

# **Beliefs about Gender and Task Performance: Replicating Bordalo et al. (2019) in an Indian Context**

ECO-3610: Experimental Economics

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# Research Question

## Context about Paper:

Bordalo et al. (2019) show how stereotypes shape beliefs through:

- **Stereotype exaggeration:** self-beliefs exceed actual ability
- **Difficulty-based distortion:** larger overestimation for hard questions

## Research Question:

- How do gender stereotypes distort beliefs about task performance in male-typed vs. female-typed domains in an Indian university context?

# Hypotheses

## **H1: Stereotype Exaggeration**

- Participants' self-beliefs align with gender stereotypes independent of actual ability.

## **H2: Gender Salience Amplification**

- Revealing partner's gender amplifies belief distortions.

## **H3: Contextual Attenuation**

- Smaller distortions at Ashoka (60% female) due to gender balance.

# Experimental Design Flow (Part 1)

1. **Consent:** Participants read study info and agree

2. **Instructions:** Overview of categories, rules, belief elicitation

3. **Comprehension Check:** MCQs on incentives

4. **Ability Task — Cars:** 10 Qs + Self-belief

5. **Ability Task — Beauty:** 10 Qs + Self-belief

# Experimental Design Flow (Part 2)

**6. Partner Assignment (TREATMENT):** Random assignment + pronoun reveal + partner score predictions

**7. Demographics:** Age, gender, major

**8. Payment + Thank You**

# Linking Hypotheses to Survey Design

## H1: Stereotype Exaggeration

Two gender-typed domains: **Cars (male)** and **Beauty (female)**.

Respondents:

- Answer 20 factual questions (actual ability)
- Predict their own score (self-belief)

If self-beliefs favour one's gender even after controlling for actual ability → stereotype distortion.

## H2: Gender Salience Amplification

Partner introduced through a gendered pronoun → triggers gender salience.

Respondents:

- Predict partner's score in each domain
- Rate confidence in predictions

If revealing gender increases stereotype-aligned predictions → amplification effect.

# Outcome Variables, Treatment, Measurement

## Outcome Variables

- ① **Self-belief:** Expected score for self
- ② **Partner Confidence** Belief about partner's ability

**Treatment: Gender Matching** using Partner Gender Reveal

- 1 = indicates female
- 2 = indicates male
- 0 = control

# Why & How We Generated Data Using GPT

## Why synthetic (artificial) data?

- Time + recruitment constraints before exam period
- Needed a larger, balanced sample to test hypotheses
- Pilot full analysis pipeline before running real study
- Stress-test design (power, robustness, coding checks)

## How we generated it

- Used a large language model (GPT) as a data generator
- Long, structured prompt specifying:
  - Experimental context + treatment structure
  - Variable names and distributions from Qualtrics
  - Behavioural patterns from Bordalo et al. (2019)
- Model instructed to output a clean CSV only



# Synthetic Data Generation Prompt: Summary

**Task to GPT** Generate a **synthetic dataset of 400 participants** for our adaptation of *Bordalo et al. (2019)* at Ashoka University, with:

- Internally coherent observations and realistic heterogeneity
- Explicit stereotype-driven patterns in beliefs
- Added noise so that regressions and tests are non-trivial

Final output: **CSV only**, no extra text.

## Sample Design & Treatments

- Total  $N = 400$  Ashoka-style undergraduates
- Treatment: `treatment_pronoun = 0/1`
  - 200 control (no pronoun / gender-anonymous partner)
  - 200 treatment (partner gender made salient)
- Within treatment:  
`partner_gender_code = 0` (control)/`1`(male pronoun)/`2`(female pronoun)
  - 100 male partners, 100 female partners
  - Indicators for `gender_matching + dyad_type`

# What the Prompt Asked GPT to Simulate

## Survey / decision structure encoded

- Two 10-question domains:
  - Cars (male-typed domain)
  - Beauty & Skincare (female-typed domain)
- After each domain:
  - Self-belief about own score in that domain
  - Randomly assigned partner + beliefs about partner's performance

