1. No Liars - People answer honestly in the presence of the sensitive item

- 1. No Liars People answer honestly in the presence of the sensitive item
 - Do respondents really understand anonymity?

- No Liars People answer honestly in the presence of the sensitive item
 - ► Do respondents really understand anonymity?
- 2. No Ceiling effects '4' means my answers are no longer anonymous; instead report '3'

- No Liars People answer honestly in the presence of the sensitive item
 - ► Do respondents really understand anonymity?
- 2. No Ceiling effects '4' means my answers are no longer anonymous; instead report '3'
- 3. No Floor Effects If the control items are rare, respondents may be reluctant to report '1' and choose '0' instead.

- No Liars People answer honestly in the presence of the sensitive item
 - Do respondents really understand anonymity?
- 2. No Ceiling effects '4' means my answers are no longer anonymous; instead report '3'
- 3. No Floor Effects If the control items are rare, respondents may be reluctant to report '1' and choose '0' instead.
- No Design Effects- Presence of the treatment item doesn't affect answers on other items
 - Bias towards a 'reasonable'/central number?

► How do people make choices between many options?

- ► How do people make choices between many options?
- ► Treatments are often 'bundles' of characteristics, but which aspect matters most?

- ► How do people make choices between many options?
- ► Treatments are often 'bundles' of characteristics, but which aspect matters most?
- Also a problem of social desirability bias in which characteristics matter

► Hainmueller et al 2013 - How do attitudes to immigrants depend on immigrant characteristics?

- ► Hainmueller et al 2013 How do attitudes to immigrants depend on immigrant characteristics?
- ► Vary education, profession, language, gender, national origin, etc.

- ► Hainmueller et al 2013 How do attitudes to immigrants depend on immigrant characteristics?
- ► Vary education, profession, language, gender, national origin, etc.
- Profiles
 - Attributes
 - Values

- ► Hainmueller et al 2013 How do attitudes to immigrants depend on immigrant characteristics?
- ► Vary education, profession, language, gender, national origin, etc.
- Profiles
 - Attributes
 - Values
- ► Randomize attribute order to prevent bias

- ► Hainmueller et al 2013 How do attitudes to immigrants depend on immigrant characteristics?
- ► Vary education, profession, language, gender, national origin, etc.
- Profiles
 - Attributes
 - Values
- ► Randomize attribute order to prevent bias
- ► Treatment is the **combination** of attributes the respondent sees

- ► Hainmueller et al 2013 How do attitudes to immigrants depend on immigrant characteristics?
- ► Vary education, profession, language, gender, national origin, etc.
- Profiles
 - Attributes
 - Values
- ► Randomize attribute order to prevent bias
- ► Treatment is the **combination** of attributes the respondent sees
- ► Millions of possible treatments

Please read the descriptions of the potential immigrants carefully. Then, please indicate which of the two immigrants you would personally prefer to see admitted to the United States.

	Immigrant 1	Immigrant 2		
Prior Trips to the U.S.	Entered the U.S. once before on a tourist visa	Entered the U.S. once before on a tourist visa		
Reason for Application	Reunite with family members already in U.S.	Reunite with family members already in U.S.		
Country of Origin	Mexico	Iraq		
Language Skills	During admission interview, this applicant spoke fluent English	During admission interview, this applicant spoke fluent English		
Profession	Child care provider	Teacher Three to five years of job training and experience		
Job Experience	One to two years of job training and experience			
Employment Plans	Does not have a contract with a U.S. employer but has done job interviews	Will look for work after arriving in the U.S.		
Education Level	Equivalent to completing two years of college in the U.S.	Equivalent to completing a college degree in the U.S.		
Gender	Female	Male		

	Immigrant 1	Immigrant 2
If you had to choose between them, which of these two immigrants should be given priority to come to the United States to live?	0	0

On a scale from 1 to 7, where 1 indicates that the United States should absolutely not admit the immigrant and 7 indicates that the United States should definitely admit the immigrant, how would you rate immigrant 1?

Absolutely Not Admit 1	2	3	4	5	6	Definitely Admit 7
0	0	0				

Using the same scale, how would you rate immigrant 2?

Absolutely Not Admit	bsolutely lot Admit					
1	2	3	4	5	6	7

Fig. 1 Experimental design: Immigration conjoint. This figure illustrates the experimental design for the conjoint analysis that examines immigrant admission to the United States.

choice outcomes hereafter. Second, in "rating-based conjoint analysis," respondents give a numerical rating to each profile which represents their degree of preference for the profile. This format is preferred by some analysts who contend that such ratings provide more direct, finely grained information about respondents' preferences. We call this latter two of outcome a ratine outcome.

