

## **Voter Analytics Pipeline**

Using Simulated Voter Records | Goodparty.org

## **Agenda**











**Executive Summary** 

Business Problem & Objective

**Technical Approach** 

Orchestration & Lineage









Data Quality & Governance

**Business Impact & Sample Analytics** 

**Surfacing Insights** 

What's Next?
Possible Extensions

# **Production-Ready Voter Analytics Platform**



Scalable Data Ingestion and Transformation to Enable Broad Electoral Analytics

### **Task**

- Build a daily analytics pipeline
- Ingest raw voter records
- Curate analytics datasets for decision-making

**Project Scope**: Take-home assessment

Timeline: 4-6 hours estimated

**Tech Stack**: Airflow | dbt | DuckDB | Streamlit

### **Automated**



2 Python Processors3 Airflow Dags8 dbt Models1 Streamlit Prototype

### Quality





Statically Typed
Unit Tested
dbt Contracted
Auditable

### **Insights**





5 Marts

4 Dims

2 Facts

**Built for Analysis** 

# From Voter Records to Campaign Strategy



Deep Electoral Analytics fueled by a pipeline built for growth

### **Challenges**

Periodic voter file (sub-weekly update)



No guaranteed validation



 Required
 Delivered

 Idempotency
 ✓ Automated EL-processor handling incremental ingestion

 Intermediate
 ✓ Sanitized raw records in contracted DIM and STAGE layers

 Mart
 ✓ Curated tables for core aggregations and targeting

 Best Practice
 ✓ Modular, tested, documented, and reproducible

Integrity risk in input files



No historic election context



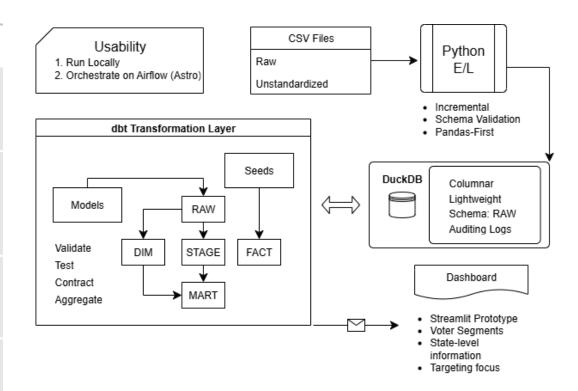
Extensions (Beyond MVP)				
Scalable Build System	3 DAGs   Setup > [Daily Pipeline, Monthly Seed]			
Integrated Election Calendar	MIT Historic   Google Civic API   Federal Schedule			
Behavioral Segmentation	6 engagement tiers, derived opportunity scoring			
<b>Production Patterns</b>	Data contracting, custom macro(s)			
<b>Prototype Dashboard</b>	Streamlit app surfacing 8 interactive visualizations			

### **Technical Approach**

Pandas ELT | Medallion Architecture

Layer	Technology	Rationale
Orchestration	Airflow (Astro) Cosmos	Rapid local development, dbt integration, portable
Ingestion	Python   Pandas	MVP: no-frills basic load strategy  Extension: MD5 deduplication, schema validation, error thresholding, batch processing
Storage	DuckDB	Analytics-optimized, embedded, no infrastructure overhead, good for early phase
Transformation	dbt	✓ MVP: essential dbt validation
Visualization	Streamlit + Plotly	Self-service analytics, no BI tool dependencies, good for prototyping



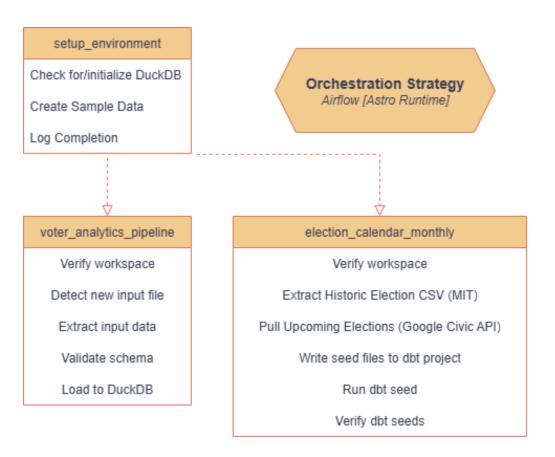


- Pipeline Stats:
- 3 Airflow DAGs
- 15+ dbt models (dim, stage, mart)
- 15+ data quality tests
- 4 production mart tables
- 2 election seed files (historic + upcoming)