## Actual performance variables (“true ability”)

- `cars_correct`, `beauty_correct`  $\in \{0, \dots, 10\}$
- Distributions with  $SD \approx 1.5$ –2 and 5–10% “weird” types

## Self-beliefs and distortions

- `selfbelief_cars`, `selfbelief_beauty` in  $[0, 10]$
- Added noise ( $SD \approx 1$ –1.5), truncated to  $[0, 10]$

# Partner Beliefs, Derived Variables & Targets

## Partner beliefs (by treatment and gender-salience)

- Latent “true” partner ability drawn from same gendered distributions
- Variables:
  - `partner_cars_belief`, `partner_beauty_belief`  $\in [0, 10]$
  - `partner_score_belief`  $\in [0, 20]$  (aggregate score out of 20)
  - `partner_confidence`  $\in [0, 100]$  (granular measure)
- Noise added (SD  $\approx 2-3$ ), truncated to feasible ranges

## Derived variables and behavioural indicators

- Distortions:
  - `distort_self_cars` = `selfbelief_cars` - `cars_correct`
  - `distort_self_beauty` = `selfbelief_beauty` - `beauty_correct`
- Overconfidence dummies:
  - `overconfident_cars` = `1(distort_self_cars > 0)`
  - `overconfident_beauty` = `1(distort_self_beauty > 0)`
- Additional variables:
  - `pref_domain_self` (cars vs beauty vs equal)
  - `time_dummy_fast` (top 25% fastest responses)

# Sample Profile

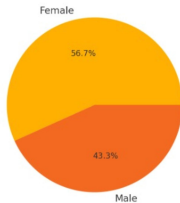


Figure A: Gender Split

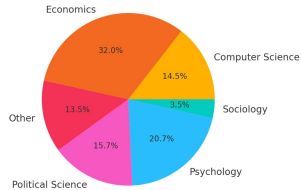


Figure B: Major Composition

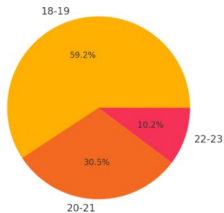


Figure C: Age Split

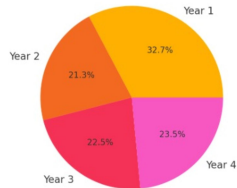


Figure D: Year of Study

# Power Analysis: Parameters and Rationale

- The original paper reports stereotype effects in belief formation, using bank-level regressions.
- We base our power calculation on the stereotype coefficient ( $\theta\sigma$ ) reported in:
  - **Table 4, Column 3:** beliefs about men  $\rightarrow \theta\sigma = 0.45$
  - **Table 4, Column 4:** beliefs about women  $\rightarrow \theta\sigma = 0.14$
- These are belief outcomes measured on a 0–1 scale, so we assume a conservative standard deviation of 0.23 which we estimated using simulated data.
- The power test evaluates whether an effect of this magnitude is detectable using a two-sample mean comparison (t-test framework), which maps cleanly onto the treatment versus control contrast in our design.

# Power Result and Sample Choice

- Using the two published stereotype coefficients:
  - Large effect:  $\Delta = 0.45$
  - Moderate effect:  $\Delta = 0.14$
- Stata estimates the required sample sizes for 80% power (two-sided test,  $\alpha = 0.05$ ) to detect each effect.
- Both required sample sizes fall well below 400 observations, meaning that any effect within the range observed in Bordalo et al. (2019) would be identifiable with substantially fewer subjects.
- Since we simulated **400 participants**, our experiment is **well powered** to detect stereotype-induced distortions in belief reporting, including smaller effects than those documented in the original study.
- Final note: our use of  $N = 400$  reflects feasibility and design planning, rather than minimum detectable power.

