

Does Adult Redeploy Illinois change the rate of prison use in jurisdictions with ARI programs?

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Adult Redeploy Summit
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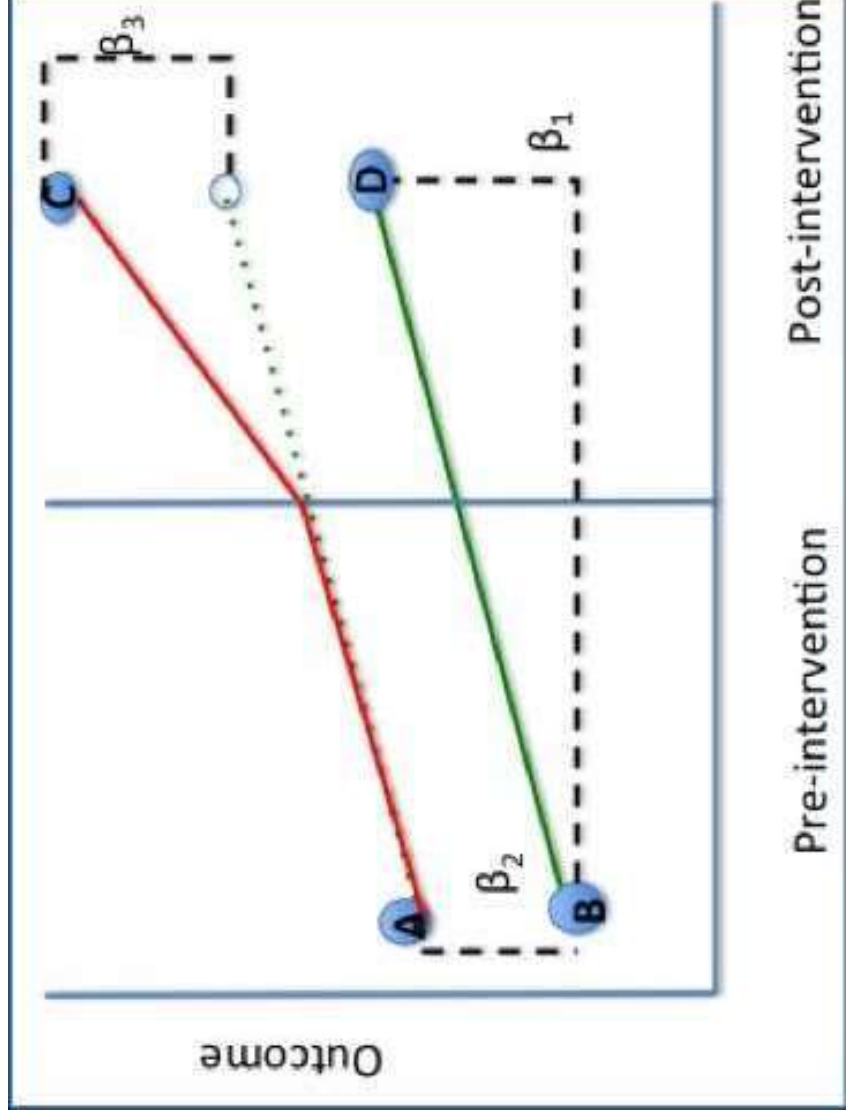
What did we attempt to learn?

- Simple question: Did prison use change for ARI counties?
- What should we see if ARI is changing prison usage?
 - Changes in incarceration rates.
 - Changes in felony sentencing.
- Can we attempt to answer this with available data?
 - To a limited extent... YES!

What data did SPAC have to answer this?

- SPAC did **not** have ARI data on participants.
- SPAC did know when each site started enrolling clients.
- SPAC did have prison and sentence data
 - Detailed IDOC prison admissions data, high quality
 - CHRl conviction and sentence data, quality that varies across counties
- SPAC started with three models in mind:
 - Difference-in-Differences of prison admission rates
 - Synthetic Control method of prison admission rates
 - Multilevel Logistic Regression of presence or absence of a prison sentence for a conviction using CHRl sentence data
 - Less preferred due to data issues