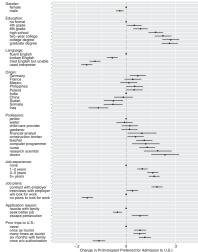


Fig. 3. Effects of immigrant attributes on preference for admission. This plot shows estimates of the effects of the randomly assigned immigrant attributes on the probability of being preferred for admission to the United States. Estimates are based on the regression estimators with clustered standard errors, but represent 95% confidence intervals. The points without horizontal bars denote the attribute value that is the reference category for each attribute.

► Estimating results uses a simple regression of respondent choices on profile attribute-values

- ► Estimating results uses a simple regression of respondent choices on profile attribute-values
- ► But each specific profile (treatment) may arise too rarely to make comparisons of individual attribute-values

- ► Estimating results uses a simple regression of respondent choices on profile attribute-values
- ► But each specific profile (treatment) may arise too rarely to make comparisons of individual attribute-values
 - So this is **not** an Average Treatment Effect

- ► Estimating results uses a simple regression of respondent choices on profile attribute-values
- ► But each specific profile (treatment) may arise too rarely to make comparisons of individual attribute-values
 - So this is **not** an Average Treatment Effect
 - Eg. the effect of gender when age, language etc. are held constant

- ► Estimating results uses a simple regression of respondent choices on profile attribute-values
- ► But each specific profile (treatment) may arise too rarely to make comparisons of individual attribute-values
 - So this is **not** an Average Treatment Effect
 - Eg. the effect of gender when age, language etc. are held constant
 - ► It is an Average Marginal Component Effect

- ► Estimating results uses a simple regression of respondent choices on profile attribute-values
- ► But each specific profile (treatment) may arise too rarely to make comparisons of individual attribute-values
 - So this is **not** an Average Treatment Effect
 - Eg. the effect of gender when age, language etc. are held constant
 - It is an Average Marginal Component Effect
 - Eg. the effect of gender averaging across all possibilities of age, language, etc.

Assumptions:

► We're still assuming people try to answer honestly

Assumptions:

- ► We're still assuming people try to answer honestly
- ► The ordering of attributes does not matter (or is randomized)

Assumptions:

- ► We're still assuming people try to answer honestly
- ► The ordering of attributes does not matter (or is randomized)
- ▶ Profiles are randomized

► How realistic are the responses?

- ► How realistic are the responses?
 - Not a behavioural measure; nothing at stake

- ► How realistic are the responses?
 - Not a behavioural measure; nothing at stake
 - Still some social desirability bias?

- ▶ How realistic are the responses?
 - Not a behavioural measure; nothing at stake
 - Still some social desirability bias?
 - Not like real-world preference-formation process
 - ► Stated preferences vs. Revealed preferences

- ► How realistic are the responses?
 - Not a behavioural measure; nothing at stake
 - Still some social desirability bias?
 - Not like real-world preference-formation process
 - ► Stated preferences vs. Revealed preferences
- Hainmueller et al 2014 compare conjoint responses to a Swiss referendum

- ► How realistic are the responses?
 - Not a behavioural measure; nothing at stake
 - Still some social desirability bias?
 - Not like real-world preference-formation process
 - ► Stated preferences vs. Revealed preferences
- ► Hainmueller et al 2014 compare conjoint responses to a Swiss referendum
- Citizens voted on specific naturalization applicants (Really!)

Figure S11: Effects of Applicant Attributes on Opposition to Naturalization Request (Unweighted Survey Sample)

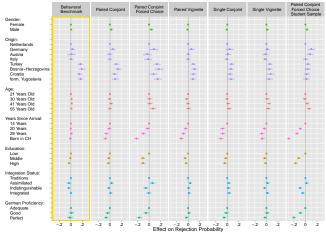


Figure shows point estimates (dots) and corresponding, cluster-robust 95 % confidence intervals (horizontal lines) from ordinary least squares regressions. The dots on the zero line without confidence intervals denote the reference category for each applicant attribute.

- ► But note the conjoint method still hugely under-estimated the overall rejection rate
- ▶ 21% versus 37% in reality

Section 3

Generalizability

Generalizability

- 1. Generalizability of our Sample
- 2. Generalizability of our Context
- 3. Generalizability of our Treatment
- 4. Generalizability of our Outcome