# Descriptive Analysis: Summary Statistics

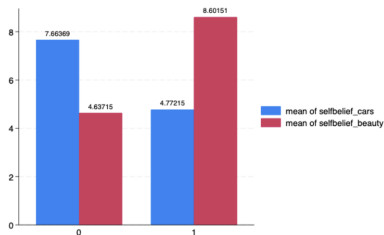


Figure 1: Self-Belief Index

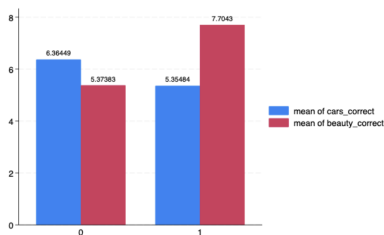


Figure 2: Score

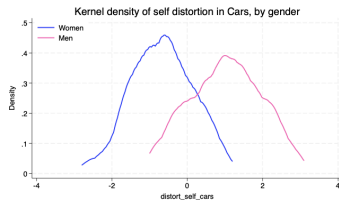
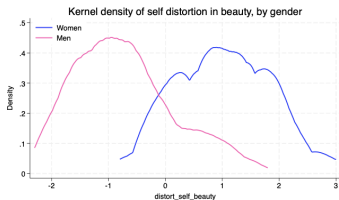
**Gender coding:** 0 = male, 1 = female.

Self-beliefs differ more by gender than actual performance: men are more confident in cars, while women show higher belief in beauty.

Score differences are small, yet belief differences are large, indicating stereotype-driven confidence rather than ability.

Overall, the data suggests that confidence is gender-typed and amplified relative to real performance outcomes.

# Descriptive Analysis: Graphs



## Self-distortion in Beauty

In Beauty, women's distortion distribution lies to the right of men's, indicating higher overconfidence relative to actual performance.

## Self-distortion in Cars

In Cars, the pattern reverses: men display stronger positive distortion, while women are more calibrated or underconfident.

This highlights confidence asymmetry: small score differences generate large, systematic belief gaps by gender.



# Descriptive Analysis: Graphs

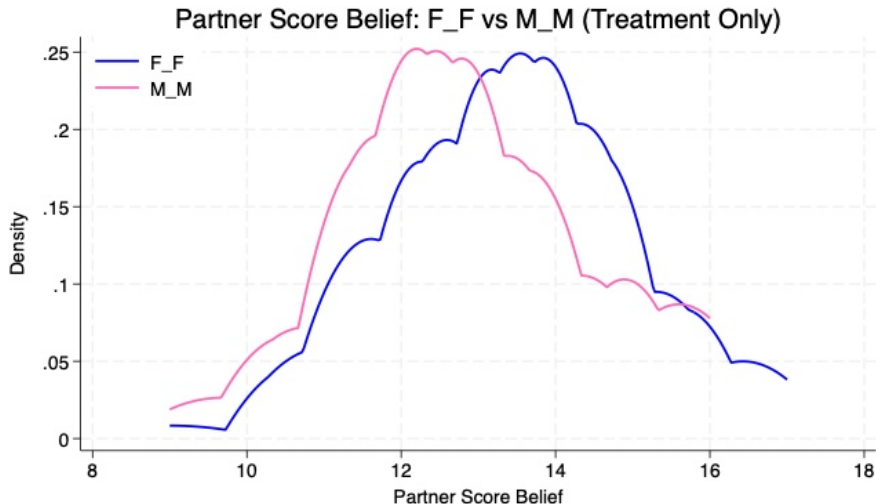


Figure 3: Graph

# Descriptive Analysis: Graphs

- We restrict attention to the **treatment group** and study gender-matched vs. gender-mismatched dyads.
- For the treatment sample, we construct a categorical dyad variable capturing partner pairings, where

$$\text{dyad\_type} \in \{M\_F, F\_M, F\_F, M\_M\},$$

with, for example, *M\_F* denoting a *male subject matched with a female partner*.

