

## Working women and caste in India

Working outside the house is a social-stigma for women in India.

Only the poorest women are working in blue-collar jobs.

Once their family income increases, they withdraw from the workforce.

Women re-enter the work force at high education levels in white-collar jobs.

Scheduled castes and Scheduled tribes (Sc/St) = socio-economically disadvantaged.

⇒ Caste is an important determinant of a woman's work-status in India.

Our Questions:

1. Is the ability to infer a woman's work-status based on her caste changing over generations?

2. What is the change in the effect of caste on work-status over generations?

## Data and Empirical Methodology

We use a nationally representative dataset from the **National Family Health Survey conducted in 2015-16 (NFHS-4)** consisting of **81,816** women aged 21-49 years across all 29 states and 7 union territories of India.

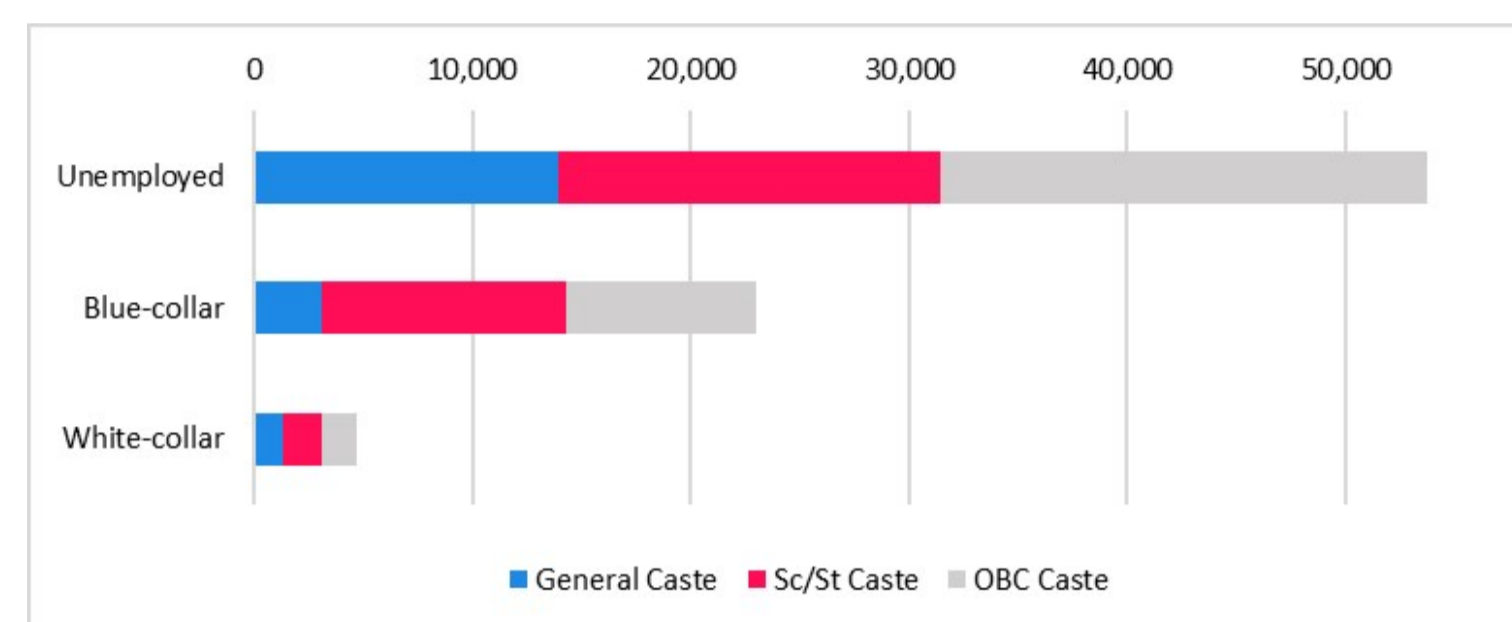


Figure 1: Distribution of number of women from each caste across work-status types

We design three binary classification experiments to predict a woman's work. We train an ensemble **Gradient Boosting Decision Tree (GBDT)** model for each experiment using LightGBM.

Table 1: Training and test set performance for GBDT models. Best parameters were found using 5-fold cross-validated grid search.

