Civil rights school-related data

Find your dedicated page and share your analysis process on your chunk of data. Include the following:

Two-three inquiry questions you generated during a survey of your data A screen shot or two of your pivot table that sheds light on your questions A screen shot or two of your produced charts A few samples of your formulas used in your analysis A bulleted list of conclusions you can (or cannot) draw from the data Doubts or further questions about the data itself What would you do next if you were developing this into a project? Two-three inquiry questions you generated during a survey of your data

Is there a relationship between frequently of vandalism and frequency of a student equipping a knife when comparing counties?

What are the summary statistics for several violations (Alcohol or drug use, Knife Equipped, Vandalism, Fighting) divided by the number of students enrolled?

A screen shot or two of your pivot table that sheds light on your questions

3		Sum of Alchohol and Drugs Divided by Enrollment	Sum of Knife Equipped Divided by Enrollment	Sum of Vandalism Divided by Enrollment	Sum of Fighting Divided by Enrollment
4	Adams	0.19%	0.10%	0.08%	0.21%
5	Allegheny	0.26%	0.07%	0.09%	0.99%
6	Armstrong	0.25%	0.12%	0.00%	0.04%
7	Beaver	0.29%	0.06%	0.05%	0.33%
8	Bedford	0.11%	0.06%	0.04%	0.26%
9	Berks	0.19%	0.10%	0.08%	0.53%
10	Blair	0.15%	0.09%	0.03%	0.25%
11	Bradford	0.19%	0.16%	0.04%	0.51%
12	Bucks	0.13%	0.03%	0.08%	0.20%
13	Butler	0.18%	0.11%	0.04%	0.15%
14	Cambria	0.10%	0.04%	0.06%	0.13%
15	Cameron	0.16%	0.00%	0.00%	0.48%
16	Carbon	0.07%	0.05%	0.05%	0.37%
17	Centre	0.14%	0.05%	0.07%	0.28%
18	Chester	0.20%	0.03%	0.04%	0.36%
19	Clarion	0.12%	0.09%	0.31%	0.14%
20	Clearfield	0.28%	0.09%	0.05%	0.57%
21	Clinton	0.38%	0.17%	0.05%	0.45%
22	Columbia	0.33%	0.08%	0.05%	0.26%
23	Crawford	0.23%	0.09%	0.03%	0.30%
24	Cumberland	0.18%	0.07%	0.03%	0.18%
25	Dauphin	0.18%	0.06%	0.04%	1.13%
26	Delaware	0.16%	0.07%	0.07%	0.98%
27	Elk	0.08%	0.14%	0.00%	0.03%
28	Erie	0.28%	0.11%	0.07%	1.34%
29	Fayette	0.23%	0.08%	0.04%	0.90%
30	Forest	0.19%	0.00%	0.00%	0.19%
31	Franklin	0.20%	0.11%	0.06%	0.18%
32	Fulton	0.05%	0.05%	0.28%	0.65%
33	Greene	0.27%	0.08%	0.06%	0.39%
34	Huntingdon	0.18%	0.09%	0.05%	0.51%
35	Indiana	0.21%	0.04%	0.02%	0.16%
36	Jefferson	0.17%	0.06%	0.06%	0.04%
72	Mean	0.17%	0.09%	0.07%	0.39%
73	Standard Error	9.77442E-05	6.58387E-05	8.48676E-05	0.000347251
74	Median	0.18%	0.08%	0.05%	0.33%
75	Mode	0.00%	0.00%	0.00%	#N/A
76	Standard Deviatio	0.000800071	0.000538913	0.000694671	0.002842371
77	Sample Variance	6.40114E-07	2.90427E-07	4.82568E-07	8.07907E-06
78	Kurtosis	0.322789668	5.417515065	7.598055289	1.454654232
79	Skewness	0.076520194	1.716041422	2.592170853	1.285177472
80	Range	0.38%	0.33%	0.37%	1.34%
81	Minimum	0.00%	0.00%	0.00%	0.00%
82	Maximum	0.38%	0.33%	0.37%	1.34%
83	Sum	11.56%	5.90%	4.46%	25.99%
84	Count	67	67	67	67

A screen shot or two of your produced charts



A bulleted list of conclusions you can (or cannot) draw from the data There is no relationship between frequently of vandalism and frequency of a student equipping a knife

1. Two-three inquiry questions you generated during a survey of your data

1) What are the total number of sexual harassments per enrollment and the total number of incidents per enrollment for schools in the Allegheny County, respectively?