- Kernel densities for all dyad combinations are reported in the log file; here we focus on the most informative case of gender-matched dyads.
- Male–male (M\_M) dyads exhibit slightly lower partner-belief densities, suggesting moderate beliefs and competitive ingroup deflation.
- Female–female (F\_F) dyads display more confidence projection consistent with altruistic evaluation.
- Women express greater performance optimism, while men remain less inflationary.

# Statistical Test for Differences across genders

Significant gaps between men and women in both domains → confidence patterns align with stereotypes rather than performance. Distortion is measured by the difference in belief about performance and actual score received.

```
. ttest distort_self_cars, by(gender)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
Female	227	-.6674809	.0545106	.8212842	-.7748148	-.559987
Male	173	1.016185	.0716771	.9427652	.8747049	1.157665
Combined	400	.06075	.0604698	1.209397	-.0581293	.1796293
diff		-1.683586	.0883945		-1.857364	-1.509807

diff = mean(Female) - mean(Male) t = -19.0463  
H0: diff = 0 Degrees of freedom = 398

Ha: diff < 0      Ha: diff != 0      Ha: diff > 0  
Pr(T < t) = 0.0000      Pr(|T| > |t|) = 0.0000      Pr(T > t) = 1.0000

Figure (a): Distortion of beliefs (cars)

```
. ttest distort_self_beauty, by(gender)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
Female	227	1.020795	.0560191	.8440125	.9103184	1.131091
Male	173	-.7236994	.0667885	.878466	-.8555301	-.5918687
Combined	400	.26625	.0609282	1.218564	.1464696	.3860304
diff		1.744404	.0867008		1.573955	1.914853

diff = mean(Female) - mean(Male) t = 20.1198  
H0: diff = 0 Degrees of freedom = 398

Ha: diff < 0      Ha: diff != 0      Ha: diff > 0  
Pr(T < t) = 1.0000      Pr(|T| > |t|) = 0.0000      Pr(T > t) = 0.0000

Figure (b): Distortion of beliefs (beauty)

# Statistical Test for Differences Within Treatments

When looking at male respondents within the treatment, gender revelation leads to partner beliefs shifting toward stereotype-consistent expectations (MF = 1 and MM = 0)

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	56	7.839286	.1502742	1.124549	7.53813	8.140442
1	32	4.71875	.1864047	1.054464	4.338575	5.098925
Combined	88	6.704545	.1987152	1.864114	6.309577	7.099514
diff		3.120536	.2437173		2.636042	3.60503
diff = mean(0) - mean(1)				t = 12.8039		
H0: diff = 0				Degrees of freedom = 86		
Ha: diff < 0		Ha: diff != 0		Ha: diff > 0		
Pr(T < t) = 1.0000		Pr( T  >  t ) = 0.0000		Pr(T > t) = 0.0000		

Figure 1: Second-order beliefs about car score

. ttest partner\_beauty\_belief, by(dyad\_MF\_MM)

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	56	5.089286	.159701	1.195093	4.769238	5.409334
1	32	9.0625	.2196841	1.242721	8.614451	9.510549
Combined	88	6.534091	.2418736	2.268975	6.053341	7.014841
diff		-3.973214	.2686864		-4.507345	-3.439083
diff = mean(0) - mean(1)				t = -14.7876		
H0: diff = 0				Degrees of freedom = 86		
Ha: diff < 0		Ha: diff != 0		Ha: diff > 0		
Pr(T < t) = 0.0000		Pr( T  >  t ) = 0.0000		Pr(T > t) = 1.0000		

Figure 2: Second-order beliefs about beauty score

# Statistical Test for Differences Within Treatments

Female participants report higher partner confidence when paired with men in the male-typed domain, and lower partner confidence when paired with women, even though this belief need not reflect true ability. (FF = 0 and FM = 1)

```
. ttest partner_cars_belief, by(dyad_FM_FF)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	70	4.871429	.1301718	1.089095	4.611743	5.131114
1	61	7	.1258848	.9831921	6.748193	7.251807
Combined	131	5.862595	.1299367	1.487193	5.605531	6.11966
diff		-2.128571	.1823673		-2.48939	-1.767753

diff = mean(0) - mean(1) t = -11.6719  
H0: diff = 0 Degrees of freedom = 129

Ha: diff < 0 Ha: diff != 0 Ha: diff > 0  
Pr(T < t) = 0.0000 Pr(|T| > |t|) = 0.0000 Pr(T > t) = 1.0000

Figure 3: Second-order beliefs about car score

