Evaluation of Seguro Popular

Design Framework

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Thanks, in Mexico, to Edith Arano, Carlos Avila, Juan Eugenio Hernández Avila, Mauricio Hernández Avila, Jorge Carreon Manuel Castro, Octavio Gómez Dantes, Sara Odette León, Héctor Hernández Llamas, René Santos Luna, Tania Martínez, Gustavo Olaíz, Maritza Ordaz, Eduardo González Pier, Héctor Peña, Raymundo Pérez, Esteban Puentes, Corina Santangelo, Sergio Sesma, Sara Uriega, Martha María (Mara) Tellez; and, at Harvard, to Emmanuela Gakidou, Jason Lakin, Ryan T. Moore, Nirmala Ravishankar, Manett Vargas

A New Member of the Evaluation Team

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Evaluation Components

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- Impact Evaluation

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- National Level Analysis

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- Process Evaluation

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- In-depth Focus Groups

Goals of SP & Evaluation Outcome Measures

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  - Places with better doctors and health administration: bigger effects

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  - Infeasible because of political preferences of local officials to affiliate some areas first

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- If politicians (in a democracy) decide which areas get MAOs
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  - Even if SP has no effect, areas with SP will be healthier

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- Generalization: its ethical to randomize at one level below that which officials care

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- Drop clusters from states that did not participate

Remaining in study: 148 clusters in 7 states

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- Any Mexican family may still affiliate at any time
- No randomization at individual level
- Without an evaluation, choices would still be made, but would be arbitrary choices made by local government officials

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- We need estimators robust not merely to statistical assumptions but to real world problems

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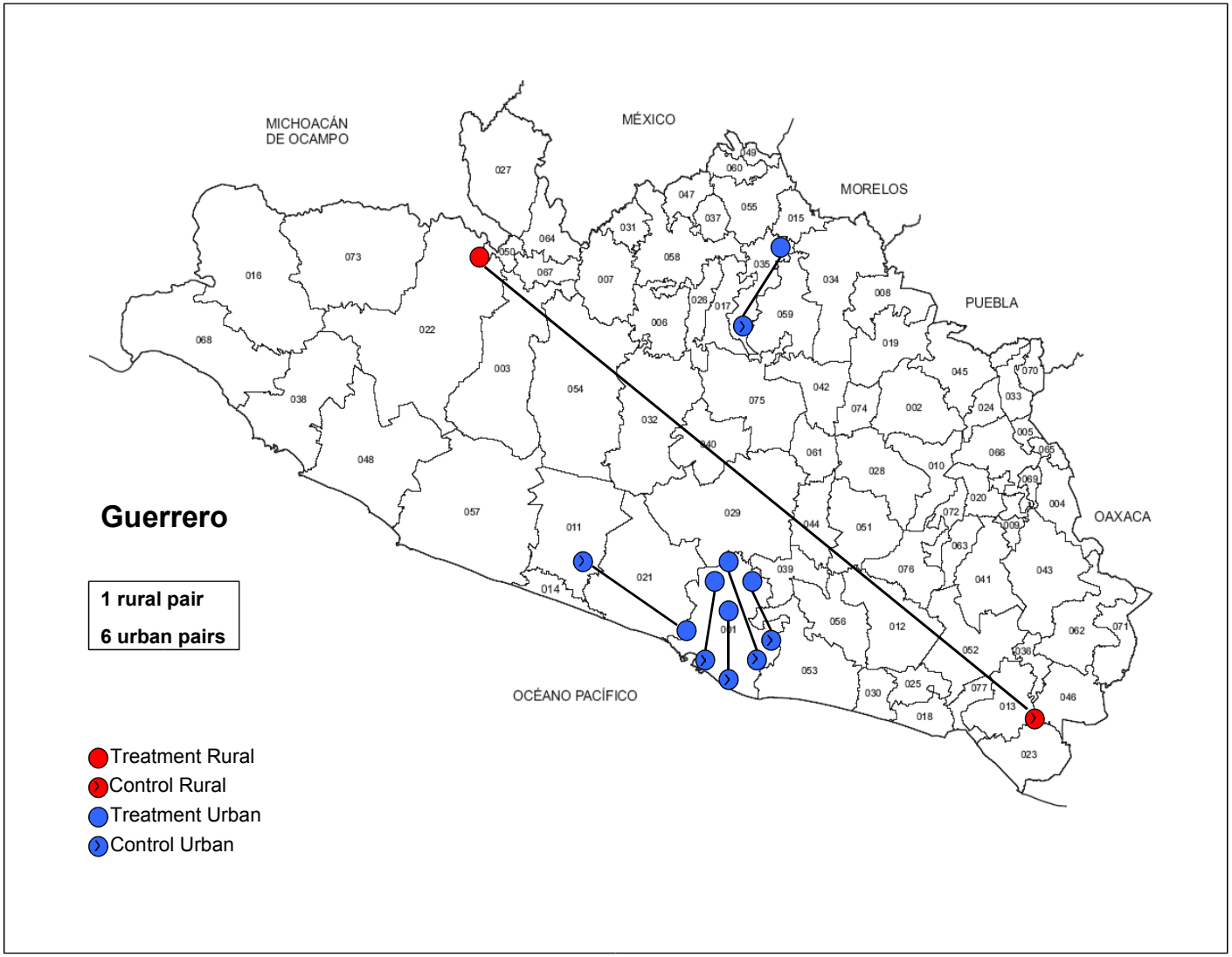
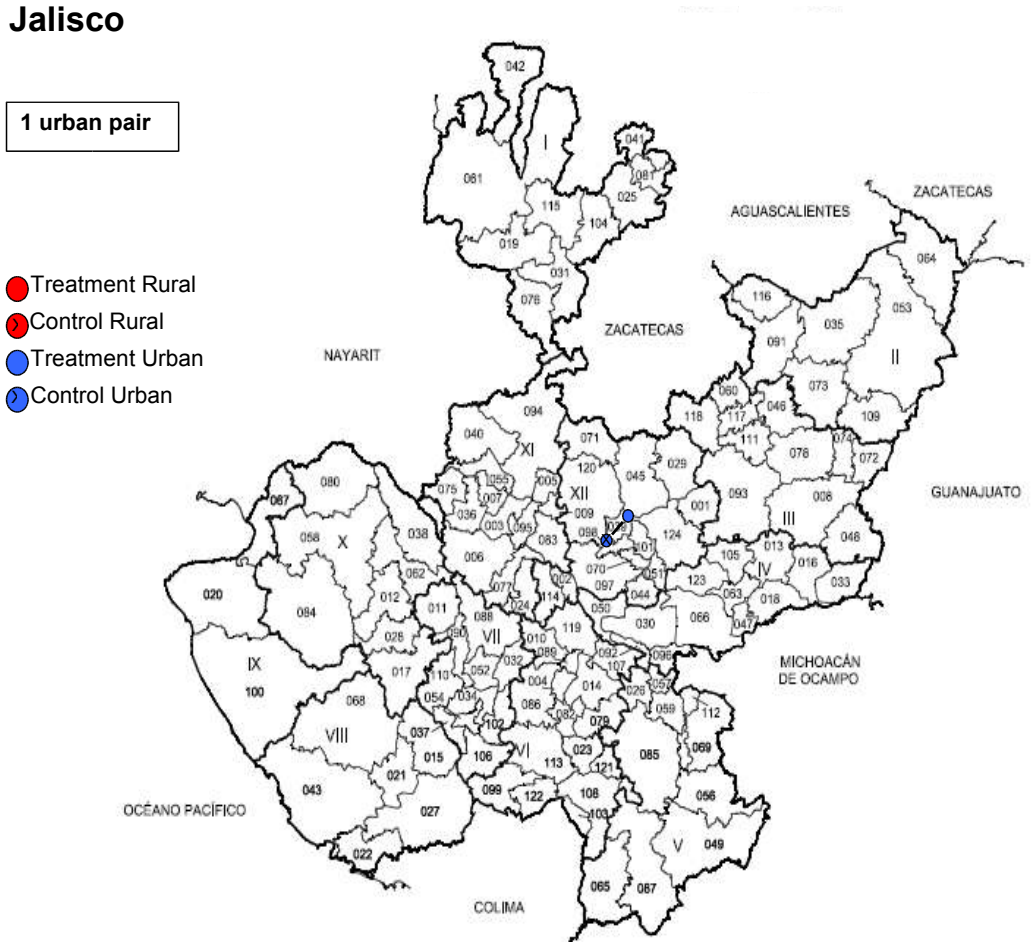
- 74 matched treatment-control pairs in the evaluation: 55 rural and 19 urban in 7 states

State	Rural Pairs	Urban Pairs	Total
Guerrero	1	6	7
Jalisco	0	1	1
México	35	1	36
Morelos	12	9	21
Oaxaca	3	1	4
San Luis Potosí	2	0	2
Sonora	2	1	3
Total	*55*	*19*	*74*