2) Which pre-selected counties have the highest and the lowest probability of bullying, fighting, and possession of Knife, respectively?

2. A screen shot or two of your pivot table that sheds light on your questions

1) What are the total number of sexual harassments per enrollment and the total number of incidents per enrollment for schools in the Allegheny County, respectively?

1.1	A	В	C	D	E					
1	County	Allegheny 🖵			PivotTable Field List 🛛 💌 🗙					
2										
3	Row Labels	Sum of Enrollment	Sum of SexualHarassmentPerEnrollement	Sum of IncidentsPerEnrollment	Choose fields to add to report:					
4	A W Beattie Career Center	655	0	0.019847328	School Name					
5	Abraham Lincoln El Sch	328	0	0.009146341	County					
6	Acadamy at Westinghouse	489	0	0.253578732	Enrollment					
7	Academy CS	188	0.010638298	0.14893617	Incidents					
8	Acmetonia Primary Sch	301	0	0	✓ IncidentsPerEnrollment					
9	Allard El Sch	187	0	0						
10	Allegheny IU 3	504	0	0.047619048						
11	Avalon El Sch	325	0	0.027692308						
12	Avonworth El Sch	779	0	0.002567394						
13	Avonworth HS	417	0.002398082	0.033573141						
14	Avonworth MS	345	0	0.002898551	Drag fields between areas below:					
15	Baker El Sch	397	0	0	Y Report Filter Column Labels					
16	Baldwin SHS	1470	0.000680272	0.019727891	County					
17	Barrett El Sch	290	0	0.106896552						
18	Bellevue El Sch	363	0	0.03030303						
19	Benjamin Franklin El Sch	338	0	0.00887574	Row Labels Σ Values					
20	Bethel Memorial El Sch	341	0	0	School Name Sum of Enroll					
21	Bethel Park HS	1561	0	0.062780269	Sum of Sexua 🔻					
22	Bon Meade El Sch	454	0	0.002202643	Sum of Incide 🔻					
14 -	▶ ► 2013-2014 / WORKING DATA	pivot table / Shee	19 / 12 /		Defer Lavout Update					
Der	ala a									
284	Woodland Hills Academy	483	0	0.00621118						
285	Woodland Hills JHS	485	0	0.094845361						
286	Woodland Hills SHS	1191	0	0.034424853						
287	Wyland El Sch	405	0	0						
288	Young Scholars of Western Pennsylva	230	0	0.004347826						
289	Grand Total	152465	0.228525473	11.65488964						

2) Which pre-selected counties have the highest and the lowest probability of bullying, fighting, and possession of Knife, respectively?

3	Row Labels	J Sum of Enrollment	Sum of BullingPerEnrollment	Sum of FightingPerEnrollment	Sum of PossOfKnifePerEnrollment	PivotTable	Field List		• ×	
4	Allegheny	142783	0.00107856	0.009889132	0.000714371				Carl I	
5	Beaver	22136	0.002258764	0.003342971	0.000632454	Choose fie	lds to add to re	eport:		
6	Clinton	4223	0.000473597	0.004499171	0.001657589				^	
7	Dauphin	34145	0.001142188	0.011334017	0.00055645	Bulling	PerEnrollme	ent		
8	Elk	3642	0.000549149	0.000274574	0.001372872		ParEnrolln	oont		
9	Franklin	19410	0.000206079	0.001803194	0.001081917		sion of a Knife	iciic		
10	Greene	5188	0.005975328	0.00385505	0.00077101	Posso	fKnifePerEnr	ollment	~	
11	Huntingdon	5514	0.003445774	0.005077983	0.000906783			0.00.00.00.00		
12	Indiana	9833	0.002339062	0.001627174	0.000406793	Drag fields	between area	s below:		
13	Jefferson	4797	0.003752345	0.000416927	0.000625391	Y Repo	rt Filter	Column	Labels	
14	Lebanon	19026	0.000420477	0.004730369	0.000578156			Σ Values	-	
15	Mifflin	5273	0.000379291	0.002655035	0.000379291					
16	Northumberlan	d 12007	0.001165987	0.003831099	0.000582993					
17	Philadelphia	137674	0.000239697	0.004263695	0.00166335					
18	Snyder	4889	0.003272653	0.001227245	0.001022704	Row I	abels	Σ Values		
19	Tioga	5429	0.001289372	0.004973292	0.000184196	County	County 👻		Sum of Enrollm 🔻	
20	Union	3978	0.001256913	0.001256913	0.001759678			Sum of Bull	ing 🔻	
21	Venango	8239	0.003277097	0.006068698	0.000242748			Sum of Figh	ntin 🔻	
22	Washington	27934	0.001288752	0.003257679	0.000894967			Sum of Pos	sof ▼	
23	York	64633	0.001438893	0.007178995	0.00125323	Defer Lavout Lindate			Update	
24	Grand Total	540753	0.035249977	0.081563213	0.017286943					
25										
26	max of bulling p	per enrollment	0.005975328							
27	min of bulling p	er enrollment	0.000206079							
28										
29	max of fighting	per enrollment	0.011334017							
30	min of fighting	per enrollment	0.000274574							
31										
32	max of knife po	ssession per enrollme	0.001759678							
33 min of knife possession per enrollmer			0.000184196							

