WEB SCRAPING FOR THE CODING-CHALLENGED

FDLP WEBINAR

May 23rd, 2019



Carl p. Olson, Librarian III, Coordinator for government publications Albert S. Cook Library, Towson University





Today's Agenda







- What is Data-Scraping?
- What Does One Do with Spreadsheet Data?
- Hard and Easy ways to scrape data;
- Data-Scraping HTML with Google Sheets;
- Further Information.









Web-scraping is a (typically automated) process which transfers content from online documents to an interactive format, such as Excel or CSV, for analysis, aggregation, or further computation.

What is Data Scraping

- □ Web-scraping is as old as the Web itself;
- □ Web-scraping: "content-harvesting lite."
- Now used by business analysts, journalists, and researchers;
- Ccoding-challenged professionals on a deadline.





Why Do People Scrape Data? TU TOWS

- Directories;
- Employment listings;
- Products & pricing;
- Web addresses;
- Site maps;
- Annual reports;
- Data tables from reports;
- Documents!



Many Federal Sites offer data in spreadsheets:

The Easiest Data-Scraping:

CENSUS oF AGRICULTURE DATA RELEASE VIDEO

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□ BLS; □ BEA; □ FRED; **G** FBI; DOA; \Box CENSUS; \Box CDC; \Box NCES; **BTS**;

Criminal Justice Information Services (CJIS)



he FBI's Criminal Justice Information Services Division, or CJIS, is a high-tech hub in the hills of West Virginia at provides a range of state of the-art tools and services to law enforcement, national security and intelligence ommunity partness, and the general public.

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Why Do People Scrape Data?



FBI, Crime in the U.S., 2017;Murders in the U.S.;By State;By Type of Weapon



Criminal Justice Information Services Division												
Table		n to Law Enfor	<u>cement</u>	<u>olent Crime</u>	<u>Property Cr</u>	ime <u>Clearan</u>	ces <u>Persons</u>	<u>Arrested</u> Poli	<u>ce Employee Dat</u>	<u>a</u>		
1	, Types of We											
State		Total murders ¹	Total firearms	Handguns	Rifles	Shotguns	Firearms (type unknown)	Knives or cutting instruments	Other weapons	Hands, fists, feet, etc. ²		
Alabama ³		2	1	0	0	0	1	0	1			
Alaska		62	37	7	3	3	24	13	8			
Arizona		404	249	162	8	9	70	50	93			
Arkansas		250	168	92	11	4	61	23	52			
California		1,830	1,274	886	37	34	317	258	195	1		
Colorado		218	137	88	7	4	38	37	22			
Connecticu	ıt	102	72	30	0	1	41	11	9			
Delaware		52	44	20	0	1	23	3	4			
District of C	Columbia	116	90	89	0	0	1	15	5			
Georgia		672	542	490	15	5	32	37	85	8		
Hawaii		39	4	1	1	0	2	9	10	16		
Idaho		28	13	8	4	1	0	6	3			

Why Do People Scrape Data?

"Interviewing" data:

- Autosum
- Transpose;
- Ranking;
- Ratios.