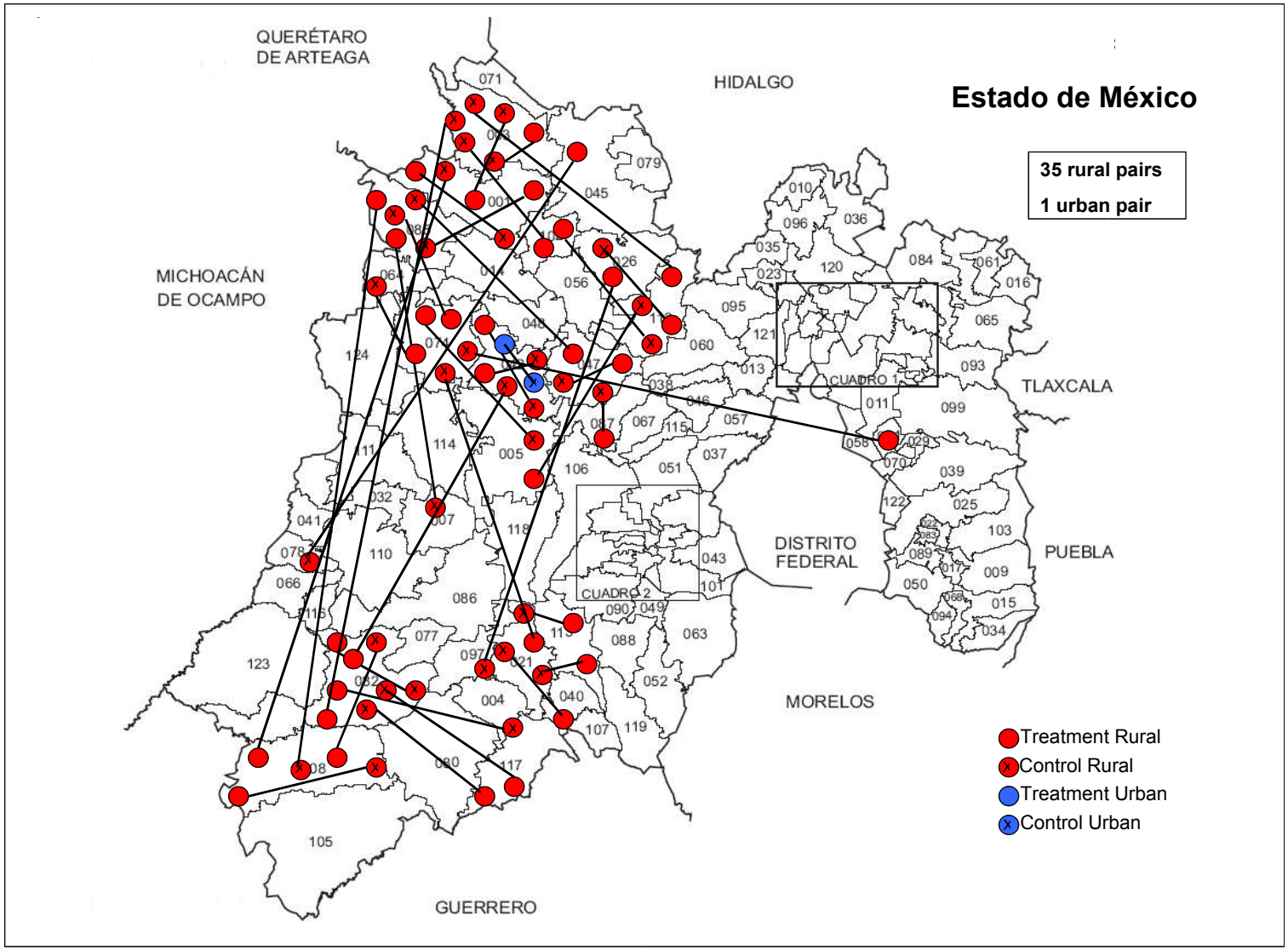
Matched Pairs, Guerrero



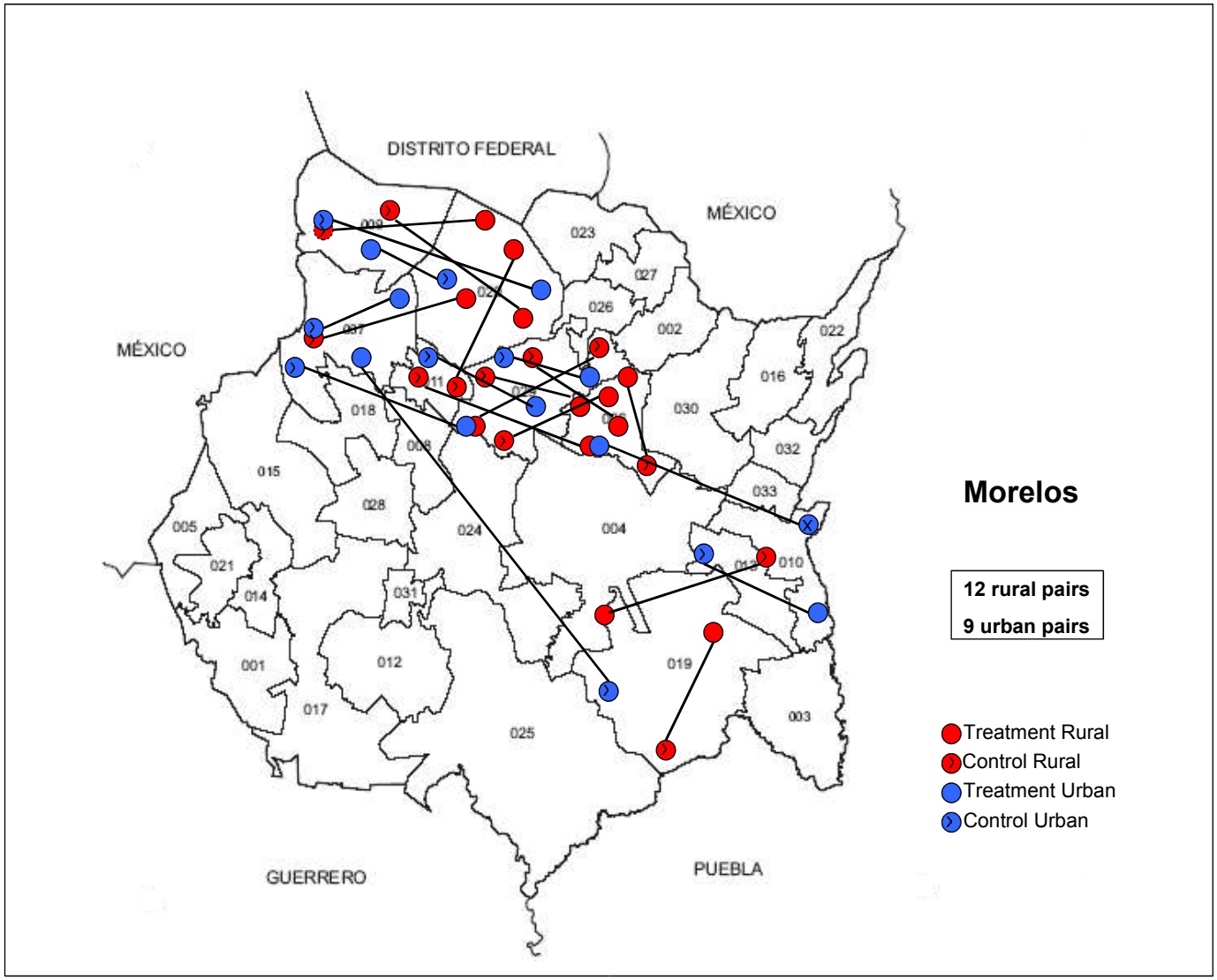
Matched Pairs, Jalisco



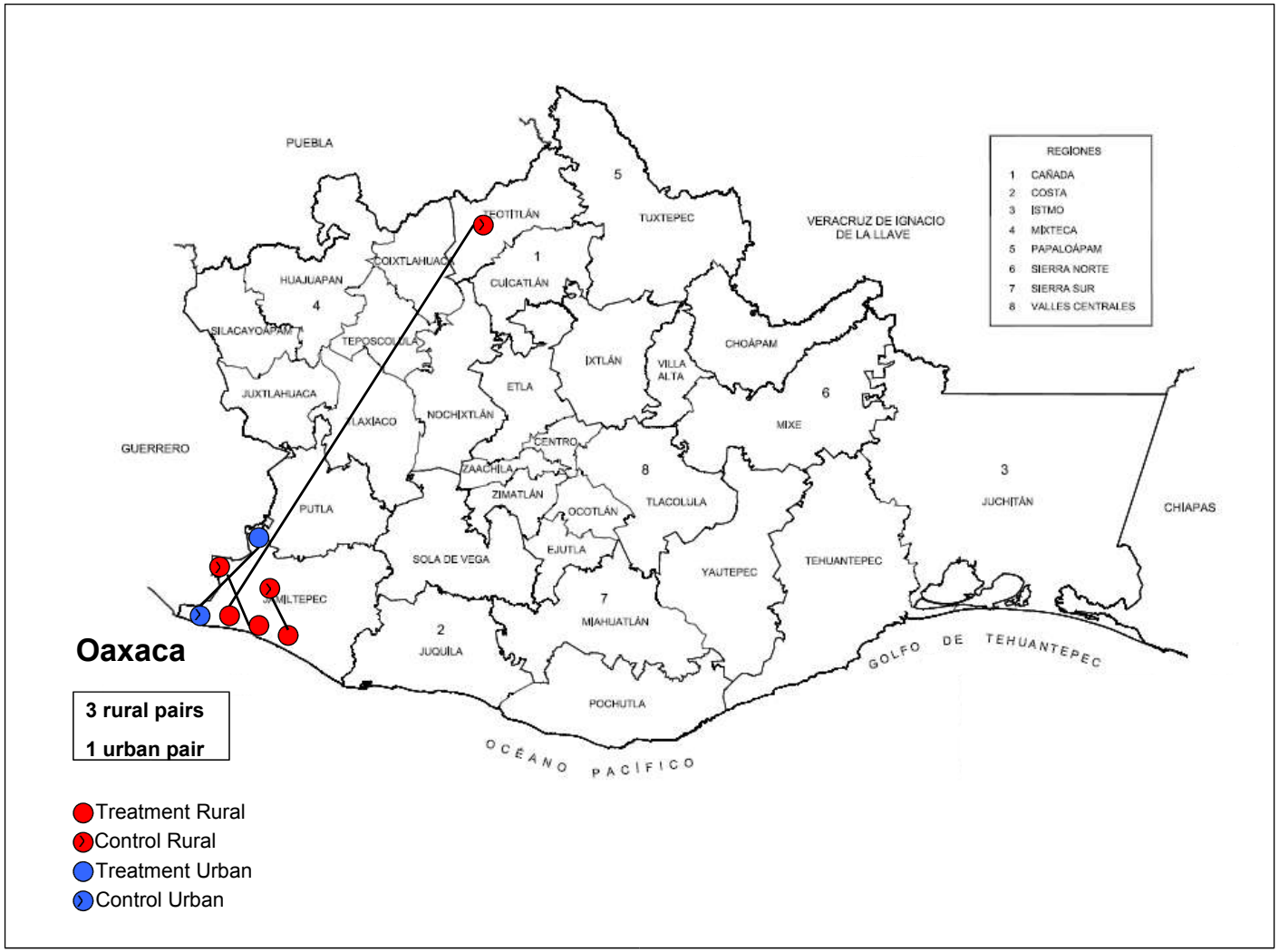
Matched Pairs, Estado de México