Experiment/Description	Class Balance	Training		Test	
		Acc.	F1	Acc.	F1
Having a job or not ( <i>work-status</i> )	0.34	0.69	0.60	0.67	0.57
Having a blue collar job or not ( <i>blue-collar</i> )	0.28	0.71	0.59	0.68	0.55
Having a white collar job or not ( <i>white-collar</i> )	0.05	0.89	0.51	0.85	0.30

## SHAP feature attribution framework

Our models uncover non-linear temporal patterns between caste and women's work-status using the SHAP (SHapley Additive exPlanation) feature attribution framework.

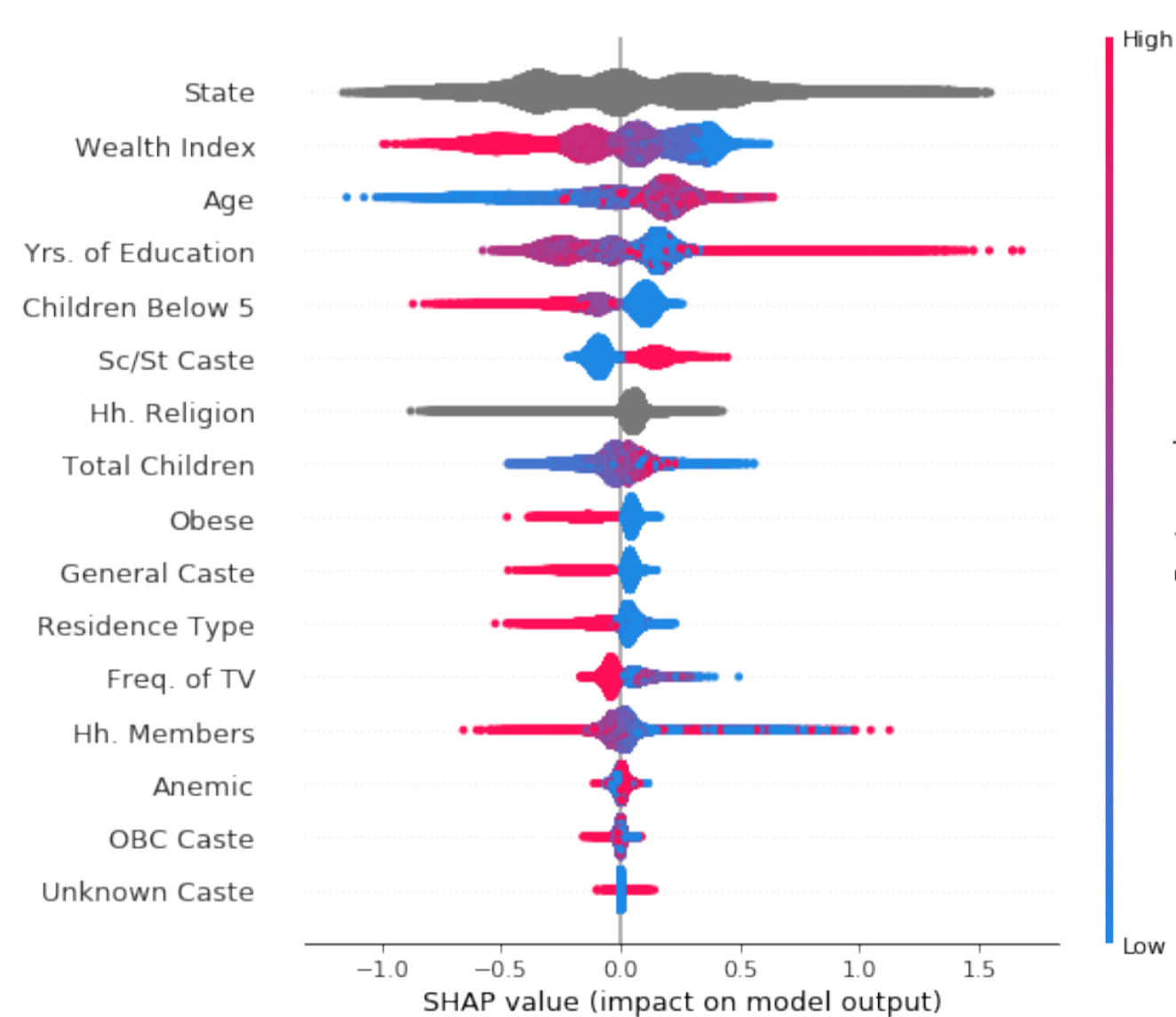


Figure 2: SHAP summary plot for *work-status* experiment. Categorical variable values are grey.

