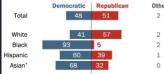
Predicting Electoral Results Using Census and RDH Data

Deven Hagen and Vibby Janardhan

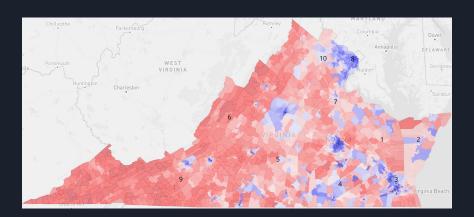
Project Goal

- Electoral precincts individual polling stations
- Predicting electoral results is a huge market
 - o Polymarket, PredictIt
- Using fundamental data rather than polling



Estimates for Asian adults representative of English speakers mily, Low effective sample size for Asian adults in 2022 (99), lottes: Based on 3.014 (2016), 7.585 (2018), 9.686 (2020) and 4.651 (2022) adult citizens for whom reliable data on turnout and ofce choice are evaliable. Turnout was verified using official state election records. Vote choice for all years is from a post-election survey with additional data from panelist profile surveys. White, lack and Asian adults include only those who are not Hispanic; ispanic adults are of any race; source: Survey of U.S. adults conducted Nov. 16-27, 2022, plus

PEW RESEARCH CENTER



Dataset

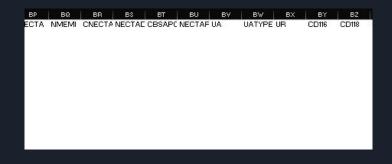
- Source (https://redistrictingdatahub.org/dataset/virginia-block-county-and-vtd-pl-94-171-2020-official)
- Very Large
- Class Transformation
- 463 Attributes and 2465 Instances

STATEFP20	COUNTYFP20	VTDST20	GEOID20	VTDI20	NAME20	NAMELSAD20	LSAD20	MTFCC20	FUNCSTAT20	ALAND20	AWATER20	INTPTLAT20	INTPTLON20	FILEID	STUSAB	SUMLEV	GEOVAR
5	1 149	10	5 51149000105	i A	Harrison	Harrison Voting Dist	tri V2	G5240	N	116568311	529708	37.1616583	-77.1821536	PLST	VA		700 0
5	1 149	20	2 51149000202	2 A	Blackwater	Blackwater Voting I	Di V2	G5240	N	78672067	7529647	37.2398647	-77.1958255	PLST	VA		700 0
5	1 149	20	5 51149000205	A	Jefferson Park	Jefferson Park Votin	ıç V2	G5240	N	30903103	1085514	37.263106	-77.3372307	PLST	VA		700
5	1 149	20	1 51149000201	A	Bland	Bland Voting District	t V2	G5240	N	37721412	4337040	37.2545759	-77.2430783	PLST	V.A		700 0
5	1 149	10	4 51149000104	A	Rives	Rives Voting Distric	t V2	G5240	N	35489092	50496	37.2080484	-77.3084528	PLST	VA		700 0
5	1 149	20	4 51149000204	A	Courts Bldg.	Courts Bldg. Voting	[V2	G5240	N	22526572	7005262	37.2728157	-77.2692636	PLST	VA		700 0
5	1 149	20	3 51149000203	I A	Brandon	Brandon Voting Dis	tr V2	G5240	N	135987153	20361457	37.2540131	-77.0730038	PLST	VA		700 0
5	1 149	10	2 51149000102	2 A	Templeton	Templeton Voting I	D V2	G5240	N	163930478	941327	37.077782	-77.3048884	PLST	V.A		700 0
5	1 149	10	1 51149000101	A	Richard Bland	Richard Bland Votin	n V2	G5240	N	26547216	0	37.1421074	-77.3749943	PLST	V.A		700 0
5	1 149	10	3 51149000103	A	Union Branch	Union Branch Votin	ıç V2	G5240	N	38886837	904747	37.1582155	-77.3103155	PLST	VA		700 0
5	1 595	40	1 51595000401	A	Precinct 4-1	Precinct 4-1		0 G5240	N	4160228	78930	36.708765	-77.5500243	PLST	VA		700 0
5	1 595	20	1 51595000201	A	Precinct 2-1	Precinct 2-1		0 G5240	N	997929	0	36.6992806	-77.5387812	PLST	VA		700 0
5	1 595	50	1 51595000501	A	Precinct 5-1	Precinct 5-1		0 G5240	N	3765223	9104	36.7036103	-77.5213484	PLST	V.A		700 0
5	1 595	30	1 51595000301	A	Precinct 3-1	Precinct 3-1		0 G5240	N	1651252	0	36.689912	-77.5469783	PLST	VA		700 0
5	1 595	70	1 51595000701	A	District 7	District 7		0 G5240	N	864079	0	36.688873	-77.5342971	PLST	V.A		700 0