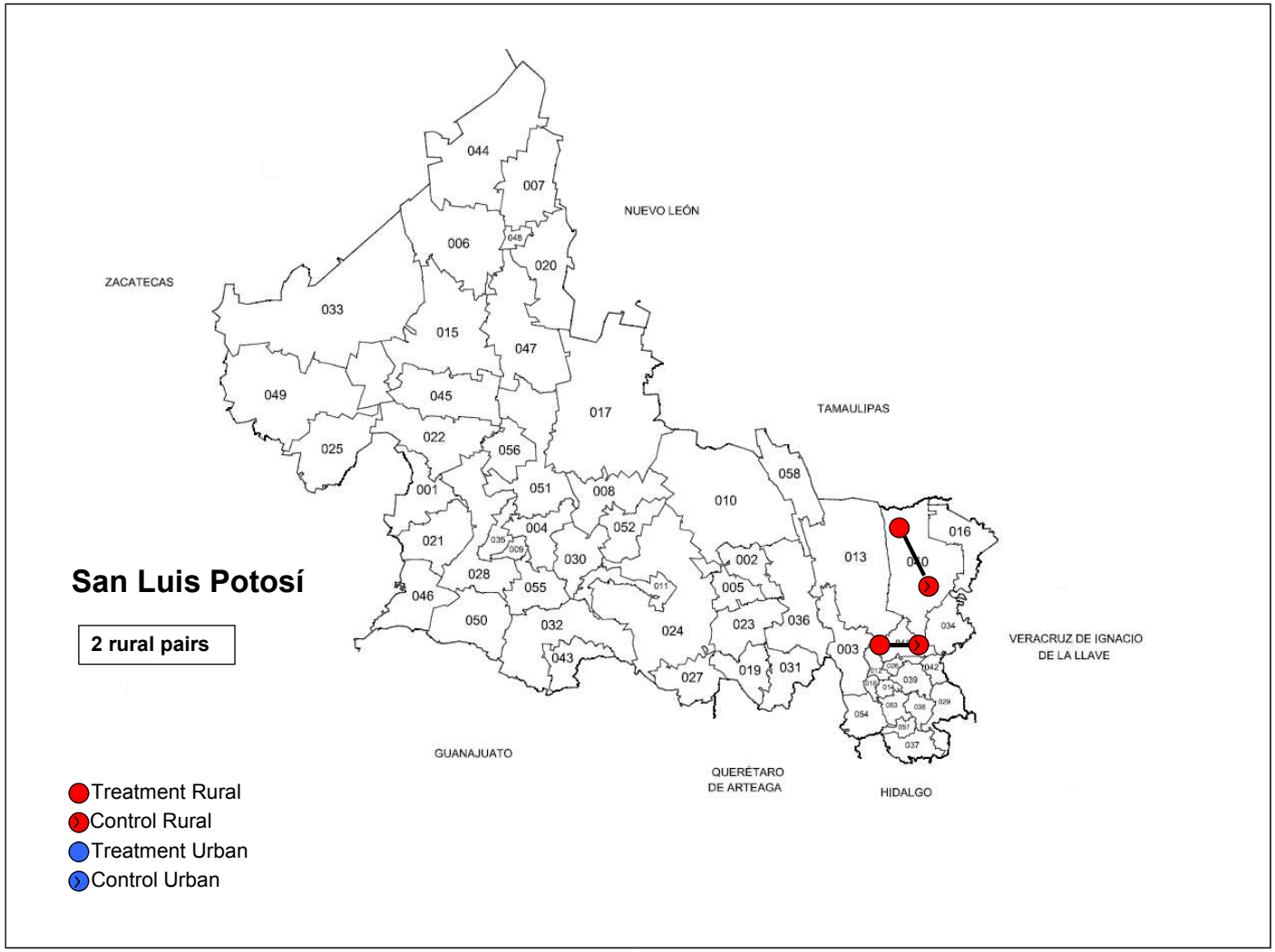
Matched Pairs, Morelos



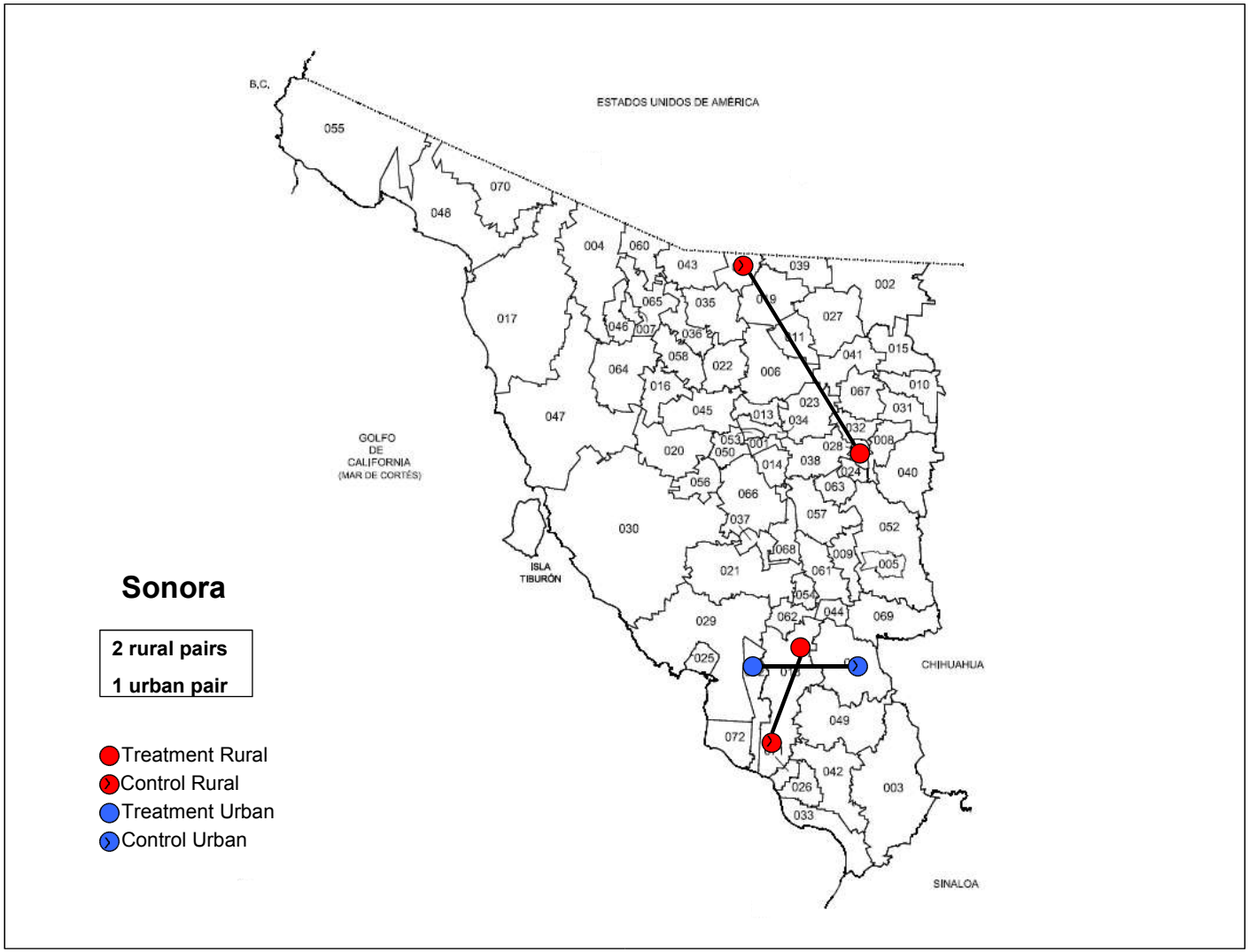
Matched Pairs, Oaxaca



Matched Pairs, San Luis Potosí



Matched Pairs, Sonora



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