3. A screen shot or two of your produced charts

X 🖌 🤊 - (° -		Bicture Tools schoolCivilRightsDataSetFromCRDC2013-2014.xlsx - Microsoft Excel																			
File Home	Insert Page Lay	out Formulas	Data Re	eview View	Format																
			ah ,	My 🍋 🗏	• 📣 🗠	0		0.0			A		A	78 .	AS)	π	0				
PivotTable Table	Picture Clip Shape	s SmartArt Screens	hot Column	Line Pie Ba	Area Scatter	Other	Line Column	n Win/Loss	Slicer	Hyperlink	Text	Header	WordArt S	ignature C	bject	Equation 1	Symbol				
Tables	Art 👻	• ations	-	Chart	* *	Charts *	Sparklij	nes	Filter	Links	Box *	& Footer	Text	Line *		* Symbo	ols				
Picture 1	- (=	fx														-,					
		2000 II		8					2	8		S2.1 :	227		9			257			
A 1	В	C	D	E	F G	ŀ	1 1	J	K	(L		M	N	0	P		Q	R	S	T	U
1																					
3 Row Labels	T Sum of Enrolling	Sum of Bullingt	um of Fightin	Sum of PossOf	nifePerEnrollm	ont						9									
4 Alloghony	1/12792	0.00107856	0.000990122	0.000714271		ent															
5 Reaver	22126	0.00107850	0.003883132	0.000714371	Sum of BullingPer	Enrollment	Sum of FightingP	erEnrolment	Sum of P	ossOfKnifePe	rEnrollme	nt									1
5 Clinton	4222	0.002238704	0.003342371	0.001657589	Chart Are	a	Conterrigitorigi	CTER CHITCH	June	obsont inter e	- Ers onne										
7 Dauphin	34145	0.0004733357	0.011224017	0.00055645	0.012																
8 Elk	3642	0.000549149	0.000274574	0.001372872	0.011												Values)-	
9 Franklin	19/10	0.000206079	0.001803194	0.001081917	2222												Sun	n of Bullin	PerEnroll	ment	
10 Greene	5188	0.005975328	0.00385505	0.00077101																	
11 Huntingdon	5514	0.002445774	0.005077992	0.000077101	0.009																
12 Indiana	9922	0.003445774	0.001627174	0.000300783	Sum of PossofKnifePerEnrollment																
12 Inforcon	7055	0.002339002	0.001027174	0.000400793	0.008																
14 Johanon	19026	0.003732343	0.000410327	0.000023331	a 0.007																
15 Aufflig	19020	0.000420477	0.004730303	0.000378130																	
15 Northumbork	3273	0.000379291	0.002033033	0.000579291	0.006					÷											
17 Dhiladalahia	127674	0.001103587	0.003831035	0.000382333	0.005					_											
19 Spydor	137074	0.000233057	0.004203035	0.00100333	0.005																0
18 Silydei	4003	0.003272033	0.001227243	0.001022704	0.004							-		9	-		-		-		
20 Union	3423	0.001289372	0.004373232	0.000134130	0.002														8 <mark>- 8</mark> - 1		
20 Onion	9320	0.001230913	0.001230313	0.001739078	0.003																
22 Washington	27924	0.003277037	0.000003038	0.000242748	0.002				-			-			-			-			
22 Vork	64622	0.001288752	0.003237075	0.000834307	0.001																155
24 Grand Total	540753	0.0352/0077	0.081563213	0.017286043	0.001										-						
25	540755	0.033243377	0.001303213	0.017200343	0						,		,						-		
26 max of bulling	ner enrollment	0.005975328			Pris	Net	tor min	Elt N	in on	e don	and	Son	non i	air ar	n an	a det	6800	non	080	the non	
27 min of bulling	ner enrollment	0.000206079			Meg	see Ch	Dank	6131.	Gre	ating	mo.	etter le	0. 41	aber	adely	Ser	~	S. 7.	eno shin	0	
28	sperententent	0.000200073			r					40.		MI 22		Athur	6km				20		
29 may of fightin	g per enrollment	0.011334017											4	0.							
30 min of fightin	a per enrollment	0.000274574			County 🖓																
31	is per en onnent	5.000274074		0				-				-0-									-0
32 max of knife	nossession ner enr	0.001759678																			
33 min of knife	ossession per enr	0.000184196																			
34	reading and a second se	0.000104130																			

