

# Prosecutorial Policy Analysis

## AI-Powered Measurement of Criminal Justice Reform in California

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## Thank You for the Second Look

Last time, you raised two concerns: whether **outcome data** could support the causal questions, and whether the project was **targeted enough** toward specific causal research questions.

Those concerns were fair. Here is what has changed:

1. **We linked outcome data and found a real signal** — pilot results survive year fixed effects; pretrial detention shows a clean causal signal ( $p_{\text{pre-trend}} = 0.90$ )
2. **We identified 6 specific causal designs** aligned to specific research questions — each one mapped to the exact outcome data source it requires
3. **We restructured the ask** so that each funded phase directly unlocks a specific causal study, not just more infrastructure

The next 15 minutes will walk you through the evidence.

# The Measurement Gap

District Attorneys are among the **most powerful actors** in the criminal justice system, yet we lack systematic measurement of:

- How their policies **vary across jurisdictions**
- How policies **change over time**
- Whether stated policy intent **affects outcomes**

Existing research relies on case outcomes or campaign rhetoric—neither captures the *stated policy intent* that guides line prosecutors daily.

## What's Missing

No systematic, comparable measurement of internal DA policy documents exists **anywhere**.

## What We Built

The first large-scale, AI-coded dataset of internal DA policy documents—infrastructure for causal inference about prosecutorial ideology.

### 2,665 internal DA policy documents

obtained by ACLU of Northern California via the California Racial Justice Act.

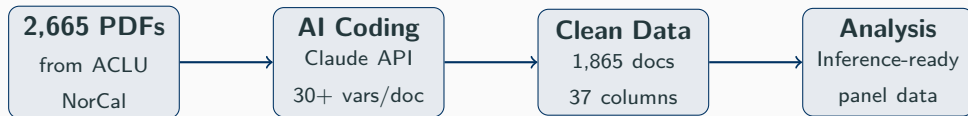
- **41 of 58** California counties represented
- Charging directives, sentencing memos, diversion protocols, racial equity initiatives, bail reform orders
- Documents span 2015–2024

These are *internal operating documents*—not press releases or campaign materials. They reveal what prosecutors actually tell their staff to do.

### Why This Is Rare

Internal policy memos are almost never publicly available. The ACLU's PRA effort created a **unique research opportunity** that may not persist indefinitely.

# What We Built: Research Pipeline



Dimension Coded	What's Measured
Ideological Orientation	7-point scale: clearly progressive → clearly traditional
Extensive Margin	Impact on <i>who enters</i> the system (charging, diversion)
Intensive Margin	Impact on <i>how severely</i> people are treated (sentencing)
Specific Policies	Diversion, bail reform, enhancements, three strikes, racial justice
Administrative Context	New policy vs. continuation, DA administration

Total API cost: ~\$80 · Processing time: ~2 hours

## Not Just “Progressive/Traditional” — We Code Specific Policies

Dimension	Values	Example from Our Data
Bail position	reform / high bail	Placer: reform-oriented bail memo; SLO: bail guideline update
Diversion	yes / no	Sacramento, Santa Clara: explicit diversion program support
Enhancements	minimize / maximize	SF, LA: minimize enhancements; 6 counties maximize
Three strikes	restrictive / expansive	Tracked per-document across all counties
Racial justice	high / moderate / low	9 counties high (post-2020 surge), 16 moderate, 10 low
Juvenile transfer	restrictive / expansive	4 documents with explicit positions
Alt. to incarceration	yes / no	Whether doc endorses treatment, community service, etc.

**This is v1.0.** Adding new dimensions (gun policy, DV protocols, immigration

# Face Validity: The Pipeline Recovers Known Patterns

## Gascón Transformation

LA County ideology score **tripled** under Gascón

Cohen's  $d = 0.75$ ,  $p < 0.001$

## Geographic Clustering

- **Progressive:** Sacramento +78%, Yolo +56%, San Diego +50%
- **Traditional:** Stanislaus -34%, Placer -21%
- Bay Area variation: Santa Clara +0.84 vs Alameda -0.15

## 2020 Racial Justice Surge

Racial justice emphasis jumped **+30pp** in one year (12% → 42%), tracking the post-George Floyd moment precisely. Documents with high racial justice emphasis are **4.6×** more likely to be progressive ( $\chi^2 = 421$ ,  $p < 0.001$ ).

