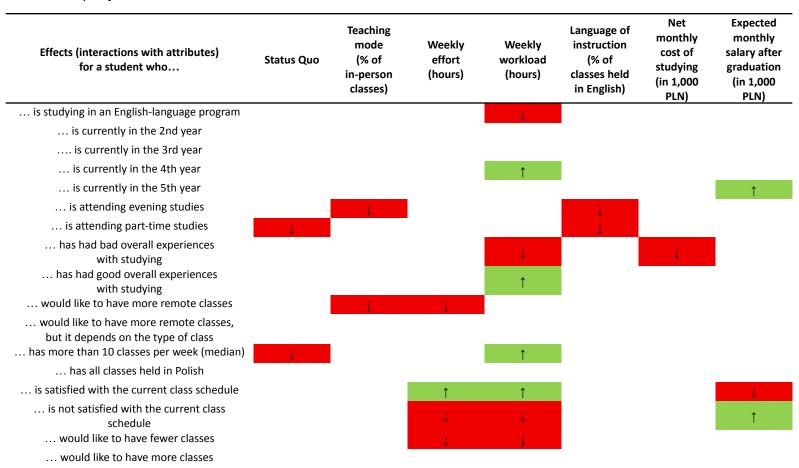
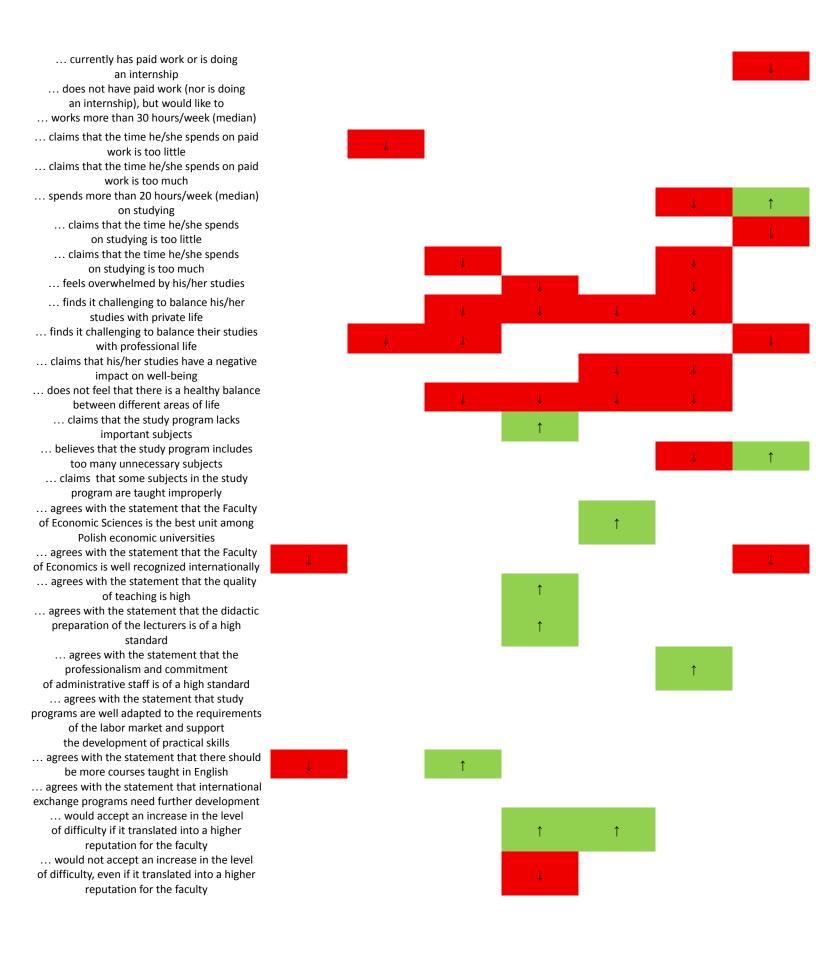
Appendix 4. Preference heterogeneity – MNL models with socio-demographic interactions

To further explain preference heterogeneity, we estimated 61 multinomial logit models with interaction terms, examining differences in preferences of various respondents. Notably, estimating a single model with multiple interactions was not feasible due to the relatively small sample size. Therefore, we estimated separate models, each including only one interaction. While this approach does not capture cross-interaction effects, it represents the most practical solution for illustrating the overall influence of all considered interactions.