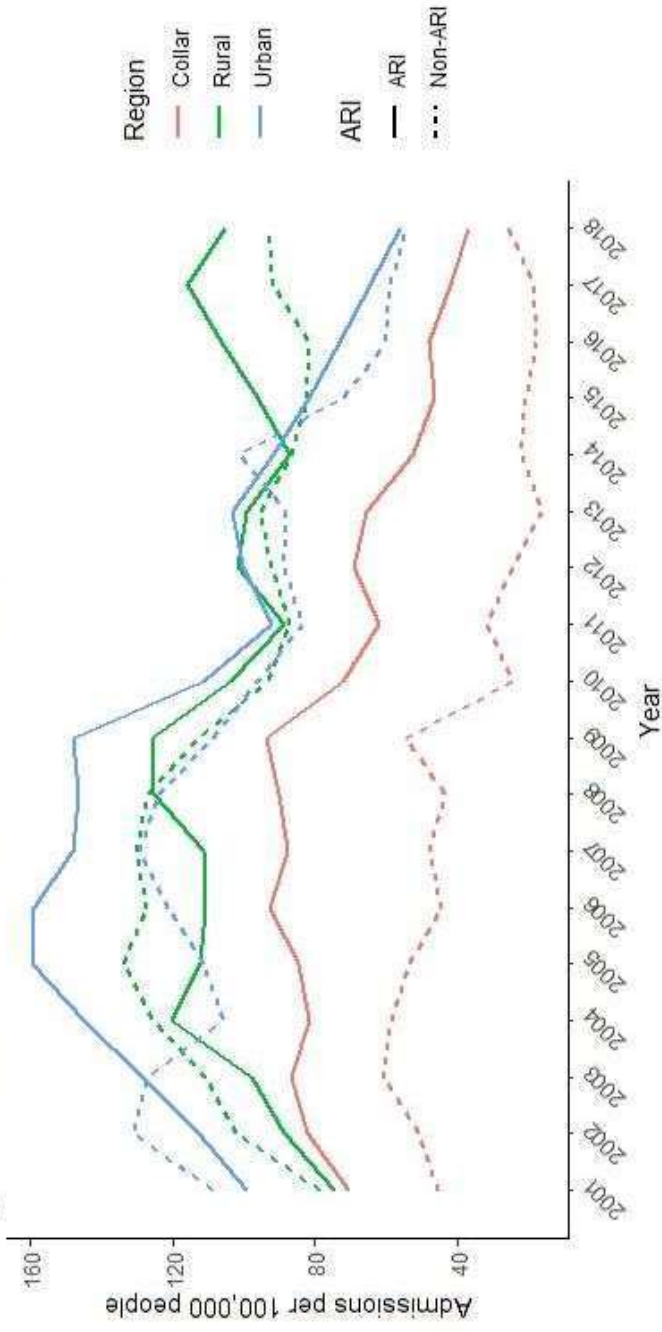
Difference-in-Differences Design



<https://www.publichealth.columbia.edu/research/population-health-methods/difference-difference-estimation>