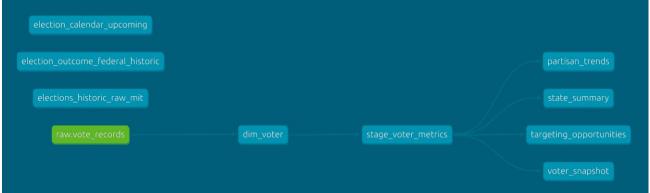
## **Orchestration & Lineage**

**(3)** 

Layered approach for flexibility and reusability



### **dbt Lineage**



Ready to leverage new seeds

- DIM\_VOTER: cleans and standardizes inbound vote\_records upon landing in RAW
- 2. STAGE\_VOTER\_METRICS: prepares voter records for aggregation
- 3. Marts: partisanship, demographics, regional insight-ready

## **Data Quality & Governance**



Production-Grade Testing, Contracts, Type Safety, and Referential Integrity at All Stages

#### **Schema Enforcement**

MVP: Validate key fields

Extension: type-safe EL pattern and

dbt contracts

# Example from voter\_snapshot

contract:

enforced: true

columns:

- name: total\_voters

data\_type: bigint

- name: pct\_current\_voters

data\_type: decimal(5,2)

#### **Quality Assurance**

**MVP**: Tests for errors and warnings

**Extension**: EL unit-testing, dbt column-tests, dbt-expectations

Test Coverage	<b>Covered Domains</b>
DB Config + IO	Storage
Pre-ingest Typing	All inbound fields
Uniqueness	Voter IDs, State Codes
Null-checking	IDs, demographics, marts
Regex Validation	Emails, State Codes
Range Validation	Age (18-120), dates, percentages (0-100)
Accepted Values	Parties, States

#### **ETL Safeguards**

**MVP**: Handle Incremental Loads

**Extensions**:

Quick error check

>5% malformed records

#### Schema Validation

Enforces 10 expected fields

#### Record Deduplication

MD5 Hash:

[ID, First/Last Name, Email]

#### Batch Process

Default 1,000 records

#### Garbage Collection |

Closes connections
Respects DuckDB 1-thread

## **Business Impact & Sample Analytics**



Evolving Raw Records into Data Strategy

#### PROD\_MART.VOTER\_SNAPSHOT

**Purpose**: current voter composition **MVP**: voter count by state, party

**Extensions:** 

Behavioral segments | Engagement tiers

#### 6 Engagement Segments:

- Current Voter (participated recently)
- Missed Last Election (lapsed once)
- Occasional Voter (2-3 lapses)
- Infrequent (4-6 lapses)
- Dormant (7+ lapses)
- Never Voted

#### Sample Insight\*:

"Pennsylvania has 12,500 high-value 'Missed Last Election' target Democrats"

#### PROD\_MART.PARTISAN\_TRENDS

Purpose: time series participation analysis

**MVP**: not required

**Extensions**:

Turnout trends over 9 election cycles (2008-2024)

#### Sample Insight:

"Independent voter participation dropped 18% from 2020 to 2022 midterms suggesting mobilization gap"

#### PROD\_MART. TARGETING\_OPPORTUNITIES

Purpose: ranked segments for GOTV campaigns

**MVP**: not required

**Extensions:** 

Opportunity score algorithm prioritizing recency

#### 

- 40% weight: Recent lapsers (1 election)

- 30% weight: Medium lapsers (2-3)

- 20% weight: Registration tenure

- 10% weight: Segment size

#### Sample Insight:

"Top 20 segments represent 45,000 recoverable voters with 78% recent engagement history"

#### PROD\_MART.STATE\_SUMMARY

**Purpose**: geographic competitive landscape

MVP: voter count by state

**Extensions:** 

Partisan lean classification |

**Engagement Opportunity Scoring** 

#### 📜 Partisan Lean Categories:

- Strong Dem/Rep (>10% margin)
- Lean Dem/Rep (5-10%)
- Competitive (2-5%)
- Highly Competitive (<2%)