## Why This Matters

Pipeline confirms patterns any expert would expect—evidence of validity, not artifact.

# Descriptive Analysis: Theory-Grounded Patterns

## Extensive > Intensive Margin

Recent reforms disproportionately emphasize **who enters** the system (**33.9%** extensive lenient) over **how severely** people are treated (**22.6%** intensive lenient). Political economy logic: diversion/declination less visible to voters than sentencing leniency—safer reforms for DAs facing reelection.

## Close Elections → Progressive Policy

Elections with margins  $\leq 15\text{pp}$  produce **+31.2pp** more progressive policies ( $p = 0.010$ ). Continuous:  $r = -0.50$  between margin and ideology ( $p = 0.009$ ). Provides credible first stage for instrumental variables designs.

**Novel contributions:** The extensive-over-intensive pattern has not been documented at this scale. The election-ideology link provides the first stage for causal designs.



# The Variation Exists: Policy Shocks = Natural Experiments

## What We Found

- **9** sharp policy disruptions (2020–23)
- **347** novel reform adoptions tracked
- Progressive docs: 18% → 56% in 4 years

## Each Feature → a Causal Design

- 9 disruptions → **staggered DiD / event study**
- ~5 close elections → **RDD first stage**
- DA transitions → **synthetic control**

## The Key Insight

**All four pillars of modern causal inference**—DiD, RDD, synthetic control, and instrumental variables—are available in this data. No other prosecutorial dataset offers this.

## What's Missing

Outcome linkage (Extension 2) and geographic scale (Extension 1) to power these designs. The policy variation is already in hand.

# Pilot Results: The Signal Is Real

Finding	Estimate	Significance	Survives Controls?
Ideology $\leftrightarrow$ Jail Pop Rate	$r = -0.222$	$p = 0.009$	Yes (year-demeaned)
Ideology $\leftrightarrow$ Jail Admissions	$r = -0.221$	$p = 0.009$	Yes (year-demeaned)
Progressive vs Traditional	$-68.5/100k$	$d = -0.81$	Yes
LA Pretrial (Gascón DiD)	$-32.1/100k$	Pre-trend $p = 0.90$	Cleanest

## What Works

All associations **survive year fixed effects**—not a COVID artifact. Pretrial detention shows the cleanest causal signal (pre-trend  $p = 0.90$ ). These results motivate the full-scale designs on the next slides.

## What Extensions Will Solve

Pilot  $N$  is small (34 CA counties, 9 years). National expansion + case-level outcome data will power the causal designs this variation makes possible.

# The Payoff: 6+ Credible Causal Studies, Ready to Run

Policy Coded	Causal Question	Design	Outcome Data
Bail reform	Does reform reduce pretrial detention without increasing FTA?	Event study; DiD	Vera ( <a href="#">done</a> ); CA DOJ
Diversion	Do diversion memos lower recidivism?	Staggered DiD	UniCourt; CJARS
Declination memos	Do “decline to prosecute” orders lower filings?	Synthetic control	CA DOJ ( <a href="#">free</a> )
Enhancement reform	Does curbing 3-strikes reduce sentence lengths?	Event study	CA Sentencing Comm.
Racial equity	Do equity directives reduce B/W disparities?	DiD; IV	CA DOJ by-race; CJARS
Close elections	Does a progressive DA <i>cause</i> lower incarceration?	RDD	Vera ( <a href="#">done</a> )

**6 designs × multiple outcomes = a research agenda.** The policy variation is in hand. Outcome linkage (Phase 1–2) unlocks all of them.