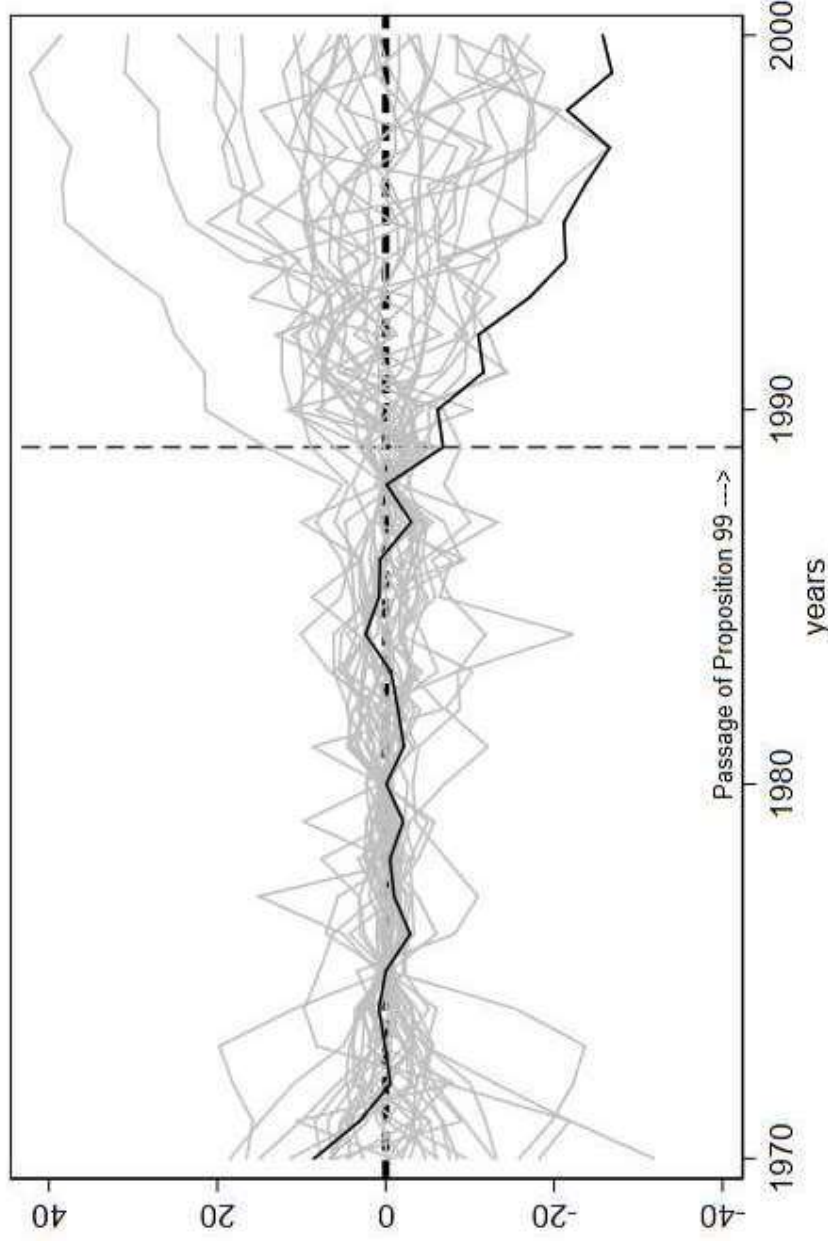
Challenge - trend data for IDOC admission rates

Figure 2 - Class 3-4 Non-Violent Admission Rates



Difference-in-differences assumes that the average outcomes of treated and control counties would have followed parallel paths if there were no intervention.... These appear problematic!

Synthetic Control Design



Re: Figure 5. Per-cap cig sales gaps in California & placebo gaps in 34 control states:

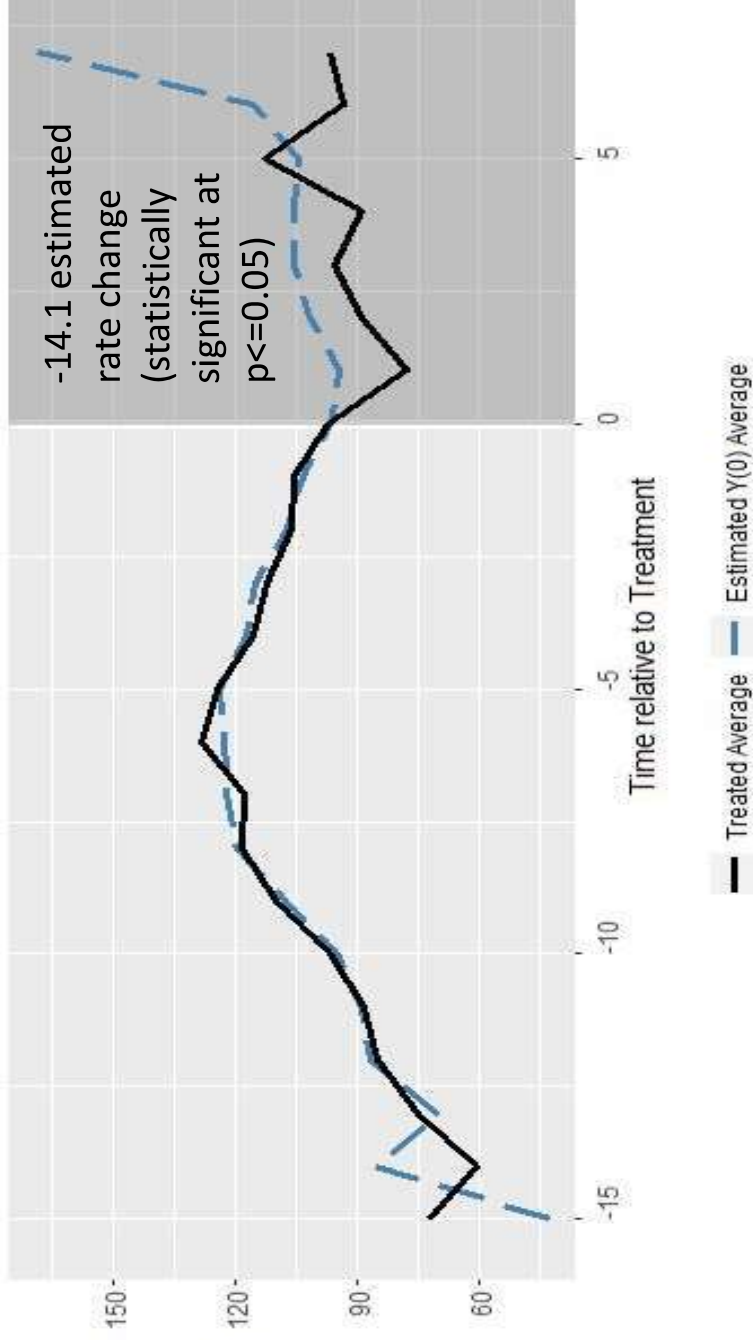
<https://www.solomonegash.com/wp-content/uploads/myreplications/synth/synthdid.html>