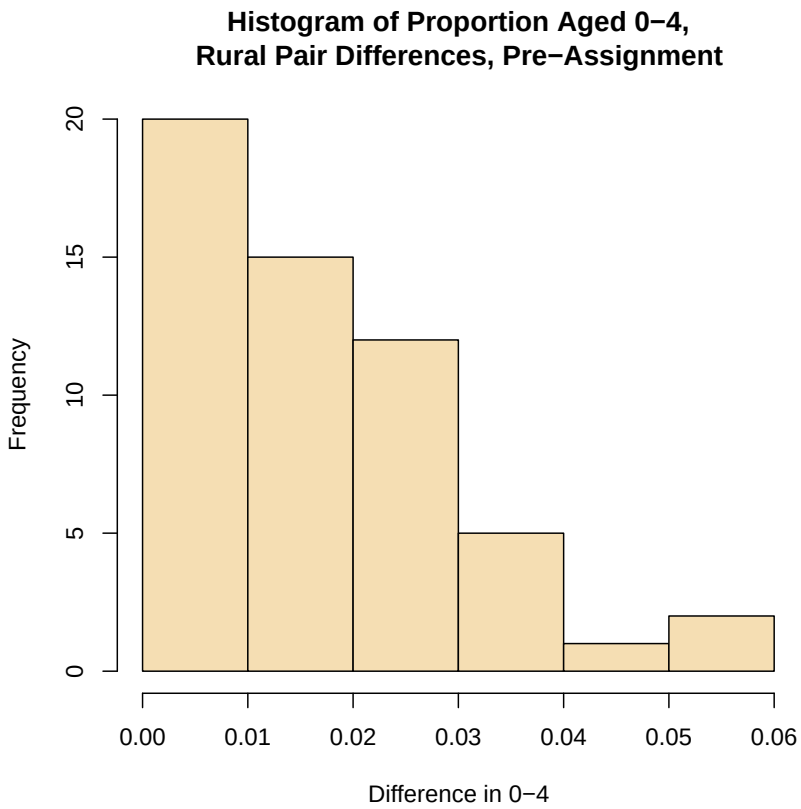
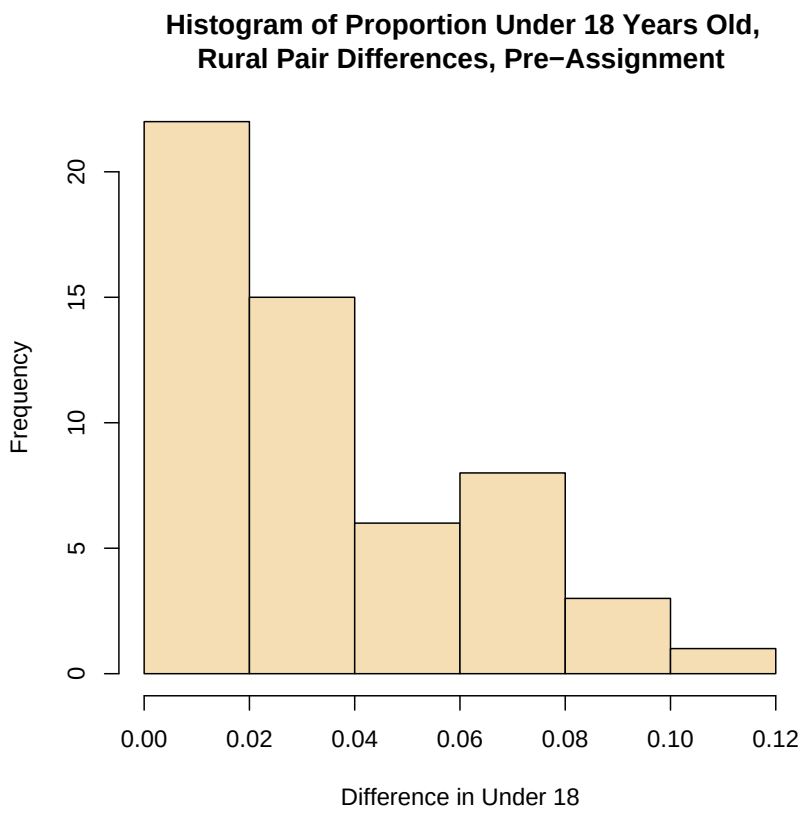
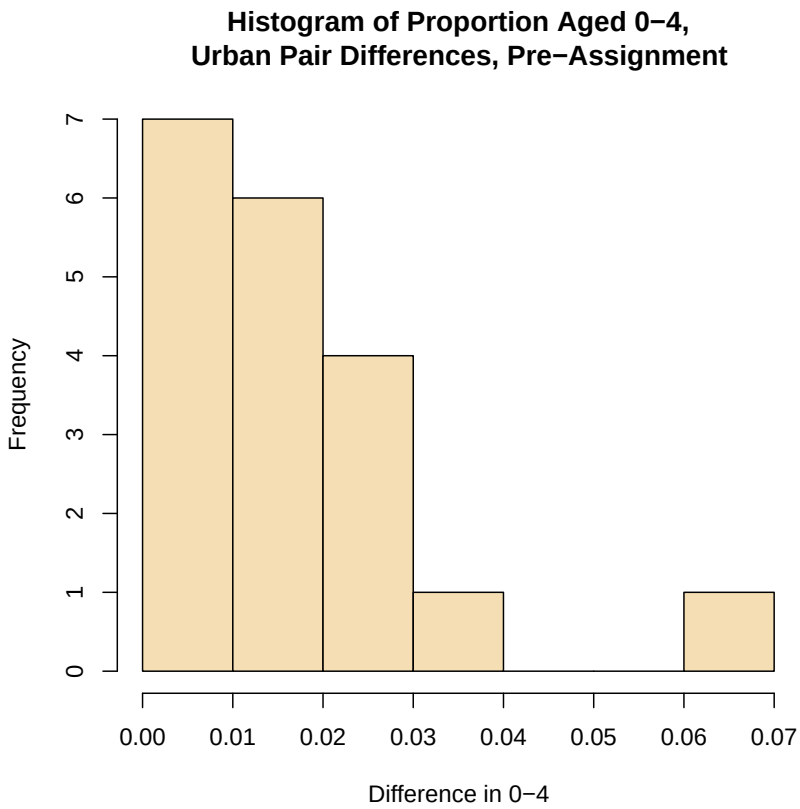
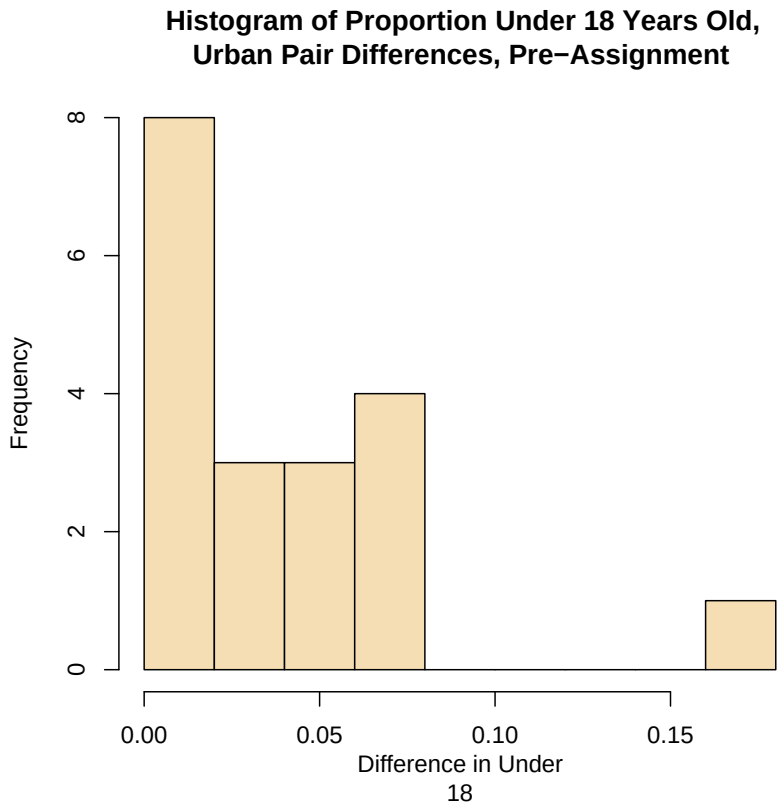
Age Differences in Rural Pairs