4. A few samples of your formulas used in your analysis

1) probability of bullying (bullying per enrollment) = number of bullying / number of enrollment

2) highest probability of fighting = MAX(D4:D23)

3) lowest probability of knife possession = MIN(E4:E23)

5. A bulleted list of conclusions you can (or cannot) draw from the data

1) The total number of sexual harassments per enrollment for schools in the Allegheny County is about 0.2285; the total number of incidents per enrollment is about 11.65;

2) Greene County has the highest probability of bullying; Dauphin County has the highest probability of fighting; Union County has the highest probability of knife possession.

3) Franklin County has the lowest probability of bullying; Elk County has the lowest probability of fighting; Tioga County has the lowest probability of knife possession.

6. Doubts or further questions about the data itself

1) Why are some of the columns left blank? For example, columns named "Assaults on Students", "Sexual Offenses", "Possession of a Firearm", etc.

2) Why does the column "Truancy Rate" shows percentages instead of numbers? How are the percentages calculated?

7. What would you do next if you were developing this into a project?

1) check the quality of the data. For example, does the spreadsheet mix data, calculation, and reporting?

Jeff P.

Find your dedicated page and share your analysis process on your chunk of data. Include the following:

- 1. Two-three inquiry questions you generated during a survey of your data
 - a. I explored the relationship between the incidence of (reported) bullying and the incidence of a several other classifications of infractions: theft, possession of a knife, possession of tobacco, possession of controlled substances, and disorderly conduct. The comparisons of these incidences were made in terms of a calculated column that divided the total number of each incident for each school district by that school district's total enrollment. I wanted to know if there were any readily identifiable anomalies between the incidence of bullying and these other categories, across the ENTIRE data set. That is, I averaged all districts' enrollment-adjusted metrics for theft, bullying, tobacco possession (etc) and compared the summary statistics of those aggregate averages. I highlighted the bullying column in both the data set and the summary statistics, in green.
 - b. I also wanted to know if there were any noticeable correlations between any of the variables I was tracking, for which I used the CORREL() function in a small table, colored in pink.