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Table 20										
Murder										
by State, Types o	of Weapons.	2017								
						Firearms	Knives or			
	Total	Total				(type	cutting	Other	Hands, fists,	
State	murders ¹	firearms	Handguns	Rifles	Shotguns	unknown)	instruments	weapons	feet, etc. ²	
Alabama ³	2	1	0	0	0	1	0	1		5
Alaska	62	37	7	3	3	24	13	8	4	161
Alaska Arizona	404	249	162	8	3	24 70	50	93	4	1.057
Arkansas	250	243	92	11	4	61	23	52	7	668
	1,830	1,274	886	37	34	317	258	195	103	4,934
California Colorado	218	1,274	000	37 7		38	250	22	22	4,534
Colorado	102	72	30	0	4	41		22	10	276
Delaware	52	44	20	0	1	23	3	4	1	148
Delaware District of Columbia	116	90	89	0	0	23	15	5	6	322
Georgia	672	542	490	15	5	32	37	85	8	1,886
Hawaii	39	4	430	1	0	2	9	10	16	82
nawaii Idaho	28	13	8	4	1	0	6	3	6	69
Illinois ³	814	693	596	24	3	70	53	50	18	2,321
Indiana	360	291	147	14	6	124	20	39	10	1,011
lowa	100	57	25	1	5	26	18 16	18	7	257
Kansas	129 263	79	44	4	7	24 52	16	26	8	337
Kentucky	263	192 460	128 216	5 23		209	25	33	13	718
Louisiana Maine	23	460	216	23	12	209	46	42	18	1,592
	475	370	339	5	3	23	44	50	4	1,320
Maryland	475	370	335	0	0	23	36	29	6	439
Massachusetts	567	381	185	13	12	171	55	23	30	
Michigan Minnesota	113	69	58	1	2	8	55	23		1,515
	149	111	90	4	3	14	14	23	r 6	409
Mississippi Missouri	596	514	224	22	8	260	25	48	9	1,706
Montana	41	17	224	22	1	200	12	40	7	99
Nebraska	41	31	27	2	2	0	4	5	3	117
Nevada	270	201	16	58	0	127	28	30	11	741
New Hampshire	14	201	4	0	1	2	5	1	1	35
New Jersey	324	242	175	7	4	56	42	29	11	890
New Mexico	113	71	20	2	Ö	49	20	19	3	297
New York	547	292	233	6	9	44	113	91	51	1,386
North Carolina	547	413	279	9	26	99	33	64	37	1,507
North Dakota	9	5	210	1	0	2	1	2	1	23
Ohio	682	485	226	5	11	243	46	128	23	1,849
Oklahoma	239	163	131	5	5	22	25	32	19	641
Oregon	100	58	34	2	2	20	17	22	3	258
Pennsylvania	735	567	452	11	8	96	63	73	32	2,037
Rhode Island	20	8	1	0	Ō	7	4	5	3	48
South Carolina	387	312	183	11	8	110	29	36	10	1,086
South Dakota	21	8	6	0	Ō	2	7	2	4	50
Tennessee	525	407	271	19	11	106	42	64	12	1,457
Texas	1,364	1,012	594	40	26	352	156	131	65	3,740
Utah	73	46	32	0	3	11	7	12	8	192
Vermont	14	6	1	0	0	5	6	1	1	34
Virginia	453	338	156	11	11	160	44	54	17	1,244
Washington	228	134	75	1	1	57	36	40	18	590
West Virginia	79	45	25	4	4	12	8	23	3	203
Wisconsin	186	149	111	4	2	32	11	17	9	521
Wyoming	14	6	5	0	0	1	3	3	2	34
Guam	1	0	0	0	0	0	0	1	0	2
¹ Total number of mu	rders for which	supplemental k	nomicide data w	ere received.						
² Pushed is included i										
	al homicide dat		4							
and supplement	ar normonde dat	a were received	a.							
Totals by Weapon	15,129	10.982	7.032	403	264	3,283	1,591	1.860	696	
recaport	10,120	10,002	.,	.00	204	0,200	,,501	.,000		



Table 20	
Murder	
by State, Types of Weapor	ns, 2017
Total Murders ¹	15,129
Total Firearms	10,982
Handguns	7,032
Rifles	403
Shotguns	264
Firearms (type unknown)	3,283
Knives or Cutting Instruments	1,591
Other Weapons	1,860
Hands, fists, feet, etc. ²	696

¹ Total number of murders for which supplemental homicide data were received.

Pushed is included in hands, fists, feet, etc.

What If It Isn't Online in XLS or CSV?



 WY 2010-2017 Resident Mothers Where the Primary Source of Payment was Medicaid by County Many agencies offer no Excel version;
Older documents;
Variable state, industry or agency standards;

Variable enforcement and compliance;

Smaller departments, sections or offices;

Federal councils, commissions or contractors.

What is the Hardest way?