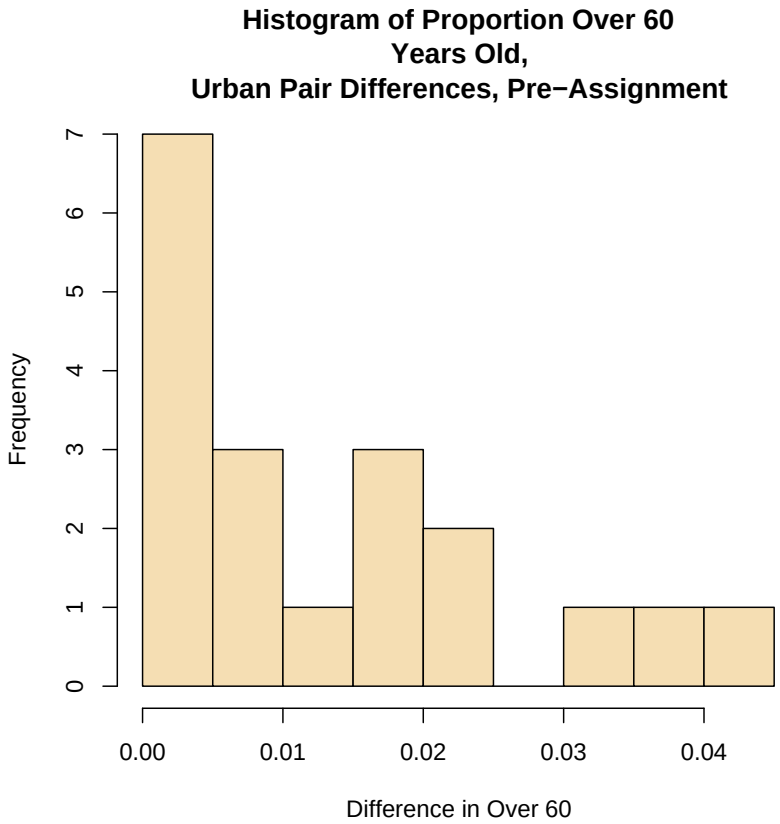


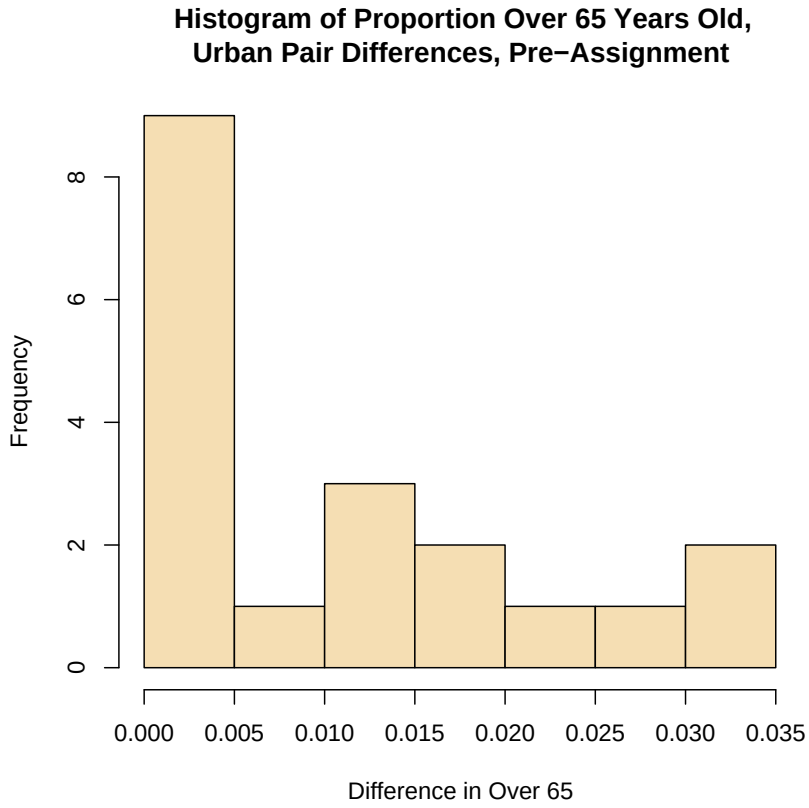
Age Differences in Urban Pairs



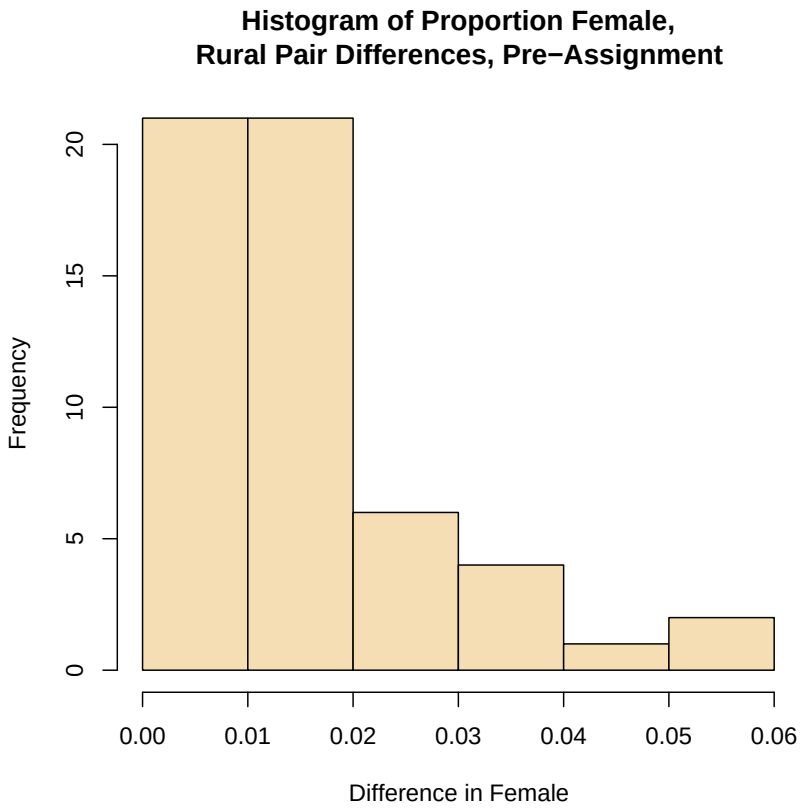


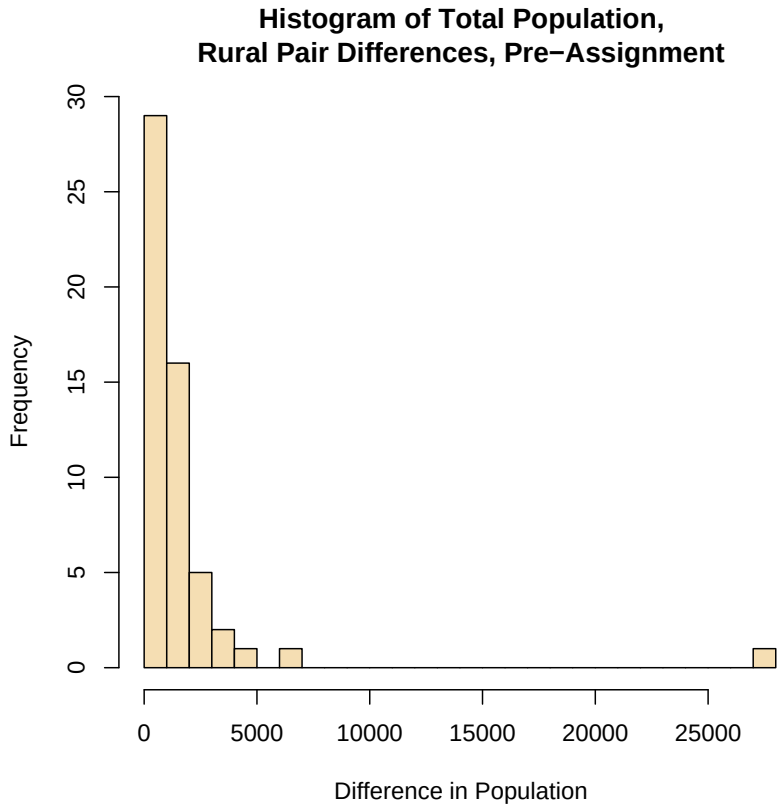
Age Differences in Urban Pairs, II





Demographic Distances in the Rural Pairs





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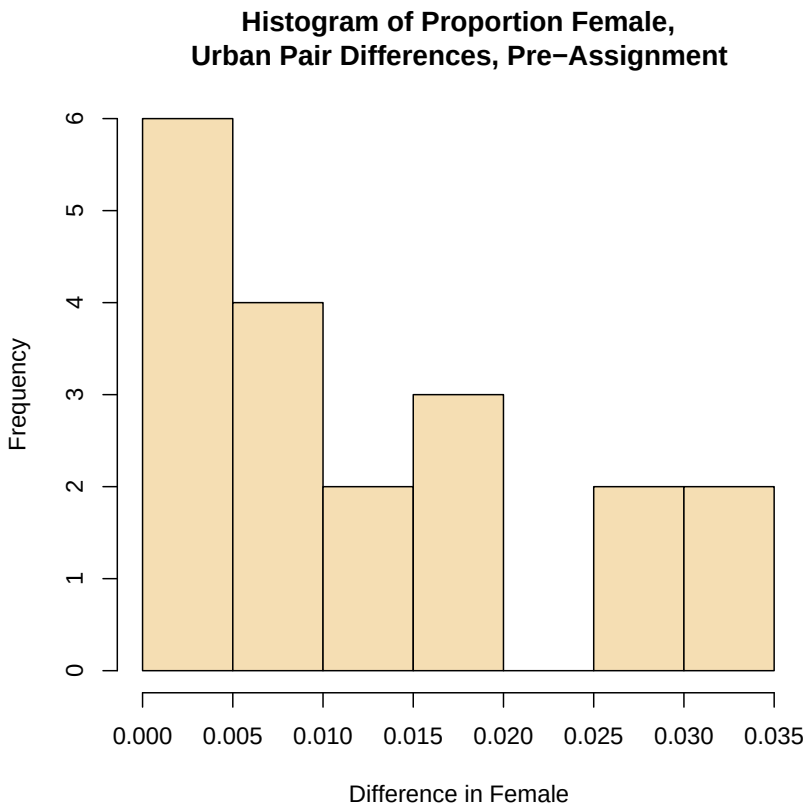
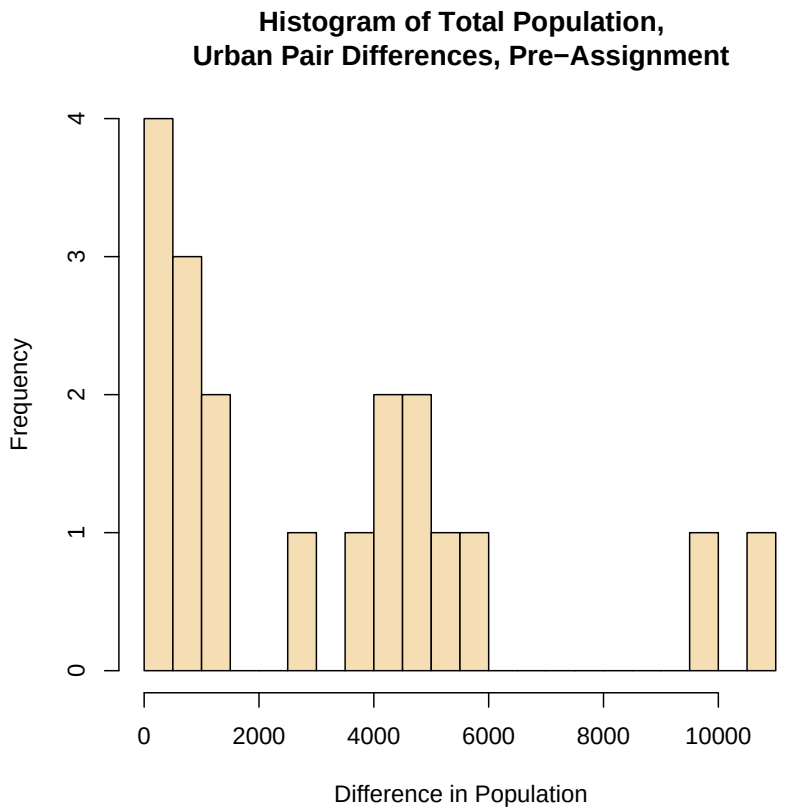
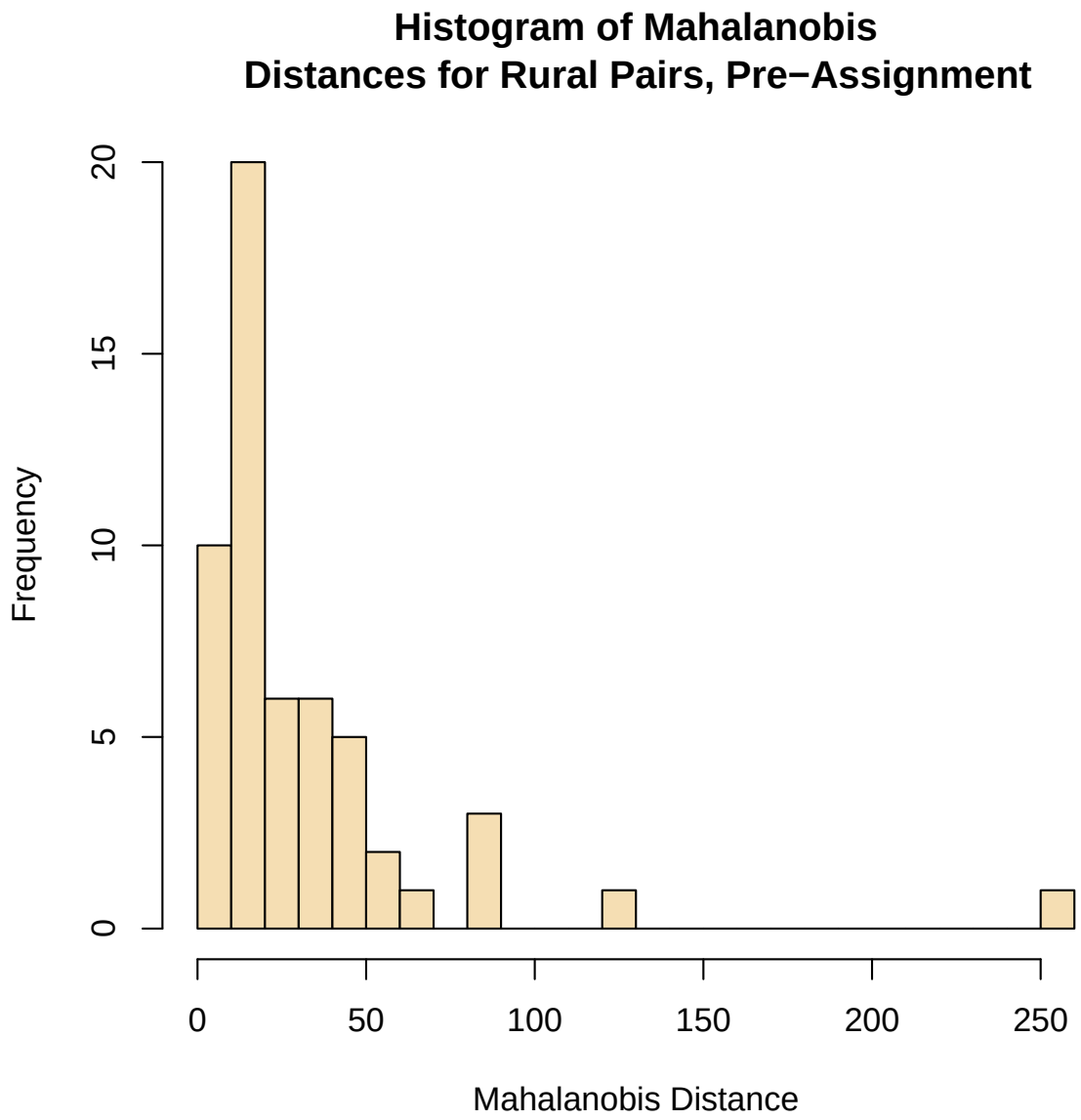




