## Evaluating affirmative action when college applications are endogenous

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Setting and data

Application responses

Shifted in vs. out 0000000

### Background

- Affirmative action (AA) policies by design change composition of student body
  - Contested
  - Potential for efficiency losses
- Application channel commonly assumed away in strategy-proof settings
  - However, AA might change preferences over programs, or there might be search frictions, or behavioural costs to applying,...
  - May matter for identification as well as policy

This paper

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## Study effects of an AA policy that seeks to reduce gender imbalance within majors

- 1. Document application responses to AA
- 2. Estimate effects of the policy
  - Academic and labor market outcomes
  - Shifted-in vs shifted-out

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## Norway's Centralized Admission System

- Centralized platform for all applications to higher public education
  - Submit rank-ordered list of up to 15 preferences
  - Specific program at specific institution, e.g. Physics at UiO
- Gale-Shapley Deferred Acceptance: Slots allocated based on application score
  - Largely comprised of high school GPA, with extra points in certain cases
  - Exact cutoff unpredictable
- Strategy proof: No incentive not to list true preferences

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## The AA policy: Gender points

• Give targeted gender an advantage by boosting their application scores

$$AppScore_{ipy} = 10 GPA_i^{HS} + 2 Target_{ip}Active_{py}$$

- Regulated by the Ministry of Education
- Requirement: One gender clearly underrepresented
  - Women: Selected STEM programs
  - Men: Selected health programs

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#### Data

- Application data 1999-2018
  - Rank-ordered lists of preferences
  - Application scores and cutoffs
  - Offers and enrollment
  - Gender point policies over time
- $\bullet$  Student trajectories + grades in higher education
- Background variables
- Labor market outcomes

# 1. Application responses

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## Do college applications respond to AA?

- Utilize within-program changes in use of gender points between 1999 and 2018
  - Flagship technical university abolishes gender points (women) for about half of civil engineering programs in 2016
  - Three nurse programs introduce gender points (men) in 2018
- Estimate effect on gender balance in pool of applicants using DiD
  - Control group: Applicants to comparable programs that do not change their policy

$$\textit{Target}_{i} = \delta_{y} + \theta_{p} + \sum_{t=-5}^{T} \beta_{y} \mathbf{1}_{\{t=y\}} \textit{Change}_{p} + \epsilon_{i}$$



Abolishment of AA for women (civil engineering)



2. What is the effect on candidates shifted in, vs. shifted out, by AA?

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#### **Parameters of interest**

The cost of AA measured in terms of an outcome Y can be written as

$$\sum_{i} Y_i(AA=1) - Y_i(AA=0)$$

Assuming that switching on AA without changing the treatment does not change the outcome, this can be written in terms of potential outcomes  $Y^1, Y^0$ 

$$=\sum_{i}Y_{i}^{1}D_{i}(AA=1)+Y_{i}^{0}(1-D_{i}(AA=1))-Y_{i}^{1}D_{i}(AA=0)-Y_{i}^{0}(1-D_{i}(AA=0))$$

$$=\sum_{i}[Y_{i}^{1}-Y_{i}^{0}]\Big[D_{i}(AA=1)-D_{i}(AA=0)\Big]$$

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#### **Parameters of interest**

Groups by treatment status:

Shifted-in  $D_i(AA = 1) = 1, D_i(AA = 0) = 0$ Shifted-out  $D_i(AA = 1) = 0, D_i(AA = 0) = 1$ Unaffected (admitted)  $D_i(AA = 1) = 1, D_i(AA = 0) = 1$ Unaffected (rejected)  $D_i(AA = 1) = 0, D_i(AA = 0) = 0$  $= \sum_i [Y_i^1 - Y_i^0] \left[ D_i(AA = 1) - D_i(AA = 0) \right]$ 

$$= \sum_{i \in \text{In}} [Y_i^1 - Y_i^0] - \sum_{i \in \text{Out}} [Y_i^1 - Y_i^0]$$
$$= N_{\text{In}} E[Y_i^1 - Y_i^0|\text{Shifted-in}] - N_{\text{Out}} E[Y_i^1 - Y_i^0|\text{Shifted-out}]$$

Hence, the ATE for winners and losers is informative of the cost of the policy



Distance to existing cutoff, minus any gender points

Observed outcomes for targeted applicants

----- Observed outcomes for non-targeted applicants









Results

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	Progress after 5 years
Effect of AA for winners	-0.155
Effect of AA for losers:	-0.106
Applicants shifted per cohort	77

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## **Conclusion and Roadmap**

- Gender points lead to higher share of targeted students
  - Some effect likely runs through application responses
- Little evidence of mismatch
  - More gender balance at little or no cost to the program
  - Yet some cost to the displaced students
- Next: Model application behaviour

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## Example

	Applicant of targeted gender				
Program		Raw	Application Score	Cutoff	Offer
1	Civil and Environmental Engineering	54	56	54.8	Yes
2	Teaching	54	54	48	No

	Applicant of targeted gender				
Program		Raw	Application Score	Cutoff	Offer
1	Civil and Environmental Engineering	54	56	54.8	Yes
2	Teaching	54	54	48	No

## Applicant of non-targeted gender

	Program	Raw	Application Score	Cutoff	Offer
1	Civil and Environmental Engineering	54	54	54.8	No
2	Teaching	54	54	48	Yes



Programs by use of gender points

	Number of programs	Percent of total
Never	1181	91.20
Time-varying	19	1.47
Always	95	7.34
Total	1295	100.00





## First stage: Enrollment



## **STEM** enrollment



Difference in LATEs: 0.045 (t-statistic: 1.25)