The hardest way is to transcribe data by hand;



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Next hardest is to copy

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Result from Data.Census.Gov:

Table A. Expectation of life, by age, race, Hispanic origin, race for the non-Hispanic population, and sex: United States, 2015

	All races and origins			White			Black			Hispanic ¹		Non	-Hispanic v	vhite ¹	Non	-Hispanic b	lack ¹	
Age (years)	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total	Male	Female
0	78.7	76.3	81.1	78.9	76.6	81.3	75.5	72.2	78.5	81.9	79.3	84.3	78.7	76.3	81.0	75.1	71.9	78.1
1	78.2	75.8	80.5	78.3	76.0	80.6	75.3	72.1	78.3	81.3	78.7	83.7	78.1	75.7	80.4	75.0	71.8	77.9
5	74.3	71.9	76.6	74.4	72.1	76.7	71.5	68.3	74.4	77.4	74.8	79.7	74.1	71.8	76.4	71.1	67.9	74.0
10	69.3	66.9	71.7	69.4	67.1	71.7	66.5	63.3	69.4	72.4	69.8	74.8	69.2	66.9	71.5	66.1	62.9	69.1
15	64.4	62.0	66.7	64.5	62.2	66.8	61.6	58.4	64.5	67.5	64.9	69.8	64.2	61.9	66.5	61.2	58.0	64.1
20	59.5	57.2	61.8	59.6	57.3	61.9	56.8	53.7	59.6	62.6	60.0	64.9	59.4	57.1	61.6	56.4	53.3	59.3
25	54.8	52.5	56.9	54.8	52.7	57.0	52.1	49.2	54.7	57.8	55.3	60.0	54.6	52.4	56.7	51.8	48.9	54.4
30	50.0	47.9	52.1	50.1	48.0	52.2	47.5	44.7	49.9	53.0	50.6	55.1	49.9	47.8	51.9	47.2	44.4	49.6
35	45.3	43.3	47.3	45.4	43.4	47.4	42.9	40.2	45.2	48.2	45.9	50.3	45.2	43.2	47.2	42.6	39.9	44.9
40	40.7	38.7	42.5	40.7	38.8	42.6	38.3	35.8	40.5	43.5	41.2	45.4	40.6	38.7	42.4	38.1	35.5	40.3
45	36.1	34.2	37.9	36.1	34.3	37.9	33.8	31.4	36.0	38.8	36.6	40.6	36.0	34.1	37.8	33.6	31.1	35.7
50	31.6	29.8	33.3	31.6	29.9	33.3	29.5	27.1	31.6	34.2	32.0	35.9	31.5	29.8	33.2	29.3	26.9	31.3
55	27.3	25.6	28.9	27.3	25.7	28.9	25.4	23.2	27.3	29.7	27.7	31.3	27.3	25.6	28.8	25.3	23.0	27.2
60	23.2	21.7	24.6	23.2	21.7	24.6	21.7	19.6	23.4	25.5			23.2	21.7	24.5	21.5	19.4	23.2
65	19.3	18.0													20.4	18.1	16.2	19.5
70	15.6	14.4		Eirct	Pov	v dr	onn	od i	nto	Eirct		lum	. .				10.0	15.9
75	12.2	11.2		First	NUV	v ui	opp	eui				IUIIII	1,				10.5	12.7
80	9.1	8.3	9.7	9.1	0.3	9.0	9.2	0.2	9.7	10.3				0.3	9.6	9.1	8.1	9.7
85	6.6	5.9	7.0	6.5	5.9	6.9	6.9	6.1	7.2	7.7		8.0	6.5	5.9	6.9	6.8	6.1	7.2
90	4.6	4.1	4.8	4.5	4.0	4.7	5.0	4.5	5.2	5.4	4.7	5.5	4.5	4.0	4.7	5.0	4.5	5.2
95	3.2	2.8	3.3	3.1	2.7	3.2	3.7	3.3	3.8	3.7	3.3	3.8	3.1	2.7	3.2	3.7	3.3	3.8
100	2.2	2.0	2.3	2.2	2.0	2.2	2.7	2.4	2.7	2.7	2.3	2.6	2.2	2.0	2.2	2.7	2.5	2.7