The SHAP summary plot for *work-status* experiment in Figure 2 shows the relative importance of features, the distribution of impacts of features on the model's prediction, as well as how the feature's value (Low to High) relates to its impact.

## Impact of caste on work-status across states of India

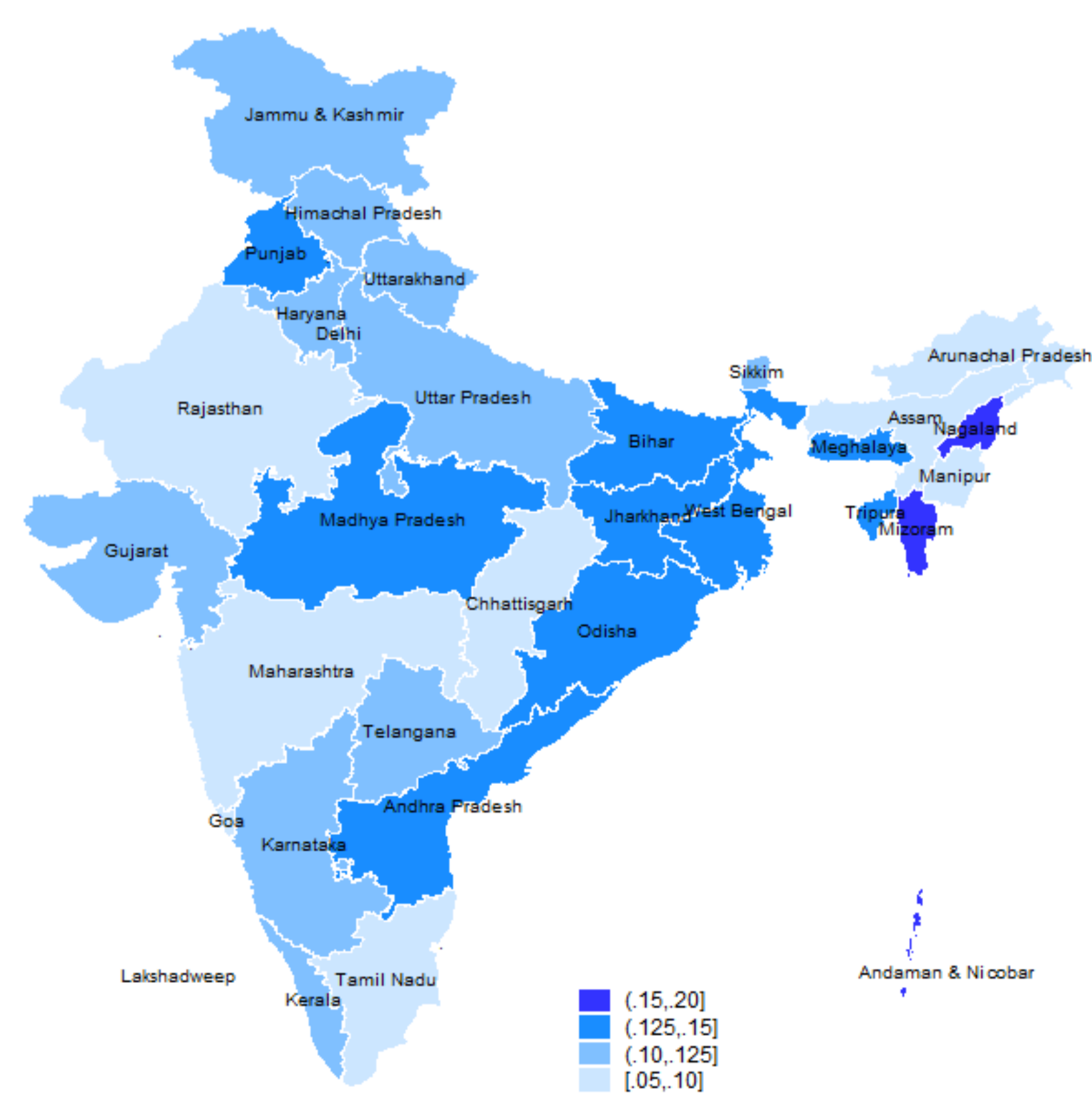


Figure 3: Mean of the magnitude of SHAP values of Sc/St Caste across states of India for *work-status* experiment