```
. ttest partner_beauty_belief, by(dyad_FM_FF)
```

Two-sample t test with equal variances

Group	Obs	Mean	Std. err.	Std. dev.	[95% conf. interval]	
0	70	8.585714	.1402347	1.173288	8.305954	8.865475
1	61	5.442623	.1594468	1.245319	5.123682	5.761564
Combined	131	7.122137	.1730602	1.980765	6.779758	7.464516
diff		3.143091	.2114687		2.724695	3.561487

diff = mean(0) - mean(1) t = 14.8632  
H0: diff = 0 Degrees of freedom = 129

Ha: diff < 0 Ha: diff != 0 Ha: diff > 0  
Pr(T < t) = 1.0000 Pr(|T| > |t|) = 0.0000 Pr(T > t) = 0.0000

Figure 4: Second-order beliefs about car score

# Statistical Tests of Belief Distortion: First- and Second-Order Effects

## First-Order Distortions (Self-Belief vs Actual Score)

- Men significantly overestimate performance in Cars, women do so in Beauty.
- Distortion gaps persist even when performance does not differ equivalently.
- Women remain relatively cautious even when skilled, while men project confidence across domains.
- This reflects stereotype-consistent belief formation rather than pure ability assessment.

# Statistical Tests of Belief Distortion: First- and Second-Order Effects

## Second-Order Distortions (Partner Belief)

- Under pronoun treatment, partner-confidence shifts toward stereotyped expectations.
- In Cars: Female participants report higher confidence when matched with men (FM) than women (FF).
- In Beauty: Female–female dyads (FF) show higher partner-belief than mixed dyads.
- Suggests identity cues trigger representativeness amplification — small score differences inflate belief gaps.
- May also reflect gender differences in competitiveness vs cooperative evaluation.

**Conclusion:** Across self- and partner-beliefs, confidence is not an unbiased reflection of performance: it is *gender-filtered*, *stereotype-aligned*, and *treatment-responsive*.

# Regression Equation

## Baseline Interaction

$$\text{BeliefOnPartnerConfidence}_i = \beta_0 + \beta_1 \text{Female}_i + \beta_2 \text{GenderMatching}_i \\ + \beta_3 (\text{Female}_i \times \text{GenderMatching}_i) + \epsilon_i$$

## Interaction with Controls

$$\text{BeliefOnPartnerConfidence}_i = \beta_0 + \beta_1 \text{Female}_i + \beta_2 \text{GenderMatching}_i \\ + \beta_3 (\text{Female}_i \times \text{GenderMatching}_i) + \beta_4 \text{Age}_i + \\ \beta_5 \text{Major}_i + \beta_6 \text{YearOfStudy}_i + \epsilon_i$$

- We construct a categorical variable for gender-matching, defined as:

$$\text{gender\_matching} \in \begin{cases} 0 & = \text{Control (no gender revealed)} \\ 1 & = \text{Gender-matched pairs (F\_F, M\_M)} \\ 2 & = \text{Gender-mismatched pairs (M\_F, F\_M)} \end{cases}$$

- Interpretation of **coefficient of interest** ( $\beta_3$ ): The female-male difference in matching relative to the female-male difference in control.



# OLS