	A	В	С	D	E	F	G	Н	I.	J	К	L		
1	LEA Type	District 🖵												
2					-						Bullying's			
3	Row Labels	Sum of Incidents/Enrc Su	um of Theft/Er Sum	n of Enrol Sum	of Bullying/Enn Su	um of Disorderly	Sum of Possession Tobacco/En	ession Tobacco/Enrc Sum of Possession of Sum of Knife		rollm	CORREL	With		
4	Abington Heights SD 👻	0.302480339	0	3306	0	0	0.03024803	4	0 0.030248034		-0.02001	Enrollment	Si	
5	Abington SD	0.287995811	0	7639	0.013090719	0	0.01309071	9 0.15708862	0.026181437		0.529081	Incidents/E	nr	
6	Albert Gallatin Area SD	2.499289974	0.056802045	3521	0.056802045	0.312411247	0.08520306	7 0.31241124	17 0.028401022		0.102172	Disorderly	Co	
7	Aliquippa SD	1.282051282	0	1170	0	0	(0.68376068	60684 0		0.058402	Drugs		
8	Allegheny Valley SD	1.210898083	0	991	0.100908174	0.100908174	0.90817356	2	0 0		0.08076	Tobacco		
9	Allegheny-Clarion Valley SD	3.698435277	0.426742532	703	0.142247511	0.142247511	0.71123755	3	0 0		0.037945	Knives		
10	Allentown City SD	18.71104316	0.687992473	17006	1.522991885	1.664118546	0.288133	0.2881336 0.111725273			0.239202	Theft		
11	Altoona Area SD	2.26342711	0.10230179	7820	0.025575448	0.051150895	0.33248081	0.0895140	6 0.179028133					
	4												₽	
£														
Jx	-VAR.P(H4:H502)													
	А	В	С	D	E	F	G		н			1		
305	Grand Total	1303.04383	03.1/300242	1003292	07.5124	5051 152.0	5/5/42/	173.834028	/1.	2705	2740 J	7.3012351	2	
504	Per Enrollment Categories>	 Incid/Enrol 	l Theft	(Enrollmnt)	Bul	llying Di	sorderly	Tobacco	Controlled	led Substance		Knit	e	
505	MEAN	2.611314328	0.126600967	3217.01804	0.17617	7351 0.26	5629207	0.352372802	372802 0.1428		3542 0	.07531314	5	
506	MEDIAN	1.843075303	0.057234432	2110	0.024	1955 0.032	2310178	0.22997317		0934	9371 0	71 0.050994391		

0

6.36877247

0.502081914

0.50258576

0.252086248

0

18.19085487

1.177069911

1.178251116

1.385493576

0

3.237410072

0.397855705

0.398254958

0.158289162

0

0.150891758 0.085674473

0.15104318 0.085760449

0.940228341

0.022768323

0

0.532386868

0.007340115

2. A screen shot or two of your pivot table that sheds light on your questions

3. A screen shot or two of your produced charts

0

27.00170358 2.221193892

3.023306611 0.222758213 6626.1007

3.026340537 0.222981754 6632.75008

9.140382862 0.049621221 43905210.5

0

214

137674

507 MIN

508 MAX

509 STDEV.P

510 STDEV.S

511 VAR.P



(on the left: average incidence by category. On the right: stDev by category.)

4. A few samples of your formulas used in your analysis
=STDEV.P(C4:C502)
=MEAN(G4:G502)
=CORREL(E4:E502,B4:B502)

- 5. A bulleted list of conclusions you can (or cannot) draw from the data
 - Just how DOES one define "disorderly conduct?" No one appears to know. Note its aberrantly high standard deviation. But (judging by its stDev, bullying's definition seems to be at least as well understood as that of "possession of tobacco," and is as close to being universally understood as possession of drugs or knives or stealing.
 - A look at correlations of Bullying:
 - CORREL With...
 - -0.02001 Enrollment Size
 - 0.529081 Incidents/Enrollment
 - 0.102172 Disorderly Conduct/Enrollment
 - 0.058402 Drugs
 - 0.08076 Tobacco
 - 0.037945 Knives
 - 0.239202 Theft

-

- There does not appear to be a correlation between enrollment and bullying's incidence. The positive correlation with Incidents/Enrollment is understandable, since bullying is a subset of this. There is a slightly positive correlation with Disorderly Conduct, and a slightly stronger correlation with Theft. I have no thoughts on the latter correlation, other than curiosity.
- I set out to sniff for inconsistencies in the definition of what bullying is. My choice of theft, knives, drugs, etc were deliberate inasmuch as I thought they were more apt to be clearly articulated offenses (several of them depend on the offender actually possessing something, rather than doing something). My choice to include disorderly conduct was slightly spurious, because I felt it to be similar to bullying inasmuch as both are subjective. I included it to see if their stDev's across the data set were both high. They are.
- I cannot say from this data that no one knows what bullying is (despite my earlier rhetoric). But I can assert that its definition seems less discrete than many other categories.

- 6. Doubts or further questions about the data itself
 - My project was based upon the doubt about the understanding of a definition (bullying). It is possible that other categories of offenses are diversely understood amongst these many Pennsylvania schools, as well.
 - Is smoking or possessing tobacco an offense at all schools? Or most?
 - Does the definition of "disorderly conduct" resemble that of law enforcement?
- 7. What would you do next if you were developing this into a project?
 - I'd look for more correlations and include more categories of incidents. That is, I'd broaden my scope of control variables beyond theft, disorderly, drugs, knives, etc.