Data Preprocessing

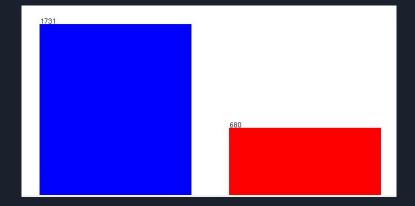
- Attribute Deletion
 - Derived Attributes
- Instance Deletion
- Normalization
 - Election results
 - Racial demographics
 - Population
 - Housing
 - Area
- 32 Attributes and 2411 Instances

1	J
MTFCC20	FUNCSTAT:
G5240	N



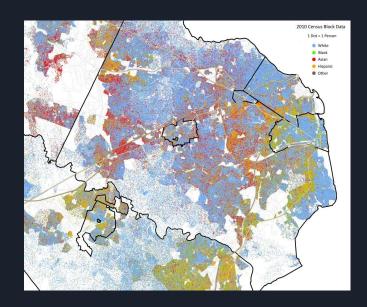
Class Distribution

- Created class labels based on number of voters for each party in the 2020 election
- Removed tied precincts
- Distribution: 71.8% Republican, 28.2% Democratic



Attribute Selection

- 4 Attribute Selection Techniques
 - OneRAttributeEval
 - CorrelationAttributeEval
 - InfoGainAttributeEval
 - ReliefFAttributeEval
- Our chosen attributes
- Similarities
 - Electoral Data
- Differences
 - o Geographic and Race Data



93.3637 26 LTGOV17DPERC 93.1564 24 AG18DPERC 92.7001 25 GOV17DPERC 91.7876 31 PRES12DPERC 91.6217 29 GOV13DPERC 91.4144 28 AG13DPERC 89.5064 30 LTGOV13DPERC 85.9394 5 VANWHTALN 84.6537 14 TNWHALN 79.4276 15 TNBLKALN 77.6856 6 VANBLKALN 77.022 1 AREALAND 76.1095 4 VAHISPANIC OneR 75.4873 13 TAHISPANIC 75.3214 2 AREAWATR 73.8698 17 TNASIANALN 72.5425 8 VANASANALN 71.2982 12 TAPERSONS 71.2153 19 TNOTHRALN 70.7175 22 TAHOCCUPID 69.8465 3 VAPERSONS 68.8096 9 VANNHPOALN 68.6437 7 VANAIANALN 68.3119 18 TNNHPOALN 67.8142 16 TNAIANALN 67.7727 10 VANORALN 67.4824 11 VANM2RACES 67.3165 21 TAHOUSING

66.1551

20 TN2MRACES

Ranked attributes:

27 PRES16DPERC

23 USSEN18DPERC

94.8569

94.7325

Correlation

0.7824	24	AG18DPERC
0.7807	31	PRES12DPEF
0.7802	28	AG13DPERC
0.7797	29	GOV13DPERC
0.7765	25	GOV17DPERC
0.7758	23	USSEN18DPE
0.7473	30	LTGOV13DPE
0.6876	5	VANWHTALN
0.6648	14	TNWHALN
0.5316	15	TNBLKALN
0.4888	6	VANBLKALN
0.4346	4	VAHISPANIO
0.4191	13	TAHISPANIO
0.3609	1	AREALAND
0.3413	17	TNASIANALN
0.3145	19	TNOTHRALN
0.3082	8	VANASANALN
0.3075	3	VAPERSONS
0.3009	12	TAPERSONS
0.2394	20	TN2MRACES
0.2107	10	VANORALN
0.1813	11	VANM2RACES
0.1606	18	TNNHPOALN
0.1324	9	VANNHPOALN
0.0742	21	TAHOUSING
0.0579	2	AREAWATR
0.0387	7	VANAIANALN
0.0287	22	TAHOCCUPII
0.0174	16	TNAIANALN