Solution – Generalized Synthetic Control (GSC)

- Xu, Y. (2017). Generalized Synthetic Control Method: Causal Inference with Interactive Fixed Effects Models. *Political Analysis*, 25(1), 57-76. doi:10.1017/pan.2016.2
- Recent development of generalization of synthetic control method.
- Allows multiple treated units in one model even with varying start times.
- Creates weighted counterfactuals from comparison counties to relax the difference-in-difference assumption of parallel trends.
- Allows SPAC to use the higher quality IDOC data as a primary model, with the logistic regression model as a check to see how sensitive the results are to modeling decisions.

GSC results for class 3-4 non-violent prison admission rates

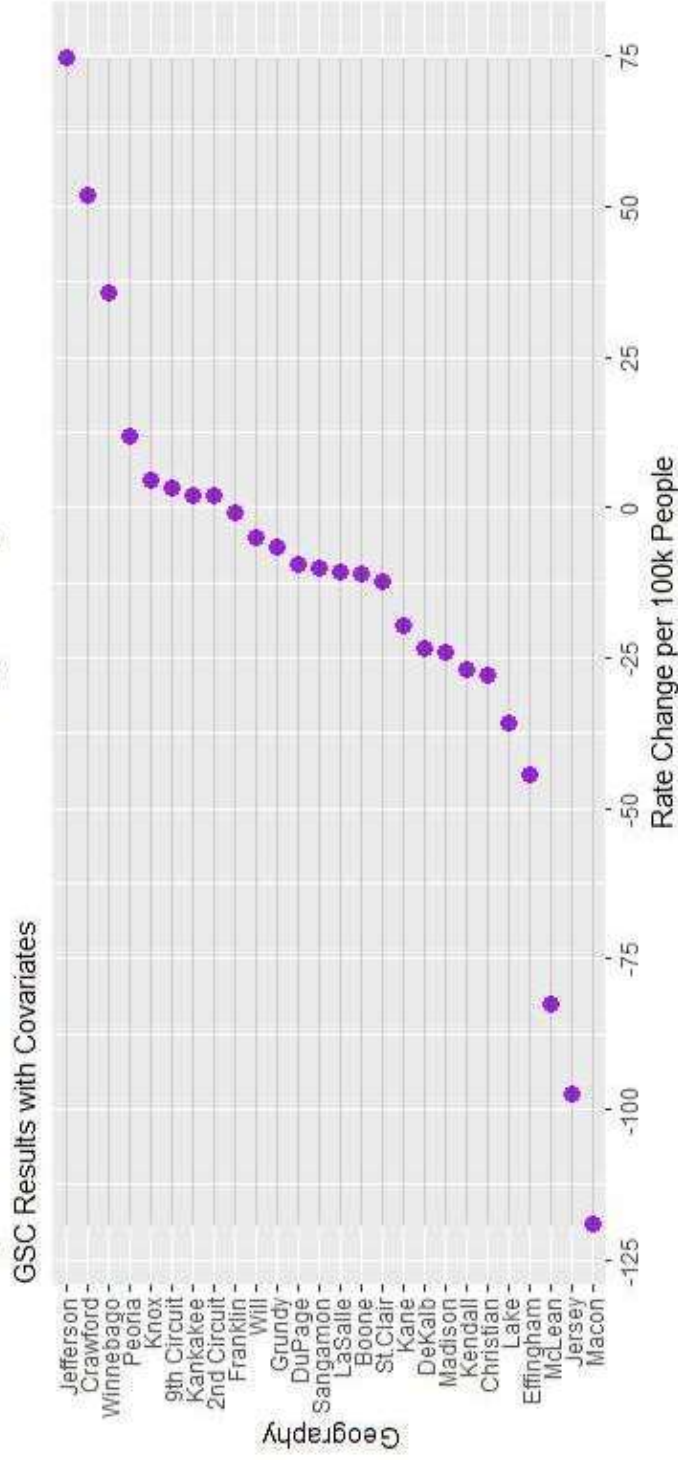
Figure 3 - Prison Admission Rates and ARI