¹Life tables by Hispanic origin are based on death rates that have been adjusted for race and ethnicity misclassification on death certificates. Updated classification ratios were applied: see Technical Notes SOURCE: NCHS, National Vital Statistics System, Mortality

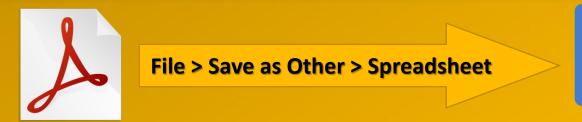
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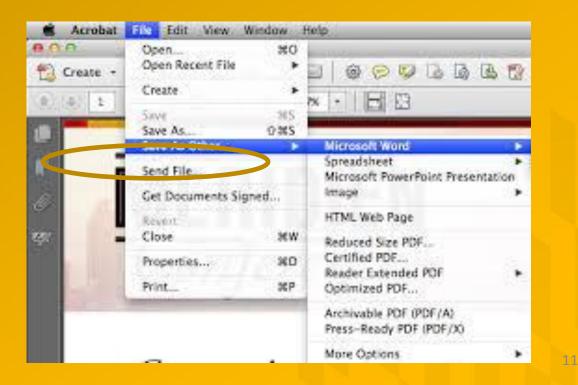
What Next?



- Adobe Acrobat can export PDF to Excel;
- Easy as File, Save as Other \rightarrow xlsx or csv
- Grey-out in Adobe Reader;
- Works only in Adobe Acrobat Pro 10.0;







Can Anyone Export PDF to Excel?

PDF Tables



LOGOUT

*PDF association

- PDF Tables exports to Excel;
- Is it quick? YES;
- Is it easy? YES;
- Is it free? No...
- Well...Is it accurate?
- That depends...



Accurately convert PDF tables to Excel

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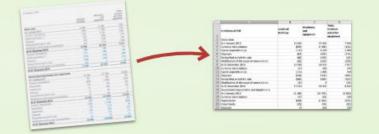
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PDF Tables – One page, One Table



- SALARY TABLE
- 2019-DCB
- One page document;
- One data table.



SALARY TABLE 2019-DCB INCORPORATING THE 1.4% GENERAL SCHEDULE INCREASE AND A LOCALITY PAYMENT OF 29.32% FOR THE LOCALITY PAY AREA OF WASHINGTON-BALTIMORE-ARLINGTON, DC-MD-VA-WV-PA TOTAL INCREASE: 2.27% EFFECTIVE JANUARY 2019

Annual Rates by Grade and Step

Grade	Step 1	Step 2	Step 3	Step 4	Step 5	Step 6	Step 7	Step 8	Step 9	Step 10
1	\$ 24,633	\$ 25,458	\$ 26,278	\$ 27,091	\$ 27,911	\$ 28,390	\$ 29,199	\$ 30,016	\$ 30,049	\$ 30,813
2	27,696	28,356	29,273	30,049	30,386	31,280	32,174	33,067	33,961	34,854
3	30,219	31,227	32,234	33,242	34,249	35,257	36,264	37,271	38,279	39,286
4	33,925	35,055	36,185	37,315	38,446	39,576	40,706	41,836	42,967	44,097
5	37,955	39,220	40,485	41,750	43,014	44,279	45,544	46,809	48,073	49,338
6	42,308	43,719	45,130	46,541	47,952	49,363	50,774	52,184	53,595	55,006
7	47,016	48,583	50,150	51,718	53,285	54,852	56,420	57,987	59,554	61,122
8	52,068	53,804	55,539	57,275	59,010	60,745	62,481	64,216	65,952	67,687
9	57,510	59,426	61,343	63,259	65,176	67,093	69,009	70,926	72,842	74,759
10	63,332	65,442	67,553	69,663	71,774	73,884	75,995	78,105	80,216	82,326
11	69,581	71,901	74,221	76,541	78,861	81,181	83,501	85,821	88,141	90,461
12	83,398	86,179	88,959	91,740	94,520	97,300	100,081	102,861	105,642	108,422
13	99,172	102,477	105,782	109,088	112,393	115,699	119,004	122,310	125,615	128,920
14	117,191	121,098	125,005	128,911	132,818	136,725	140,632	144,538	148,445	152,352
15	137,849	142,443	147,038	151,633	156,228	160,822	165,417	166,500 *	166,500 *	166,500 *