Total Multivariate Distances Within All 55 Rural Pairs



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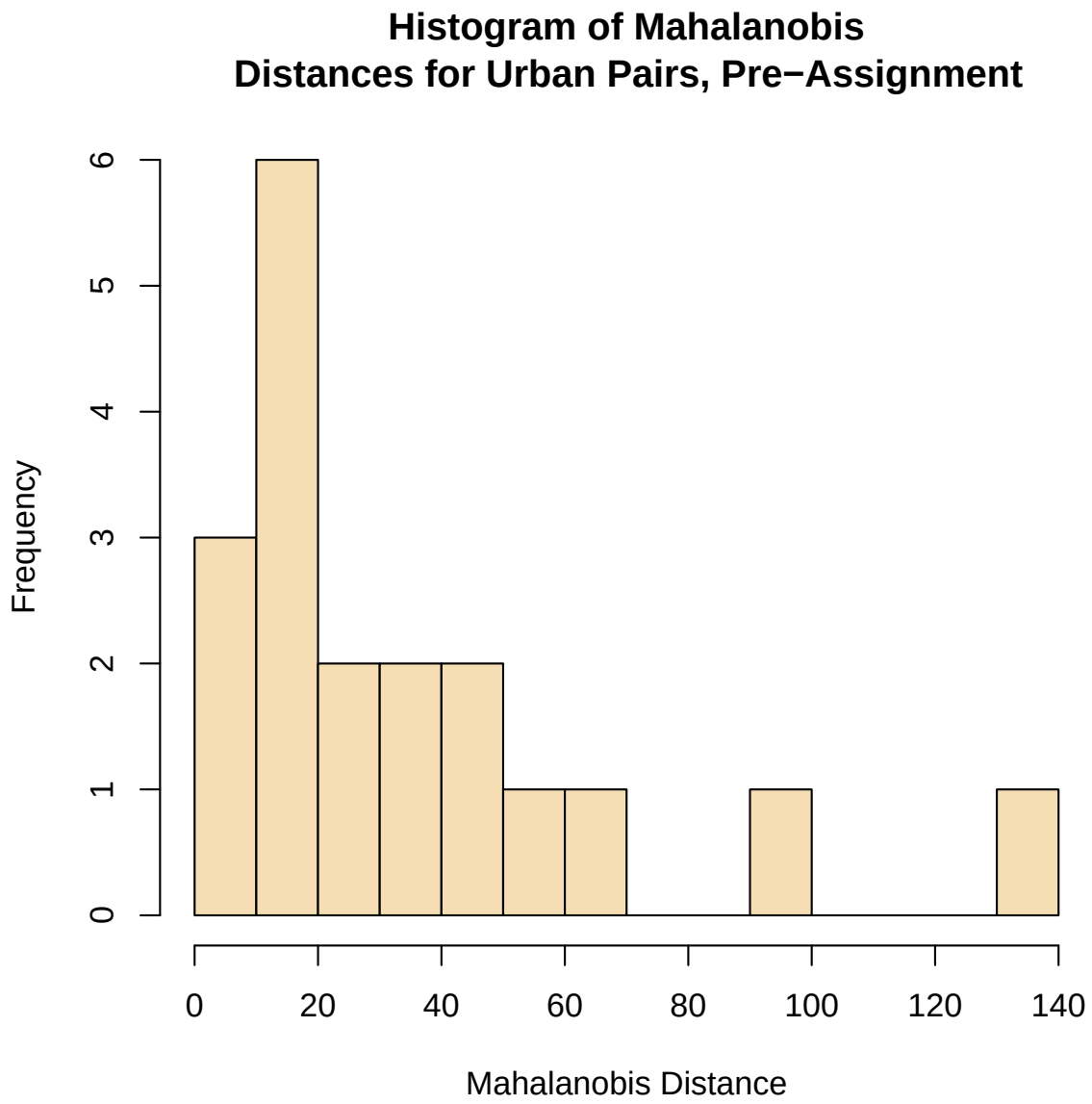
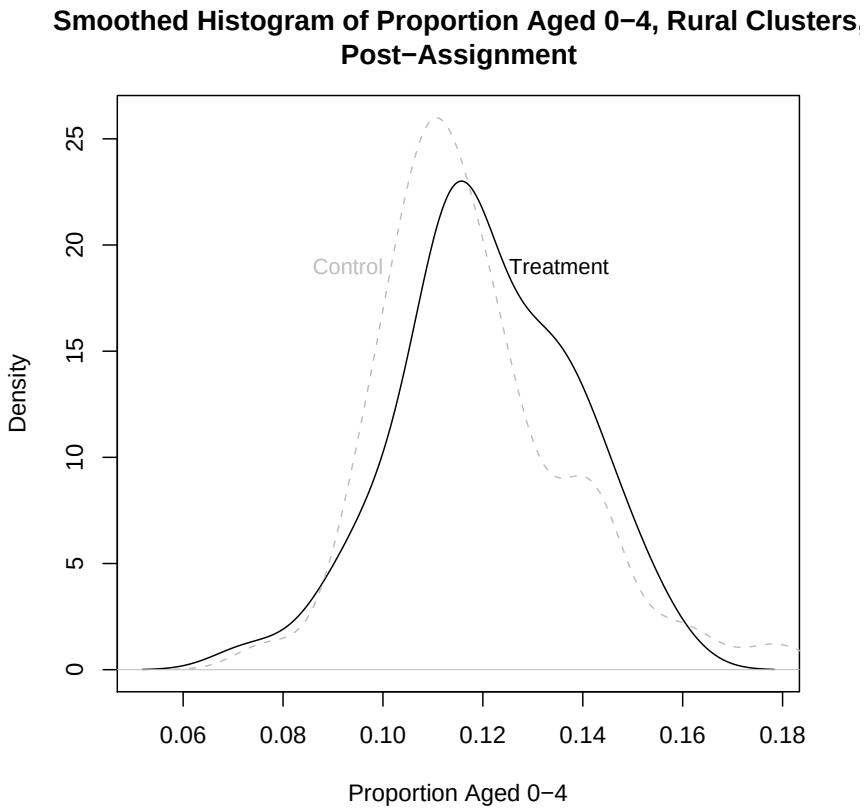
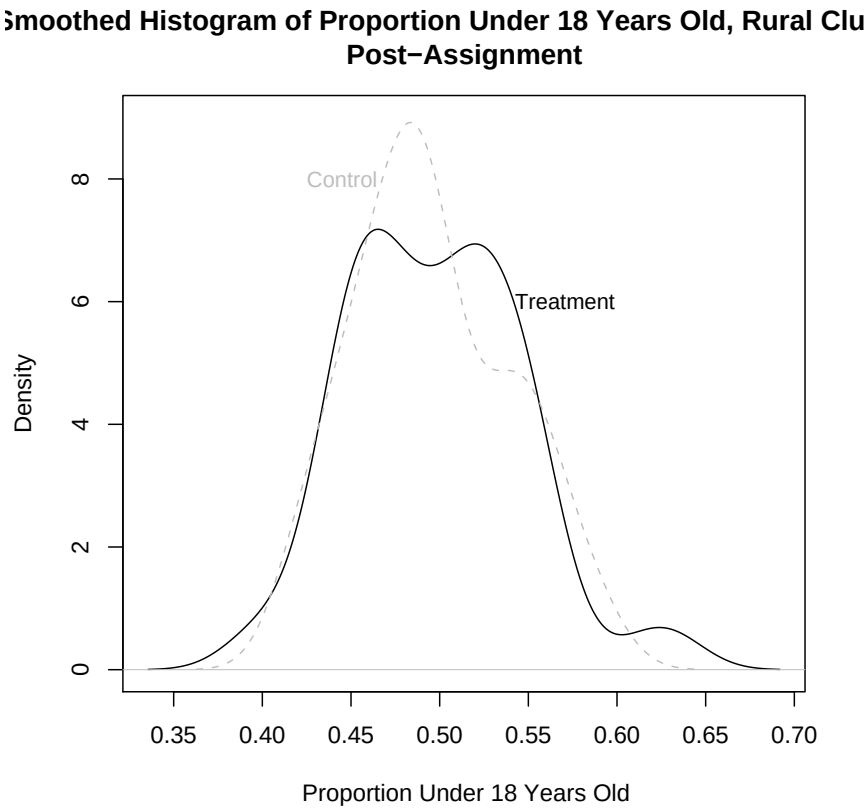