# Deep Dive: Identifying the Causal Effect of Diversion Policies

## The Question

Does adopting a diversion program **reduce recidivism** and **lower case volumes** without increasing public safety risk?

## Identification Strategy

- **Treatment:** date a DA issues an internal diversion directive (coded by our pipeline)
- **Design:** staggered DiD across counties adopting at different times
- **Robustness:** synthetic control for Sacramento, Santa Clara

## Outcome Data Needed

- **CJARS:** individual trajectories—does diversion reduce  $P(\text{re-arrest})$ ?
- **UniCourt:** case outcomes—do filing rates drop? Do plea rates change?
- **CA DOJ:** county-level prosecution counts (free, immediate)

## Why Both

CJARS = **individual mechanism**.

UniCourt = **case-level detail**. Together: “what happens to diverted people *and* to case flow.”

## Extension 1: Scale → Statistical Power for Causal Designs

### Complete California

- Add **17 remaining counties** → full 58-county panel
- **Pre-2015 docs** → 10+ year pre-periods for event study
- **200-doc human validation** → publication-ready IRR

### 100 Largest US DA Offices

- **30–50 close elections** → statistically powered RDD
- Staggered treatment across states → robust DiD
- Cross-state variation → external validity

### What Funding Unlocks

CA alone: ~5 close elections (underpowered RDD). National expansion: **30–50 close elections**—enough for a credible regression discontinuity that can answer “does electing a progressive DA *cause* lower incarceration?”

## Extension 2: Outcome Linkage → True Causal Estimates

### Aggregate → County-Level Causal

Source	Causal Question It Answers
CA DOJ	Do filing rates drop after “decline to prosecute” memos?
Sentencing Comm.	Do enhancement filings fall after reform?
Vera (done)	DiD/event study on jail populations

### Case-Level → Mechanism Evidence

Source	Causal Question It Answers
UniCourt	Do diversion memos change plea bargaining?
CJARS	Does a diversion memo reduce $P(\text{prosecution} \mid \text{arrest})$ ?

### What Funding Unlocks

Without outcome linkage, we have **policy variation without a dependent variable**. Each data source above converts one row of the “6+ causal studies” table from potential into a publishable result.

## Value for Policymakers

- **DA accountability scorecards**—comparable measures of how each office's stated policies align with its goals
- **Rhetoric vs. practice**—does a “decline low-level drugs” memo actually reduce drug prosecutions?
- **Benchmarking tools**—a DA considering bail reform can see how similar policies performed elsewhere
- **Causal evidence on the progressive prosecutor model**—answering the central question: does it work?

### Related Work

Felix Owusu's AV-funded project exploits two specific internal DA memos for a causal design. Our contribution scales this logic: **systematic measurement across 41+ offices and 1,865 documents**, creating variation needed for average causal effects and heterogeneity analysis.

# Potential Funding Package

Phase	Components	Time	Budget	Causal Unlock	Deliverable
1	CA DOJ linkage; complete CA panel; human validation	6 mo	\$45–50k	DiD + synthetic control with validated measurement	First causal estimates of DA policy effects
2	100 US DA offices; national election DB; CJARS app	12–18 mo	\$100–150k	Powered RDD (30–50 elections); robust staggered DiD	National DA ideology database
3	Case-level data (Uni-Court/CJARS)	6–12 mo	\$50–65k	IV + mechanism decomposition	Evidence of <i>how</i> policy changes court-rooms

**Today:** suggestive correlations (34-county pilot) → **Phase 1:** first credible causal estimates → **Phase 3:** full research program.



# Why Fund This Now

## Unique Advantages

- **Infrastructure is built**—pipeline, schema, analysis framework operational
- **Data window is closing**—ACLU PRA archive may not persist
- **No competitor**—no other systematic, AI-coded DA policy database exists
- **Pilot-tested**—demonstrated linkage to outcomes; identified the right methods

## Policy Impact

- Answers the question **funders care about most**: do progressive reforms actually work?
- Enables **evidence-based evaluation** of DA accountability efforts
- Creates **benchmarking tools** for jurisdictions
- First **causal estimates** of prosecutorial policy effects

# Let's Talk

**41 counties · 1,865 documents · 6+ causal designs ready to run**

The policy variation exists. The causal signal is real.  
Fund the outcome linkage, and this becomes a research program.