* Rate limited to the rate for level IV of the Executive Schedule (5 U.S.C. 5304 (g)(1)).

Applicable locations are shown on the 2019 Locality Pay Area Definitions page: http://www.opm.gov/policy-data-oversight/pay-leave/salaries-wages/2019/locality-pay-areadefinitions/

PDF Tables – One Page, One Table

- SALARY TABLE ullet
- 2019-DCB ۲
- Imported to PDF Tables; •
- Preview; •
- Download to Excel. ullet



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INCORP	ORATING 1	THE 1.4% GE	NERAL SCH	EDU	LE INCRE	ASE	E AND A LO	OC.	ALITY PAYM	ENT OF	29.3	2%								
FOR THE		PAY AREA	OF WASHIN	GTO	N-BALTIM	10F	RE-ARLING	GTO	DN, DC-MD-	VA-WV-P	A									
TOTAL IN	VCREASE: 2	2.27%																		
EFFECTI	VE JANUA	RY 2019																		
Annual R	ates by Gra	de and Step																		
Grade	Step 1	Step 2	Step 3		Step 4		Step 5		Step 6	Step 7		Step 8		Step 9	St	ep 10				
1 \$	24,633	\$ 25,458	\$ 26,27	8 \$	27,091	\$	27,911	\$	28,390 \$	29,19	9 \$	30,016	\$	30,049	\$	30,813				
2	27,696	28,356	29,27	3	30,049		30,386		31,280	32,17	4	33,067		33,961		34,854				
3	30,219	31,227	32,23	4	33,242		34,249		35,257	36,26	4	37,271		38,279		39,286				
4	33,925	35,055	36,18	5	37,315		38,446		39,576	40,70	6	41,836		42,967		44,097				
5	37,955	39,220	40,48	5	41,750		43,014		44,279	45,54	4	46,809		48,073		49,338				
6	42,308	43,719	45,13	0	46,541		47,952		49,363	50,77	4	52,184		53,595		55,006				
7	47,016	48,583	50,15	0	51,718		53,285		54,852	56,42	0	57,987		59,554		61,122				
8	52,068	53,804	55,53		57,275		59,010		60,745	62,48		64,216		65,952		67,687				
9	57,510	59,426			63,259		65,176		67,093	69,00		70,926		72,842		74,759				
10	63,332	65,442			69,663		71,774		73,884	75,99		78,105		80,216		82,326				
11	69,581	71,901	74,22		76,541		78,861		81,181	83,50		85,821		88,141		90,461				
12	83,398	86,179			91,740		94,520		97,300	100,08		102,861		105,642		108,422				
13	99,172	102,477			109,088		112,393		115,699	119,00		122,310		125,615		128,920				
14	117,191	121,098			128,911		132,818		136,725	140,63		144,538		148,445		152,352				
15	137,849	142,443	147,03	8	151,633		156,228		160,822	165,41	.7	166,500*		166,500*		166,500*				

Applicable locations are shown on the 2019 Locality Pay Area Definitions page: http://www.opm.gov/policy-data-oversight/payleave/salaries-wages/2019/locality-pay-area-

definitions/

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6	42,308	43,719	45,130	46,	41	47,952		49,363		50,774		52,184		53,595		55,006	
7	47,016	48,583	50,150	-		53,285		54,852		56,420		57,987		59,554		61,122	
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10	63,332	65,442	67,553	69,	63	71,774		73,884		75,995		78,105		80,216		82,326	
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13	99,172	102,477	105,782	109,0	88	112,393		115,699		119,004		122,310		125,615		128,920	
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PDF Tables –Long Scholarly Article