## Impact of caste on work type over generations

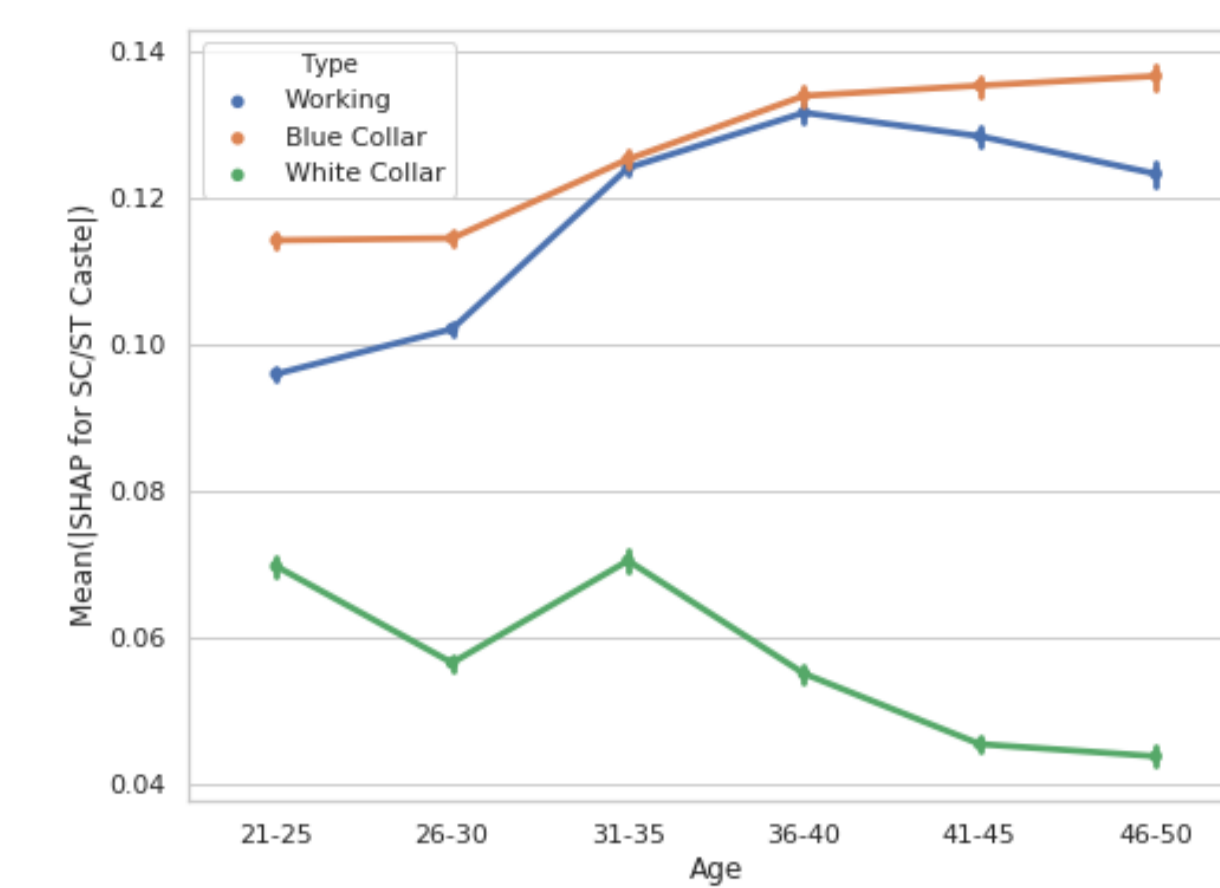


Figure 4: Mean of the magnitude of SHAP values of Sc/St Caste across 5-year age-bins.

Caste is more important in predicting work-status of older women than of younger women.

Same pattern is observed for blue-collar jobs. For white collar jobs, there isn't a clear monotonic pattern, but overall, caste is more important for women younger than 35 years of age.