For clarity, we chose not to report the coefficient estimates and p-values in Table A5.1. Instead, we visually highlight statistically significant interactions (p-value < 0.10). Red cells with downward arrows indicate a negative effect, while green cells with upward arrows indicate a positive impact. This visual representation allows for an intuitive understanding of how each interaction affects the corresponding attribute. For example, respondents enrolled in an English-language program (first row) tend to prefer programs with a smaller weekly workload.

Table A5.1. Interaction effects in the MNL Models in preference-space with linear attribute specification





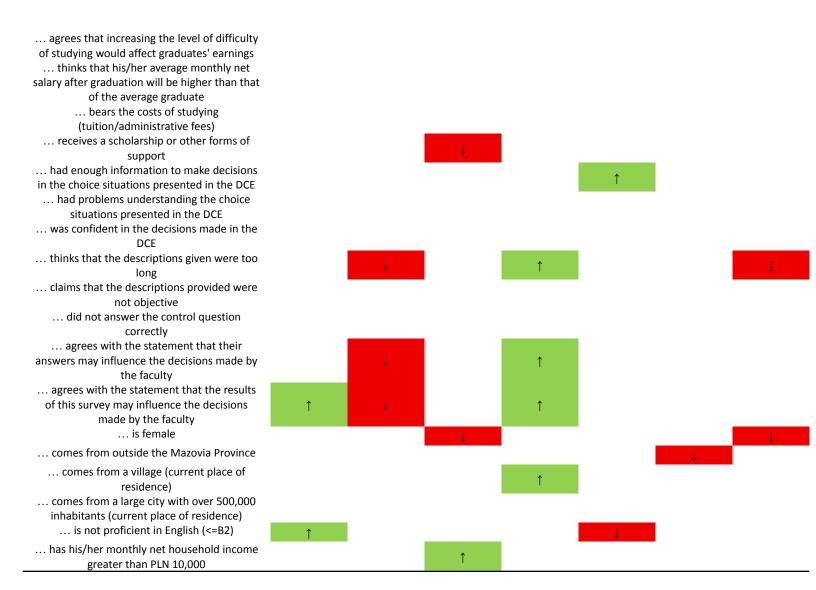


Table A5.1 provides valuable insight, as there are 427 interactions in total (61 models and 7 interactions – one for each attribute), of which 75 are statistically significant. A notable insight is that fifth-year students place significantly greater emphasis on their expected salary after graduation compared to those in earlier years of study. This suggests that students approaching the end of their education are more confident in realizing their earnings potential, whereas those in earlier stages may perceive such expectations as more distant or uncertain.

Students with generally weaker satisfaction from studying tend to prefer dedicating more time to class preparation and express stronger preferences for lower study costs. In contrast, students with greater satisfaction from studying display positive preferences for a higher time commitment to class preparation, possibly reflecting a greater intrinsic enjoyment of learning and a desire for academic engagement. As anticipated, respondents who prefer a higher share of remote learning exhibit a negative preference for an increase in in-person classes. Furthermore, satisfaction with the current study plan appears to systematically influence

preferences. Students satisfied with their existing plan show positive interactions with weekly effort and workload, but have negative interactions with increases in expected salary after graduation. Conversely, those dissatisfied with their plan display the opposite pattern. Similarly, individuals who would prefer fewer classes demonstrate negative preferences for higher weekly effort and workload, aligning with theoretical expectations. Finally, respondents who are currently employed or completing an internship exhibit lower preferences for higher expected post-graduation salaries. This is plausible, as direct engagement with the labor market may provide them with a more grounded understanding of salary prospects and job-related trade-offs.