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		22	650 Harvard Journal of Law & Public Policy	[Vol. 30	
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* Don B. Kate	s (LL.B., Yale, 1966) is an American criminologist and constitutional	40 France		1.65 [2003]	
lawyer associate tacted at dbkate	d with the Pacific Research Institute, San Francisco. He may be con- stearthlink.net; 360-666-2588; 22608 N.E. 269th Ave., Battle Ground,	41 Derimark		1.21 [2003]	
WA 98604.		42 Greece		1.12 [2003]	
** Gary Maus	er (Ph.D., University of California, Irvine, 1970) is a Canadian crimi- iversity professor at Simon Fraser University, Burnaby, BC Canada.	43 Switzerland		0.99 [2003]	
He may be on	tacted at www.garymauser.net_mauser@shi.ca, and 604-291-3652.	44 Germany		0.93 [2003]	
We gratefully a	knowledge the generous contributions of Professor Thomas B. Cole lorth Carolina at Chapel Hill, Social Medicine and Epidemiology);	45 Norway		0.81 [2001]	
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(Ohio State Un	iversity, History); Professor Thomas Velk (McGill University, Eco-	48 the two data sets given are both available. In every case, we have given			
Weisherg (Stan	airman of the North American Studies Program); Professor Robert ford Law School); and John Whitley (University of Adelaide, Eco-	49 the homicide data for 2003 or the closest year thereto because that is the			
nomics). Any n	serits of this paper reflect their advice and contributions; errors are	50 year of the publication from which the gun ownership data are taken. Gun	l		
entirely ours.		51 ownership data comes from GRADUATE INSTITUTE OF INTERNATIONAL			
		52 STUDIES, SMALL ARMS SURVEY 64 tbl.2.2,65 tbl.2.3 (2003).			
		53 The homicide rate data comes from an annually published report,			6
		54 CANADIAN CENTRE FOR JUSTICE STATISTICS, HOMIGIDE IN CANADA,			6





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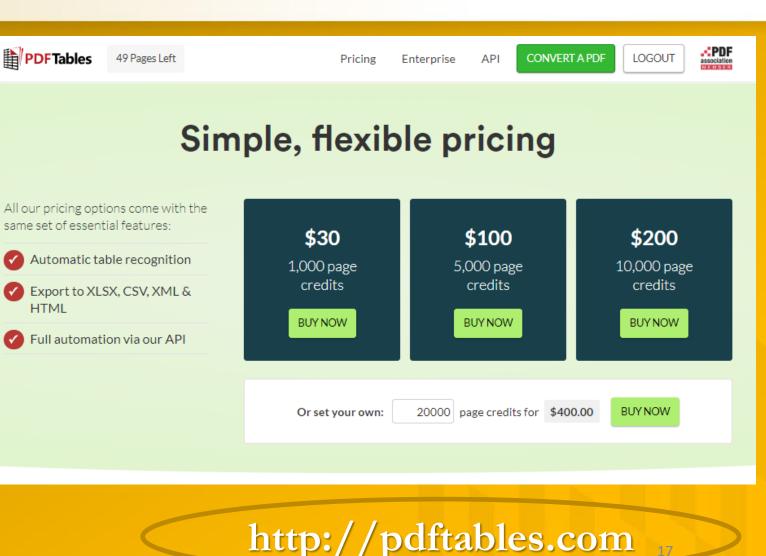
DEPARTMENTS OF LABOR, HEALTH AND HUMAN SERVICES, AND EDUCATION, AND RELATED AGENCIES APPROPRIATIONS FOR FISCAL YEAR 2013

U.S. SENATE, SUBCOMMITTEE OF THE COMMITTEE ON APPROPRIATIONS, Washington, DC. [CLERK'S NOTE.—The subcommittee was unable to hold hearings on nondepartmental witnesses. The statements and letters of those submitting written testimony are as follows:]