## Impact of caste on work-status using SHAP interaction effects

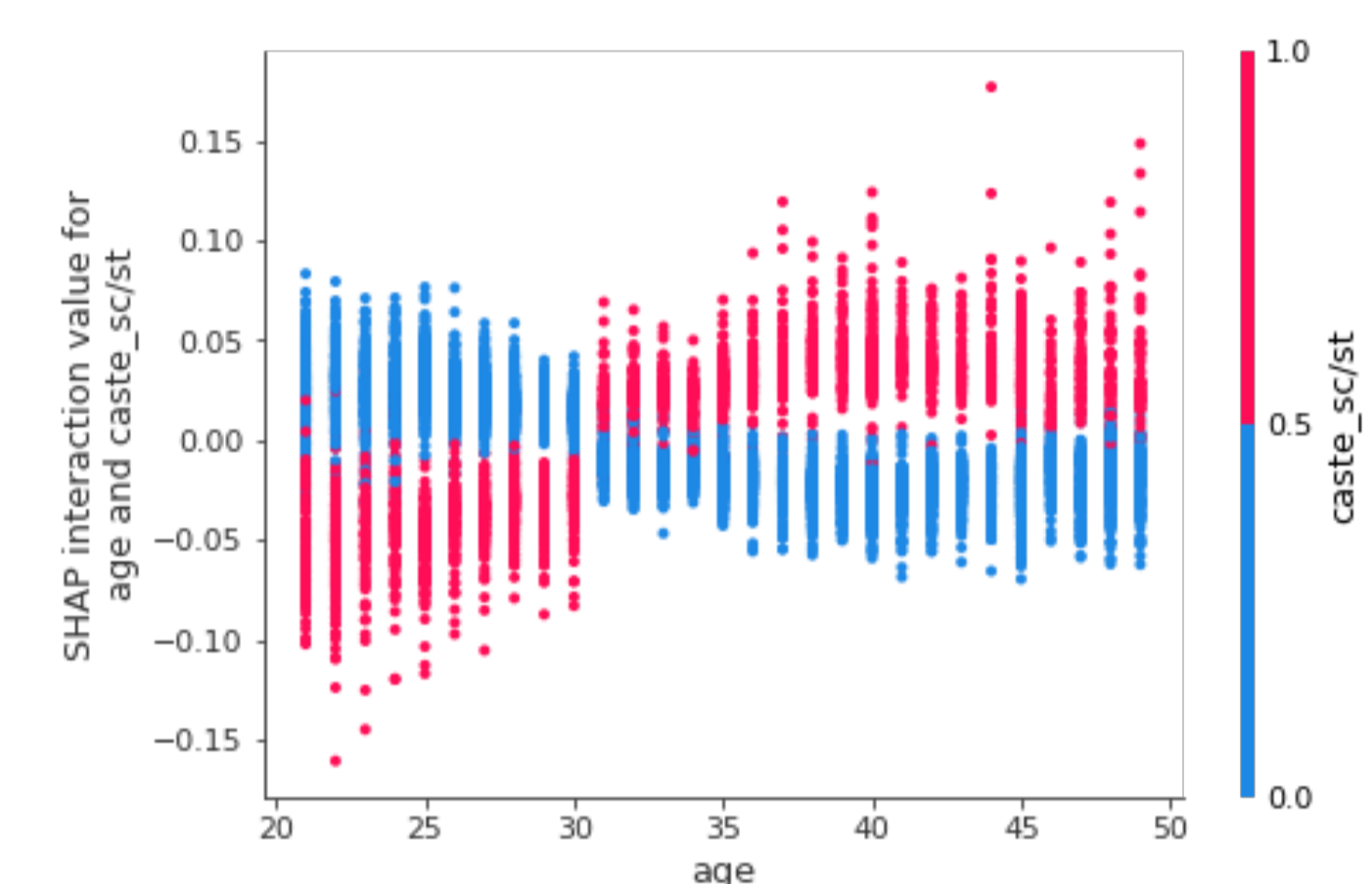


Figure 5: SHAP interaction plot of age and Sc/St Caste for *work-status* experiment.

For younger Sc/St women, SHAP interaction values are larger negatives, implying a high prediction for not-working.

For older Sc/St women, SHAP interaction values are larger positives, implying a high prediction for working.