```
. reg partner_confidence i.female##i.gender_matching
```

Source	SS	df	MS	Number of obs	=	400
Model	12242.5181	5	2448.50362	F(5, 394)	=	24.27
Residual	39745.2794	394	100.876344	Prob > F	=	0.0000
				R-squared	=	0.2355
				Adj R-squared	=	0.2258
Total	51987.7975	399	130.295232	Root MSE	=	10.044

partner_confidence	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.female	- .7140931	1.495853	-0.48	0.633	-3.654945	2.226758
gender_matching						
1	13.62668	1.728625	7.88	0.000	10.2282	17.02516
2	6.077574	2.083067	2.92	0.004	1.982258	10.17289
female#gender_matching						
1 1	-3.525193	2.340946	-1.51	0.133	-8.127501	1.077115
1 2	4.57731	2.653992	1.72	0.085	-.6404459	9.795067
_cons	60.14118	1.089395	55.21	0.000	57.99942	62.28293

Figure: OLS Regression Output 1

# OLS Regression: Partner Confidence (Baseline)

- We regress second-order beliefs (partner confidence) on gender of subject, gender-matching, and their interaction.
- Female coefficient is near zero and insignificant  $\Rightarrow$  no baseline gender gap in partner expectations.
- GenderMatching is large and significant ( $\beta_2 \approx 13.6$ ,  $p < 0.001$ )  $\Rightarrow$  matched partners are rated higher.
- Female  $\times$  GenderMatching positive but weak  $\Rightarrow$  women do not inflate partner-belief as strongly as men.

**Takeaway:** Matching drives confidence, and not gender alone. In-group expectations boost perceived partner ability.

# OLS

```
. reg partner_confidence i.female##i.gender_matching i.agenumeric i.major_num i.year_of_study
```

Source	SS	df	MS	Number of obs	=	400
Model	13230.1447	15	882.009645	F(15, 384)	=	8.74
Residual	38757.6528	384	100.931388	Prob > F	=	0.0000
				R-squared	=	0.2545
				Adj R-squared	=	0.2254
Total	51987.7975	399	130.295232	Root MSE	=	10.046

partner_confidence	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.female	-1.453279	1.529537	-0.95	0.343	-4.460595	1.554037
gender_matching						
1	13.99977	1.786404	7.84	0.000	10.48741	17.51212
2	6.112524	2.13285	2.87	0.004	1.918998	10.30605
female#gender_matching						
1 1	-3.408218	2.38048	-1.43	0.153	-8.088625	1.272188
1 2	4.913878	2.680378	1.83	0.068	-.3561766	10.18393

Figure: OLS Regression Output 2

# OLS w/ Controls: Age, Major, Year

- Adding controls barely changes estimates. Gender-matching remains highly predictive.
- $\beta_2 \approx 14$  with  $p < 0.001 \Rightarrow$  partner-belief increases sharply when genders align.
- Interaction term now positive and marginally significant: female-matching amplifies belief slightly in pooled tasks.
- Controls increase model fit ( $R^2 \approx 0.25$ ) but do not change core inference.

**Inference:** Partner-belief differences are driven by identity cues, not demographic structure or ability signals.