Ranked attributes:

27 PRES16DPERC

26 LTGOV17DPERC

0.7888

0.7865

0.6722 25 GOV17DPERC 0.6669 24 AG18DPERC 0.6639 26 LTGOV17DPERC 0.6253 31 PRES12DPERC 0.6228 28 AG13DPERC 0.6164 29 GOV13DPERC 0.5746 30 LTGOV13DPERC 0.4144 5 VANWHTALN 0.3728 14 TNWHALN 0.2569 1 AREALAND 0.2382 15 TNBLKALN 0.2034 6 VANBLKALN 0.1572 4 VAHISPANIC InfoGain 0.1487 2 AREAWATR 0.1407 13 TAHISPANIC 0.1061 17 TNASIANALN 0.101 8 VANASANALN 0.0831 19 TNOTHRALN 0.0673 3 VAPERSONS 0.0639 12 TAPERSONS 0.0427 10 VANORALN 0.0384 20 TN2MRACES 0.038 22 TAHOCCUPID 0.0259 18 TNNHPOALN 0.0237 9 VANNHPOALN 0.0226 11 VANM2RACES 21 TAHOUSING 7 VANAIANALN 16 TNAIANALN

0.6916

0.6796

23 USSEN18DPERC

27 PRES16DPERC

0.183269 27 PRES16DPERC 0.180663 26 LTGOV17DPERC 0.178971 25 GOV17DPERC 0.175165 24 AG18DPERC 0.123077 28 AG13DPERC 0.113753 29 GOV13DPERC 0.110577 30 LTGOV13DPERC 0.10999 31 PRES12DPERC 0.082153 5 VANWHTALN 0.075585 14 TNWHALN 0.061844 15 TNBLKALN 0.056405 6 VANBLKALN 0.04872 1 AREALAND 0.041586 20 TN2MRACES ReliefF 0.038933 17 TNASIANALN 0.037806 22 TAHOCCUPID 0.036286 8 VANASANALN 0.027004 11 VANM2RACES 0.02451 13 TAHISPANIC 0.024192 4 VAHISPANIC 0.020693 12 TAPERSONS 0.01809 3 VAPERSONS 0.014764 10 VANORALN 0.014508 9 VANNHPOALN 0.012837 21 TAHOUSING 0.009908 18 TNNHPOALN 0.009466 19 TNOTHRALN 0.000832 16 TNAIANALN 0.00083 7 VANAIANALN -0.0001772 AREAWATR

0.206555

23 USSEN18DPERC

Our Chosen Attributes

- Combination of each selection method
 - USSEN18DPERC, AG18DPERC, GOV17DPERC, LTGOV17DPERC, PRES16DPERC, AG13DPERC, GOV13DPERC, LTGOV13DPERC, PRES12DPERC, VANBLKALN, AREALAND