## Impact of caste on blue collar and white collar jobs

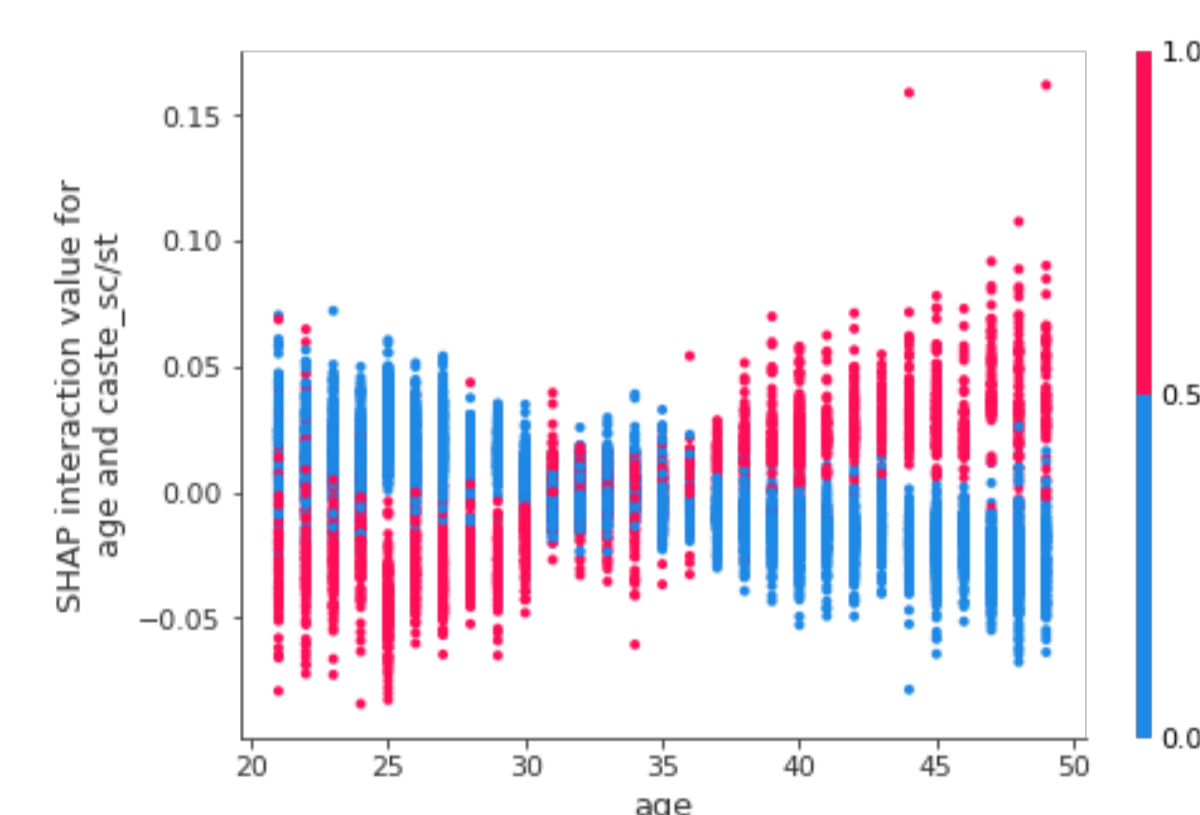


Figure 6: SHAP interaction plot of age and Sc/St Caste for *blue-collar* experiment.