Rural Age Balance After Randomization





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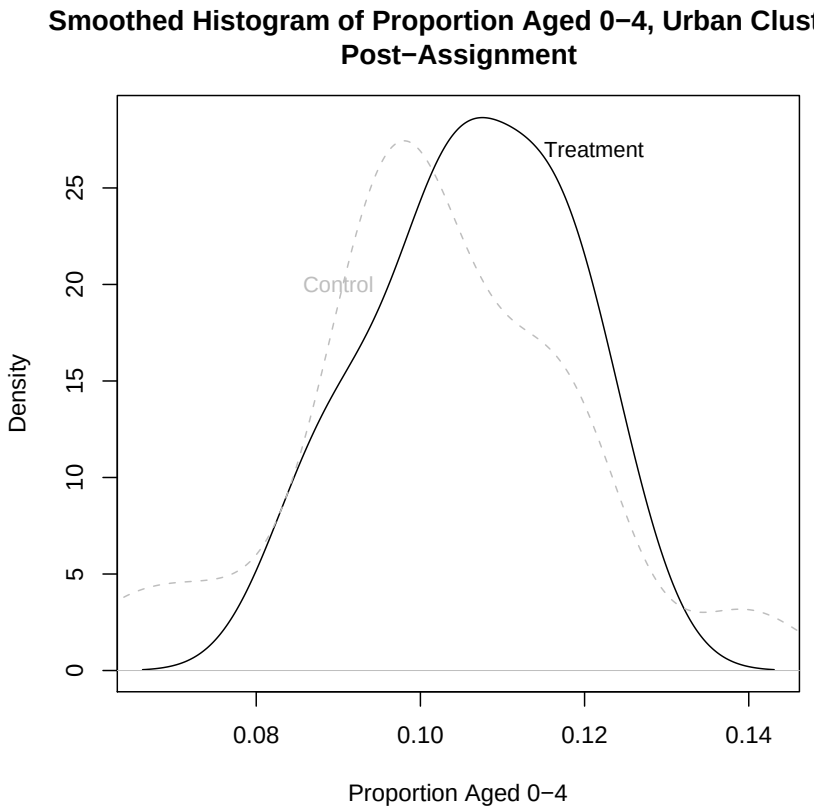

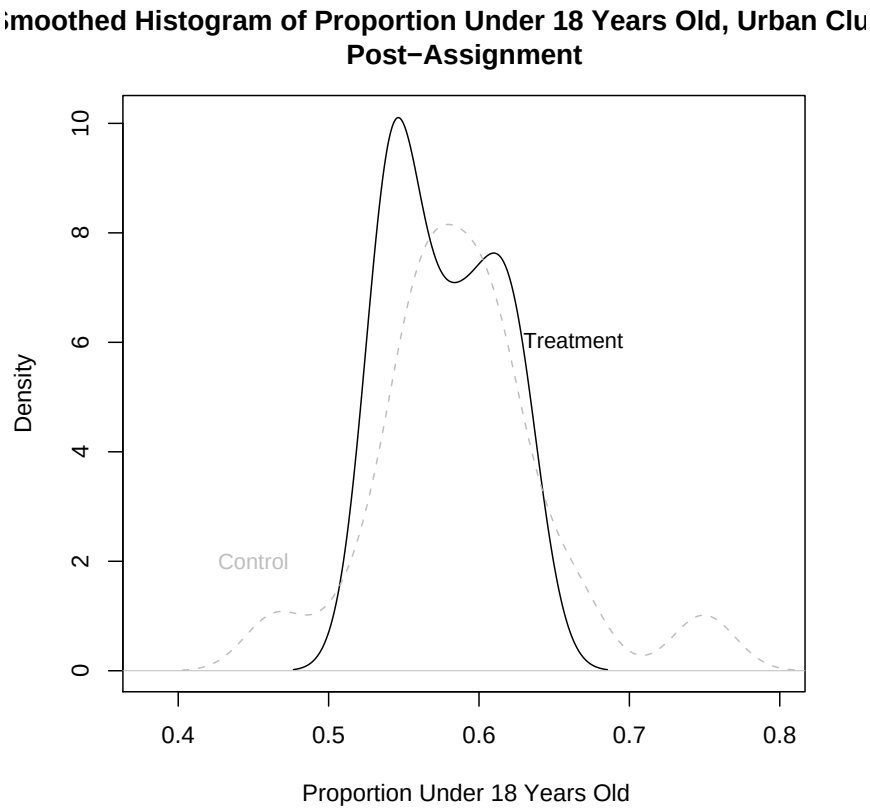

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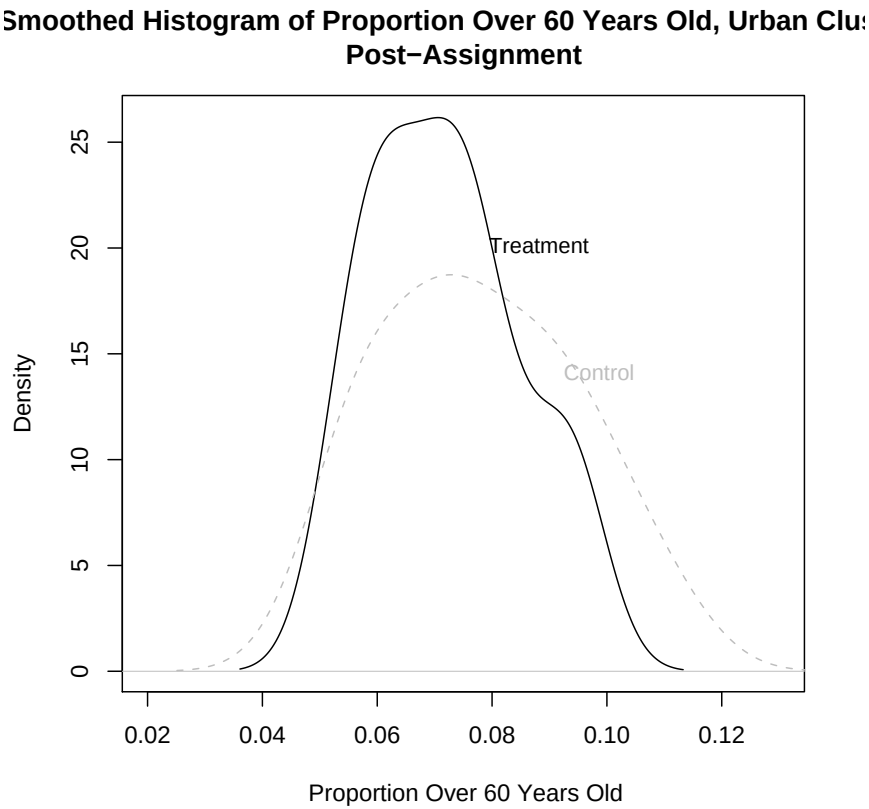

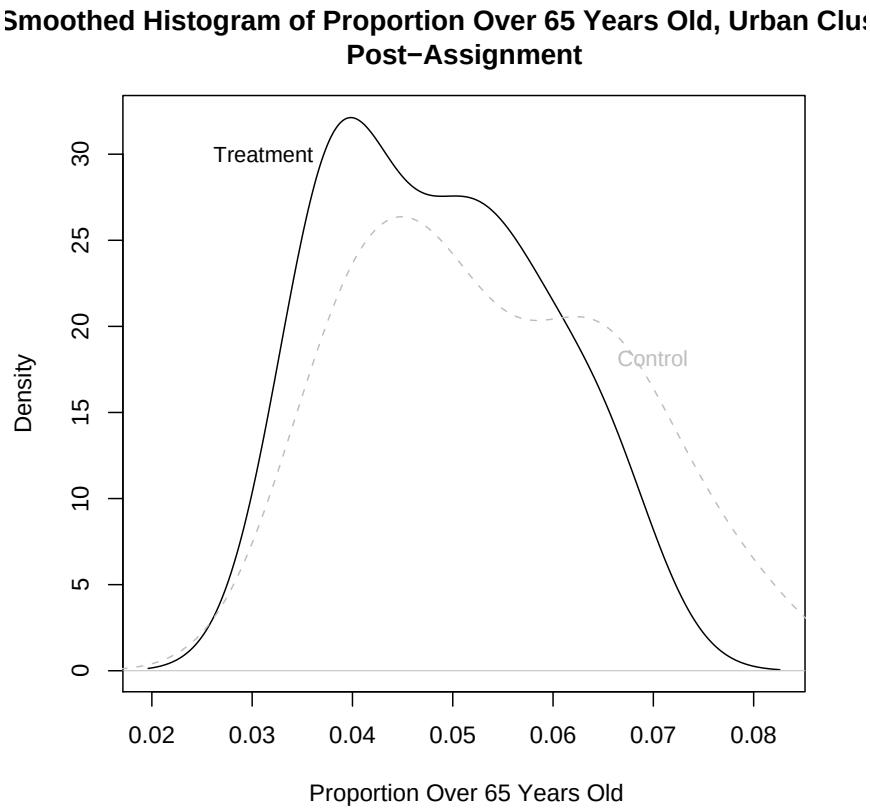

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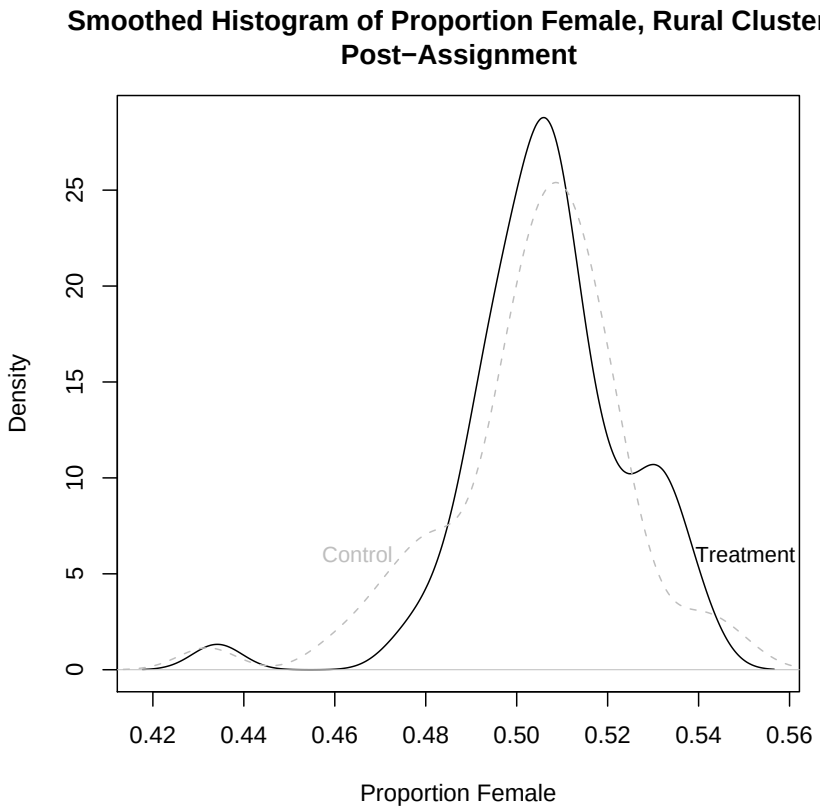