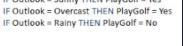
AREALAND	VANWHTALN	VANBLKALN	USSEN18DPERC	AG18DPERC	GOV17DPERC	LTGOV17DPERC	PRES16DPERC	AG13DPERC	GOV13DPERC	LTGOV13DPERC	PRES12DPERC	WIN_PARTY
0.116568	0.760894	0.139665	0.239468	0.219072	0.21883	0.205656	0.237258	0.316568	0.248521	0.393491	0.299484	R
0.078672	0.707191	0.202622	0.342271	0.333975	0.329828	0.317919	0.293536	0.330208	0.300306	0.420063	0.339479	R
0.037721	0.668154	0.210554	0.390798	0.351495	0.359974	0.350365	0.341407	0.350951	0.309257	0.423186	0.352407	R
0.035489	0.419699	0.298236	0.578216	0.573723	0.578871	0.574344	0.538462	0.573333	0.539474	0.623529	0.551509	R
0.022527	0.48436	0.381043	0.549313	0.514329	0.522846	0.515813	0.501065	0.510061	0.484321	0.570423	0.528642	R
0.135987	0.768549	0.169435	0.315488	0.314031	0.323661	0.329621	0.299509	0.335784	0.305825	0.424691	0.326367	R
0.16393	0.751823	0.164731	0.262866	0.245961	0.255344	0.244165	0.238056	0.31178	0.287724	0.37669	0.301431	R
0.026547	0.705838	0.203593	0.398542	0.34104	0.333333	0.325145	0.317497	0.367107	0.340708	0.474852	0.354031	R
0.038887	0.626954	0.282636	0.44164	0.430037	0.42844	0.428304	0.405617	0.474006	0.457372	0.541117	0.479739	R
0.00416	0.41226	0.496394	0.535032	0.43662	0.437931	0.43554	0.531317	0.506452	0.52322	0.602649	0.530612	R
0.001651	0.402821	0.492163	0.45679	0.440678	0.413502	0.405983	0.445714	0.395161	0.416667	0.585062	0.470297	R
0.001239	0.530928	0.360825	0.450355	0.381107	0.362745	0.388889	0.409091	0.397924	0.405405	0.566434	0.42446	R
0.002935	0.581509	0.043777	0.656716	0.596072	0.598131	0.605646	0.563649	0.555246	0.529086	0.600561	0.532468	R
0.00228	0.563664	0.032687	0.700523	0.668194	0.671208	0.66819	0.615273	0.601385	0.558824	0.626131	0.57537	R
0.002018	0.638917	0.028426	0.682975	0.651079	0.646515	0.650179	0.610406	0.602904	0.569257	0.63503	0.571257	R
0.004435	0.559259	0.044907	0.680664	0.633929	0.641905	0.636999	0.598681	0.587189	0.568614	0.632184	0.565117	R
0.082556	0.908634	0.015088	0.297041	0.298438	0.288596	0.277821	0.229406	0.303872	0.297678	0.352743	0.325872	R

Classifiers

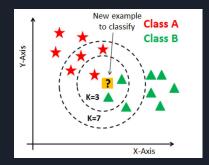
- J48 decision tree
- Naïve Bayes
- OneR Classifier
- K-nearest neighbor

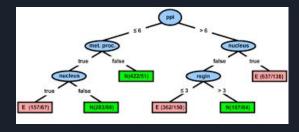


IF Outlook = Sunny THEN PlayGolf = Yes









Accuracy Table

	OneR Attribute Eval	Correlation	InfoGain	ReliefF	Our Chosen Attributes
J48	0.954791	0.944836	0.951058	0.942762	0.944836
OneR Classifier	0.948569	0.948569	0.948569	0.948569	0.948569
K-nearest neighbor	0.941518	0.941103	0.933637	0.941103	0.9382
Naïve Bayes	0.92866	0.929905	0.92949	0.934467	0.92949

Results

- K-fold value = 10
- OneR Classifier had identical results
- Accuracy better metric than False
 Positives and Negatives
- Consistent 92-95% values

Best Model: OneRAttribute + J48 Decision Tree Classifier

Correctly Classi	fied Inst	ances	2302		95.4791	8			
Incorrectly Clas	sified In	stances	109		4.5209	8			
Kappa statistic			0.88	75					
Mean absolute er	ror		0.05	89					
Root mean square	d error		0.19	51					
Relative absolute	e error		14.53	86 %					
Root relative sq	uared err	or	43.34	86 %					
Total Number of	Instances		2411						
=== Detailed Accuracy By Class === TP Rate FP Rate				Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.973	0.093	0.964	0.973	0.969	0.888	0.961	0.966	R
	0.907	0.027	0.931	0.907	0.919	0.888	0.961	0.905	D
Weighted Avg.	0.955	0.074	0.955	0.955	0.955	0.888	0.961	0.949	
=== Confusion Ma	trix ===								
a b <	classifi	ed as							
1685 46	a = R								
63 617	b = D								

Conclusion

- Previous elections, racial data are good predictors
- No difference between Voting Age vs Total Population
- Increasing geographic polarization

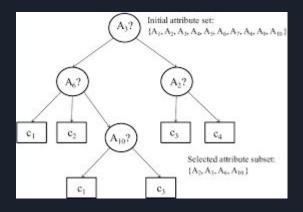


Future Research

- More Attribute Selection Algorithms/Classifiers
- Different K-Fold values
- Expand to other states
 - Use counties instead of precincts?







Thank you!

Any Questions?