		GSC Prison Admissions Rate, with covariates	
		ATT	s.e.
Overall Average		-14.1	6.5
Year	ARI Sites		
-9	-	2.2	4.5
-8	-	-1.0	4.5
-7	-	-4.4	4.6
-6	-	5.7	4.7
-5	-	0.7	4.7
-4	-	-1.4	4.5
-3	-	-2.3	4.3
-2	-	-1.0	3.8
-1	-	1.9	4.0
0	-	0.1	4.1
1	26	-17.8	5.9
2	26	-14.4	6.5
3	24	-10.1	7.3
4	22	-18.0	8.5
5	20	-3.3	9.9
6	14	-3.2	12.3
7	9	-40.9	16.7
8	4	-66.9	25.2

Effect varies by geography

Figure 6 - Median Estimated Rate Change During ARI Years



Although there are some outliers in both direction that largely cancel each other out, most counties have estimated changes between -50 and 15, with most showing reductions in the prison admission rate.

Model Two: Multilevel Logistic Regression

SPAC used a multilevel model with a random intercept at the county level and for the conviction year as well as a random slope for the ARI effect on counties. The unit of analysis was a non-violent class 3 or 4 conviction. Individual-level variables are:

- Prison admission (presence or absence, **outcome variable**)
- ARI active at the time of conviction (yes or no)
- Offense Class (3 vs. 4)
- Black (black vs. non-black)
- Male (male vs. female)
- Arrest age mean-centered
- Years between the arrest and conviction mean-centered
- Weapons conviction indicator (any weapons conviction on the DCN)
- Any prior prison sentence indicator (yes or no)
- Prior arrests
- Prior convictions
- Juvenile arrest indicator (first arrest was prior to 18th birthday)

Model Two: Multilevel Logistic Regression results

- Convictions in counties with ARI had a 19% reduction in the odds of receiving a prison sentence for Class 3 or 4 non-violent convictions compared to when ARI was not active.
- The marginal effect over the entire data set was also calculated from the model, predicting the outcome for each observation and finding the change in probability. This arrives at a statistically significant 3.3 percentage point reduction.
- For context, the baseline probability of a prison sentence for Class 3 or 4 non-violent convictions is around 35%—thus a 3.3 percentage point reduction appears reasonable.

Summarizing what we learned

- Both models produce similar results with different but closely related outcomes
 - GSC estimated a reduction of 14 admissions per 100,000 people from about 105 baseline...roughly a 13% decline
 - Regression estimated a reduction of 3.3% chance of receiving a prison sentence, from a 35% baseline...roughly a 10% decline
- The overall estimated effect from ARI intervention is not trivially small or excessively large
 - ARI sites did not always target all class 3 and 4 non-violent offenders and instead tailored target populations based on capacity and programming
- Although the analysis is positive for ARI, it has limitations.
 - Limited to class 3-4 non-violent
 - Hard to derive cost savings without more data
 - Not a replacement for an impact evaluation

Analysis will soon be available!

<https://spac.illinois.gov/>