DEPARTMENTAL WITNESSES

RAILROAD RETIREMENT BOARD

PREPARED STATEMENT OF MICHAEL S. SCHWARTZ, CHAIRMAN OF THE BOARD



Web Scraper

Scraping for Tech Tigers





 Tiger teams scrape from dynamic pages (Census, Amazon, Monster Jobs, etc.);
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https://www.webscraper.io/

(\$) Scraping Tables (\$)



Best for high-volume harvesting;
Choose the program :

safest; most reliable;
shortest learning curve;
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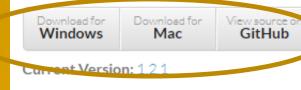
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- Troubleshoot...



Tabula

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Cowmoad the version of Tabula for your operating system:

- Windows: tabula-win.zip
- Mac OS X: tabula-mac.zip
- Linux/Other: tabula-jar.zip, view README.txt inside for instruction

2. Extract the zip file. (Instructions: Windows, Mac)

3. Go into the folder you just extracted. Run the "Tabula" program inside.

A meb browser will open. If it doesn't, open your meb browser, and go to

http://localhost:8080. There's Tabula!

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- 1. Upload a PDF file containing a data table.
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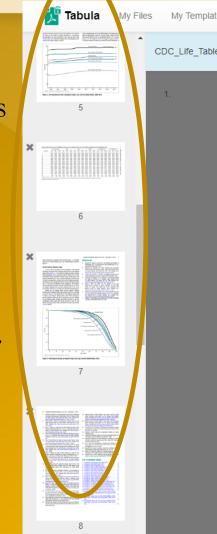
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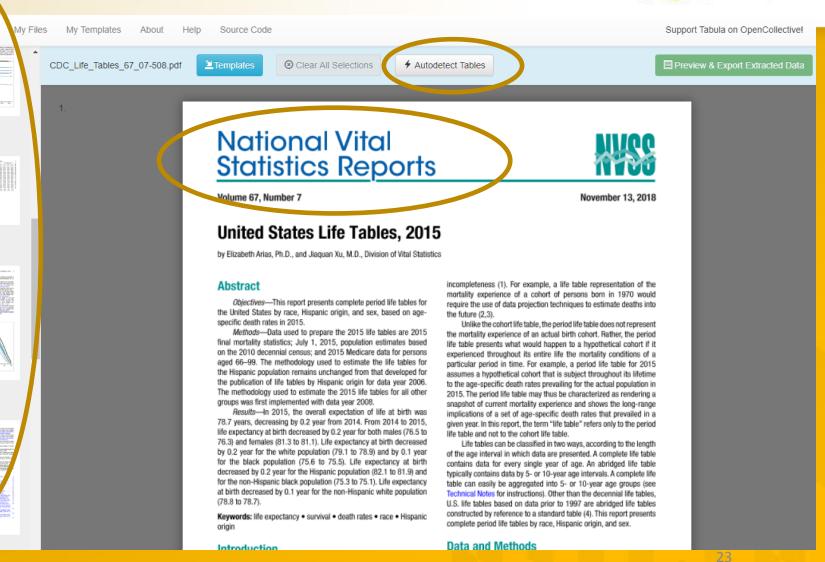
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- X in upper right removes unwanted elements.





4 National Vital Statistics Reports, Vol. 67, No. 7, November 13, 2018

Eigarette smoking (7,8). Between 1979 and 2010, the difference in life expectancy between the sexes narrowed from 7.8 years to 4.8 years (Table 19). The general decline in the sex difference since 1979 reflects proportionately greater increases in lung cancer mortality for women than for men and proportionately larger decreases in heart disease mortality among men (7,8).

The 2015 life table may be used to compare life expectancy at any age from birth onward. On the basis of mortality experienced in 2015, a person aged 65 could expect to live an average of 19.3 more years for a total of 84.3 years; a person aged 85 could expect to live an additional 6.6 years for a total of 91.6 years; and a person aged 100 could expect to live an additional 2.2 years, on average (Table A).