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[github.com/dyo2112/prosecutor-policies-causal-inference](https://github.com/dyo2112/prosecutor-policies-causal-inference)

Full pilot report, figures, and replication code available in repository

**Additional Detail**

# Overcoming Pilot Limitations

Limitation			Root Cause	Solution
COVID DiD	confounds		All CA counties declined in 2020	Multi-state data; synthetic control
Parallel trends	violated		LA converging pre-Gascón	Longer pre-period + case-level data
TWFE significant	non-		34 counties, 9 years	National expansion increases $N$
Aggregate data coarse	too		Cannot distinguish mechanisms	Case-level data from CJARS

# Causal Identification Strategies

Design	Current Status	What's Needed	Extension
RDD	~5 close CA elections	30–50 close elections	Ext 1 (national)
Stacked DiD	9 disruptions; Vera only	More outcomes + transitions	Ext 1 + 2
Synthetic Control	Gascón pilot works	More outcome data	Ext 2
Event Study	6-year pre-period	Longer pre-period	Ext 1
IV	Margin $\rightarrow$ ideology ( $r = -0.50$ )	Outcome data for 2nd stage	Ext 2

## Outcome Data Sources (Detail)

Source	Records	Timeline	Cost	Priority
CA DOJ OpenJustice	County-yr arrest, filing, incarceration	Immediate	Free	High
CA Sentencing Comm.	Enhancement filings, 3-strikes	1–2 mo	Free	High
Vera (done)	Quarterly jail pop and admissions	Done	Done	Done
UniCourt / PACER	Case dispositions, plea bargains	2–3 mo	\$30–50k	Very high
CJARS	Arrest → parole, 24 states	3–12 mo	\$5k	Highest

**Case-level data is the gold standard:** it links policy documents directly to line-prosecutor behavior.

## Policy Disruption Detection: Methodology

Signal	Weight	Method
Ideology Velocity	30%	Rate of ideology change vs. prior 2-year baseline
Novelty Index	25%	Proportion of first-time policy types
Topic Shift	20%	Jensen-Shannon divergence of topic distributions
Margin Reversal	15%	Flips in extensive/intensive leniency direction
DA Transition	10%	New administration detection

**Top disruptions:** SF 2020 (Boudin, 0.572), LA 2021 (Gascón, 0.549), Sacramento 2022 (0.412)

## Anticipated Objections (1/2)

### “Policy changes aren’t random. Where’s the identification strategy?”

(1) **Close elections are as-good-as-random**—RDD at 50% cutoff; national expansion gives 30–50 races. (2) **Election margin is an IV**— $r = -0.50$  with ideology. (3) **Memo timing is plausibly exogenous**—the exact date a DA issues a specific memo has quasi-random administrative variation; staggered DiD exploits this. (4) **Synthetic control** needs parallel trends, not randomization—LA pilot:  $p_{\text{pre-trend}} = 0.90$ . (5) **Triangulation**: when all four designs agree, combined evidence is strong.

### “Policy documents $\neq$ practice. How do you know memos change anything?”

That’s why Extension 2 is Priority 1. We test whether “diversion-positive” memos predict more diversions. Even the stated-policy variation is *itself* a novel descriptive contribution.



## Anticipated Objections (2/2)

### **“You only have California. Isn’t this just a Blue State story?”**

That’s the point of Extension 1. CA has internal variation (Bay Area vs. Central Valley), but external validity requires the 100 largest DA offices nationally.

### **“AI coding isn’t reliable enough for research.”**

Face validity is strong ( $d = 0.75$  on Gascón, geographic clustering matches expectations). Phase 1 includes 200-doc human validation for gold-standard IRR. Pipeline cost: \$80, runtime: 2 hours.