#### Sample Insight:

"3 highly competitive states (NC, GA, AZ) have 35% recoverable voter populations"

#### Strategic Impact

- 1. Trend participation rates across cycles and partisanship
- 2. Identify high-priority re-engagement targets among core voter segments
- 3 . Prioritize competitive states for resource allocation
- 4. Segment voters for tailored messaging

<sup>\*</sup> Example insights, not based on the provided sample data ( $n=^1.4k$  unique voters)

# **Surfacing Insights**

Interactive Streamlit dashboard prototype, no SQL required



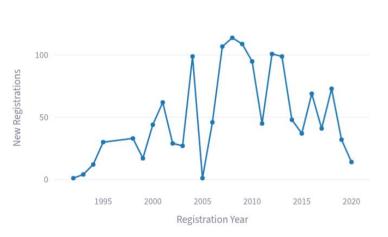
#### **Top Targeting Opportunities**

State	Age Group	Party	Opportunity Score
WA	50-64	Republican	56.7%
OR	30-49	Independent	56.7%
NH	30-49	Democrat	56.7%
AK	30-49	Independent	53.4%
PA	30-49	Republican	53.4%
OR	50-64	Democrat	53.4%
NH	50-64	Republican	53.4%
ИЛ	30-49	Democrat	47.5%
DE	30-49	Democrat	47.5%
CA	30-49	Democrat	47.5%

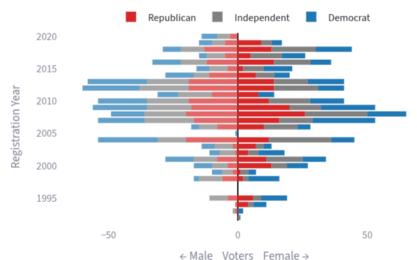
#### Dashboard Features:

- 8 interactive charts (bar, line, heatmap, diverging, pie)
- Real-time filtering (state, party, engagement tier)
- Drill-down from state → demographic segments
- Export-ready tables for campaign teams

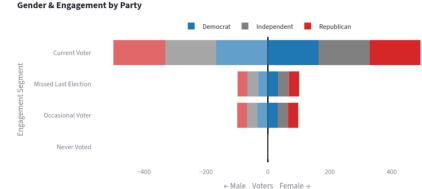
#### **Partisanship by Gender Over Time**



**Voter Registrations by Year** 



#### **©** Targeting by Engagement & Demographics



### What's Next?

Scaling the platform from MVP to enterprise

### Phase 1 [Complete]

- Idempotent ETL pipeline
- Medallion architecture (raw->dim->stage->mart)
- Quality testing and data contracts
- Behavioral segmentation and opportunity scoring
- Interactive dashboard
- Portable distribution











#### **Potential Extensions**

- Phase 2: Production Hardening [2-4 weeks]
  - Refactor/integrate into existing platform (DB, BI, docs)
  - Quality Improvements: Deeper testing, automated Airflow monitoring (Slack/email), audit suite
  - Performance improvements: partitioning, snapshotting (e.g. address changes), query optimizations (clustering, indexes, materialized views)
  - Security & Compliance: PII hashing, RBAC, privacy compliance
- Phase 3: Advanced Analytics [2-3 months]
  - Predictive modeling: turnout prediction, churn risk modeling
  - Enhanced dimensions: household clustering, social listening
  - Real-Time Operations: automate anomaly detection. CDC from upstream systems

## Thank You | Discussion



How does the team balance velocity and technical rigor when making architectural decisions? Framework for evaluating build-vs-buy and 'good enough' vs. production-hardened tradeoffs?

What are the biggest data quality challenges the team is facing today? How do you handle schema evolution in production? How does the team approach net-new data models? (For example, Serve) When something breaks in production, what is the incident response process?

What strategic or operational decisions are hardest to make with data you have today? How do you balance data investments between Win (mature) and Serve (emerging)?

Material	S
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GitHub Repository

PEW Research Party Affiliation

Fact Sheet (NOPRS)

### Socials

Find me on LinkedIn
Check out my GitHub
Read on Medium