```
. reg partner_beauty_belief i.female##i.gender_matching i.agenumeric i.major_num i.year_of_study
```

Source	SS	df	MS	Number of obs	=	400
Model	<b>679.901136</b>	<b>15</b>	<b>45.3267424</b>	F(15, 384)	=	<b>33.47</b>
Residual	<b>520.008864</b>	<b>384</b>	<b>1.35418975</b>	Prob > F	=	<b>0.0000</b>
				R-squared	=	<b>0.5666</b>
				Adj R-squared	=	<b>0.5497</b>
Total	<b>1199.91</b>	<b>399</b>	<b>3.00729323</b>	Root MSE	=	<b>1.1637</b>

partner_beauty_belief	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.female	<b>-.0492488</b>	<b>.1771685</b>	<b>-0.28</b>	<b>0.781</b>	<b>-.3975907</b>	<b>.299093</b>
gender_matching						
1	<b>-2.147954</b>	<b>.2069219</b>	<b>-10.38</b>	<b>0.000</b>	<b>-2.554796</b>	<b>-1.741112</b>
2	<b>1.857201</b>	<b>.2470511</b>	<b>7.52</b>	<b>0.000</b>	<b>1.371458</b>	<b>2.342943</b>
female#gender_matching						
1 1	<b>3.57362</b>	<b>.2757345</b>	<b>12.96</b>	<b>0.000</b>	<b>3.031481</b>	<b>4.115758</b>
1 2	<b>-3.545534</b>	<b>.3104722</b>	<b>-11.42</b>	<b>0.000</b>	<b>-4.155973</b>	<b>-2.935096</b>

Figure: OLS Regression Output 3

# OLS: Partner Belief in Beauty Domain

- Strong effects appear when looking only at beauty (a female-typed task.)
- Female dummy remains insignificant, but GenderMatching is large and highly significant ( $p < 0.001$ ).
- Female  $\times$  GenderMatching is strongly positive ( $\approx 3.58$ ,  $p < 0.001$ )  $\Rightarrow$  females trust female partners more in Beauty.
- In mixed dyads, confidence is lower, confirming stereotype congruence mechanics.

**Conclusion:** When stereotypes align (Beauty–Female), matching boosts belief formation dramatically, implying identity amplifies expectation.

```
. reg partner_cars_belief i.female##i.gender_matching i.agenumeric i.major_num i.year_of_study
```

Source	SS	df	MS	Number of obs	=	400
Model	425.088314	15	28.339221	F(15, 384)	=	22.52
Residual	483.289186	384	1.25856559	Prob > F	=	0.0000
				R-squared	=	0.4680
				Adj R-squared	=	0.4472
Total	908.3775	399	2.27663534	Root MSE	=	1.1219

partner_cars_belief	Coefficient	Std. err.	t	P> t	[95% conf. interval]	
1.female	- .6781079	.1707988	-3.97	0.000	-1.013926	-.34229
gender_matching						
1	1.221488	.1994824	6.12	0.000	.8292735	1.613703
2	-1.905547	.2381689	-8.00	0.000	-2.373825	-1.437268
female#gender_matching						
1 1	-2.375269	.265821	-8.94	0.000	-2.897915	-1.852622
1 2	2.931442	.2993098	9.79	0.000	2.34295	3.519933

Figure: OLS Regression Output 4

# OLS Regression: Partner Beliefs in Cars

- Outcome: second-order beliefs about partner ability in the car domain.
- Female dummy is negative and highly significant ( $\beta_1 \approx -0.67, p < 0.001$ ), indicating lower partner-confidence among women overall.
- Gender matching increases confidence ( $\beta_2 \approx +1.22, p < 0.001$ ) i.e. same-gender pairs are rated higher.
- Interaction splits sharply by stereotype:
  - FF pairs  $\Rightarrow$  strong down-weighting of partner ability ( $-2.38, p < 0.001$ )
  - FM pairs  $\Rightarrow$  large confidence uplift for male partners ( $+2.93, p < 0.001$ )

**Conclusion:** In Cars, confidence is allocated along gender-stereotype lines: female–female teams are penalized, male partners are rewarded.



# Caveats & Interpretation

## Limitations

- Reflects model priors, not real cognition
- Lower behavioural noise
- Stereotype effects may appear cleaner
- limited to 2 categories
- Did not incorporate DIM
- Ashoka not representative of broader Indian University

## How to use the data

- Pilot for coding and design
- Not a substitute for real data

# Conclusion and Policy Takeaways

- Our major contribution: proving the gender salience hypothesis. In Bordalo's study, the results on this were insignificant.
- Broadly, the expectations that we had for beliefs based on gender and gender type of questions hold.
- This has important implications for policy:
  - Designing environments that are conducive to gender-neutral contributions so as to control for "perceptions" or "stereotype" led discourse and outcomes.
  - Designing curriculum that starts accounting for stereotype-led belief formation, particularly with regards to certain topics that may be associated with genders and lead to the emphasising of gender norms.
  - Both can be done by framing narratives differently.