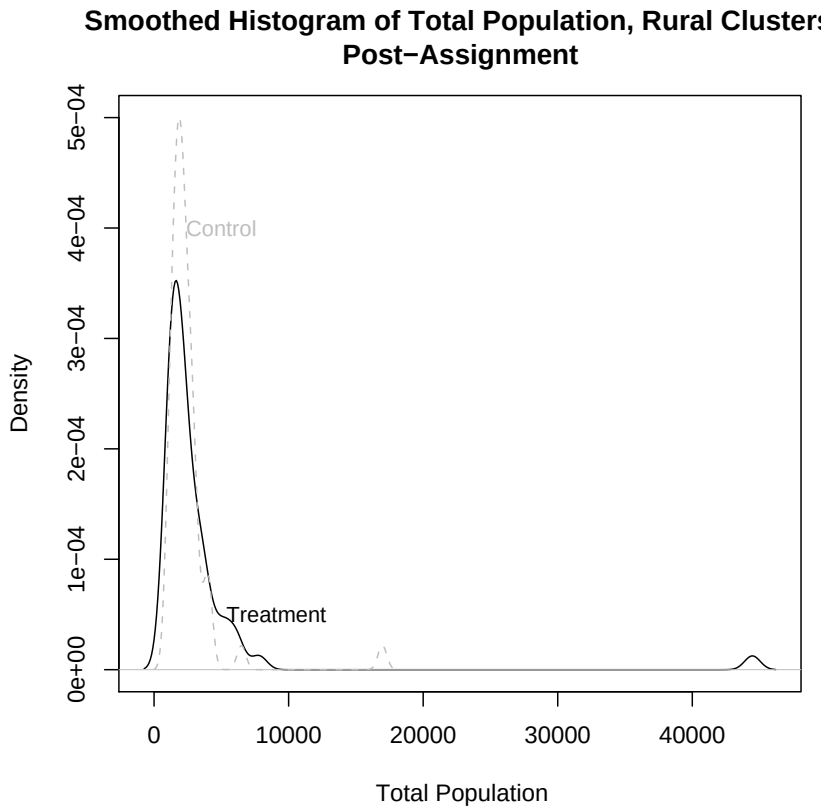

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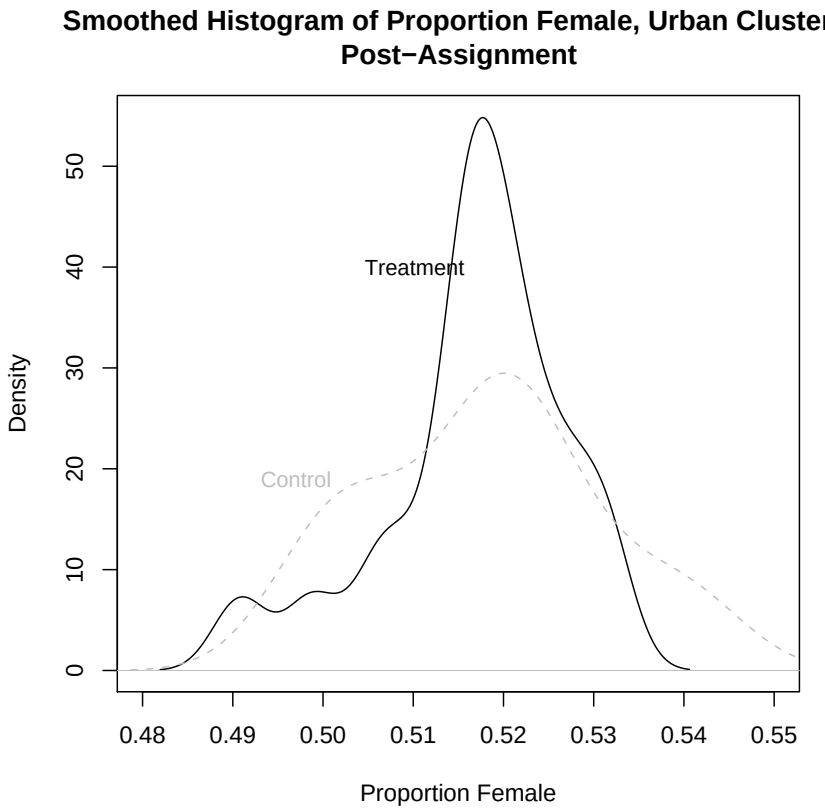

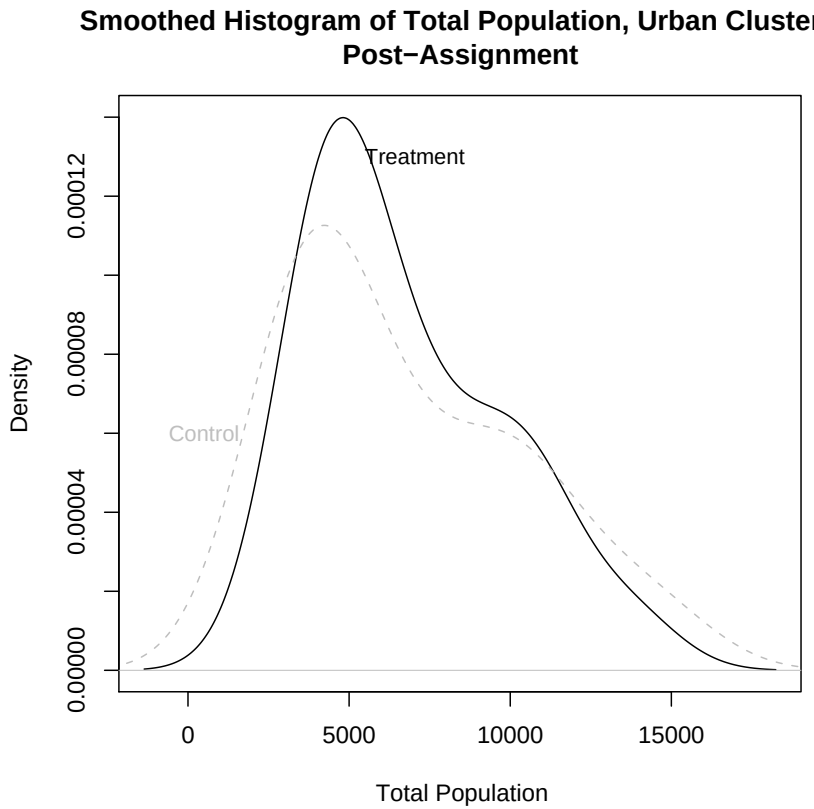

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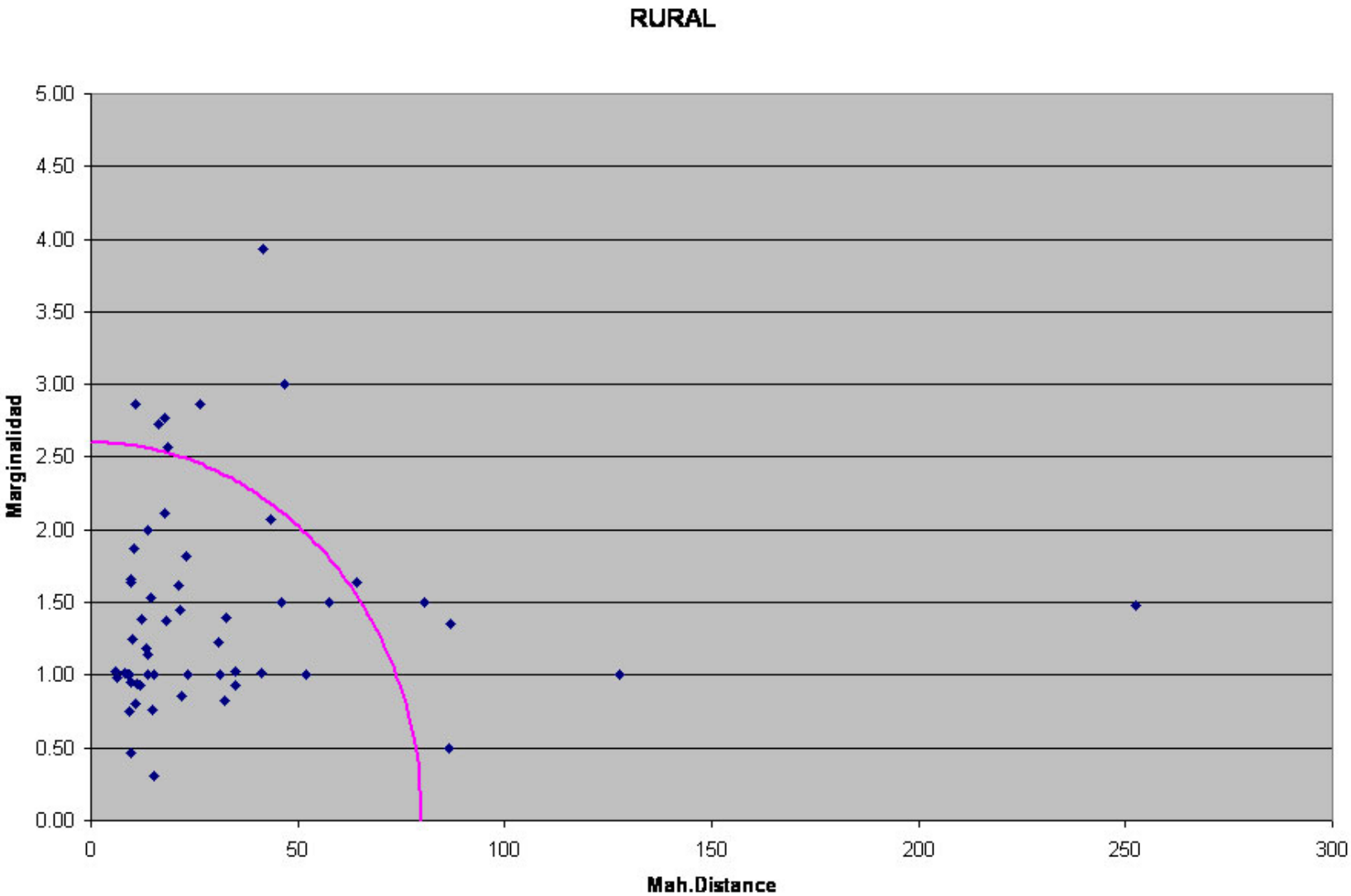
Choosing Pairs for the Survey



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For more information

http://GKing.Harvard.edu