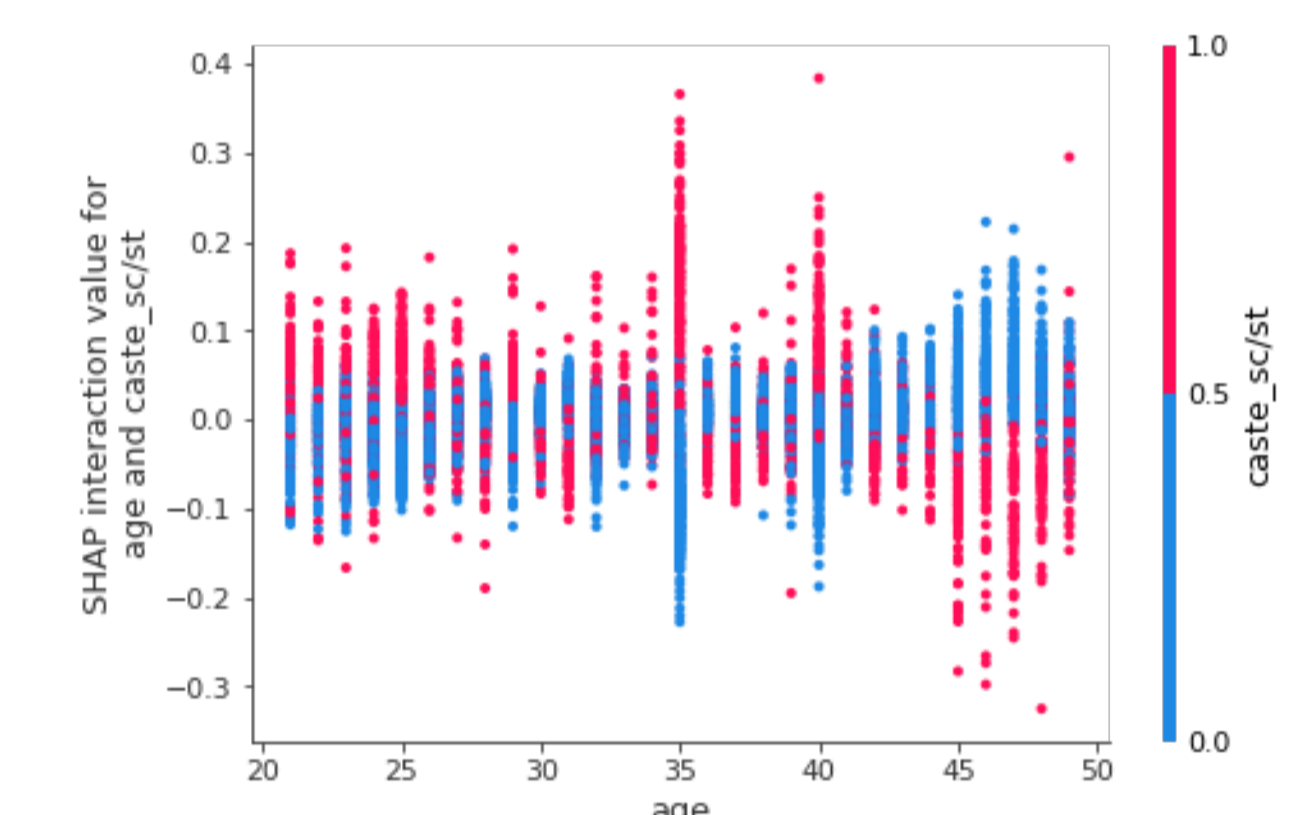


Figure 7: SHAP interaction plot of age and Sc/St Caste for *white-collar* experiment.

The pattern observed in Figure 5 is only relevant for blue-collar jobs (Figure 6).

Opposite pattern is observed for white-collar jobs (Figure 7). Younger Sc/St women are more likely to be working in white-collar jobs while older Sc/St women are less likely to be working in white-collar jobs.

## Conclusions and implications

### Conclusions

- Caste has become a less important determinant of younger women's work-status, specially their participation in blue-collar jobs.
- Younger women of Sc/St caste are less likely to be working in blue collar-jobs, and more likely to be working in white-collar jobs.

### Implications

1. **Monitoring:** A cost-effective tool to monitor the impact of the existing caste-based quotas set by the Government of India in public education and jobs.
2. **Targeting:** Future work will look deeper into geographical variations to identify regions where younger women are lagging, and job creation can be targeted.
3. **Discovering:** Can be used to study the nuanced patterns underlying other social disadvantage and bias in both developing and developed countries.

### Code and Data

All code, datasets and results available at [github.com/chaitjo/working-women](https://github.com/chaitjo/working-women)

## References

- [1] International Institute for Population Sciences. National family health survey (nfhs-4), 2015-16: India. Technical report, Mumbai: International Institute for Population Sciences and ICF, 2017.
- [2] Guolin Ke, Qi Meng, Thomas Finley, Taifeng Wang, Wei Chen, Weidong Ma, Qiwei Ye, and Tie-Yan Liu. Lightgbm: A highly efficient gradient boosting decision tree. In *Advances in Neural Information Processing Systems*, pages 3146--3154, 2017.
- [3] Scott M Lundberg and Su-In Lee. A unified approach to interpreting model predictions. In *Advances in Neural Information Processing Systems*, pages 4765--4774, 2017.