Life expectancy by race

From 2014 to 2015, life expectancy decreased by 0.1 year for the black (75.6 to 75.5) and 0.2 year for the white (79.1 to 78.9) populations (Table 19). The difference in life expectancy between the white and black populations was 3.4 years in 2015, a historically record low level. The white-black difference in life expectancy narrowed from 14.6 years in 1900 to 5.7 years in 1982, but increased to 7.1 years in 1993 before beginning to decline again in 1994 (Table T9). The increase in the gap front 1983 to 1993 was largely the result of increases in mortality among the black male population due to HIV infection and homicide (8).

Among the four race-sex groups, white females continued to have the highest life expectancy at birth (81.3 years), followed by black females (78.5), white males (76.6), and black males (72.2) (Figure 1). Between 2014 and 2015, life expectancy decreased by 0.3 years for black males (72.5 to 72.2). It remained unchanged for black females (78.5). Black males experienced a decline in life expectancy every year for 1984–1989 (8), followed by annual increases in 1990–1992 and 1994–2012. Between 2014 and 2015, life expectancy declined by 0.1 year for white males (76.7 to 76.6) and for white females (81.4 to 81.3). Overall, gains in life expectancy from 1980 through 2015 were 8.4 years for black males, 6.0 years for black females, 5.9 years for white males, and 3.2 years for white females (Table 19).

Life expectancy by Hispanic origin

From 2014 to 2015, life expectancy decreased by 0.2 year for the Hispanic population (82.1 to 81.9) and the non-Hispanic black population (75.3 to 75.1). It decreased by 0.1 year for the non-Hispanic white population (78.8 to 78.7) (Table 19). In 2015, the Hispanic population had a life expectancy advantage at birth of 3.2 years over the non-Hispanic white population and 6.8 years over the non-Hispanic black population. The U.S. life tables by Hispanic origin are based on death rates that have been adjusted for race and ethnicity misclassification on death certificates (see Technical Notes for a detailed description of the methodology).

Repeat t

Among the six Hispanic-origin race-sex groups, Hispanic females continued to have the highest life expectancy at birth (84.3

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[1]



Autodetect Tables:

- Searches and highlights what it believes is tabular data;
- OK to shape tables.



1 of the reporting year are on average exact age x + 1/2.

Preliminary adjustment of the data Adjustment for unknown age

An adjustment is made to account for the small proportion of deaths each year for which age is not reported on the death certificate. The number of deaths in each age category is adjusted proportionally to account for those with not-stated ages. The following factor (F) is used to make the adjustment. F is calculated for the total and for each sex group within a racial and ethnic population for which life tables are constructed:

	D		
-	$\overline{D^{a}}$		

where *D* is the total number of deaths and *D*^a is the total number of deaths for which age is stated. *F* is then applied by multiplying it by the number of deaths in each age group. Table I shows values for *F* by sex used to adjust mortality data for the total, white, black, Hispanic, non-Hispanic white, and non-Hispanic black populations in 2015.

Adjustment for misclassification of race and Hispanic origin on death certificates

The latest research to evaluate race and Hispanic-origin

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Female	1,339,220	43	1.00003211	×	
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Female	1,142,685	37	1.00003238		
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Black	320,072	26	1.00008124		
Male	164,670	21	1.00012754		
		5	1.00003218		
Female	100/405	9	1.00003210		
Hispanic	179,457	6	1.00003344		
		-			
Male	98,170	5	1.00005093		
Female	81,287	1	1.00001230		
Non-Hispanic white	2,123,631	57	1.00002684		
Male		37	1.00003479		
Female		20	1.00001887		
1 GINERO	1,000,000	20	1.00001002		
Non-Hispanic black.	315,254	17	1.00005393		
Male	161,850	12	1.00007415		
Female	153,404	5	1.00003259		

SOURCE: NCHS, National Vital Statistics System, Mortality.

Repeat this Selection

decedent. Further, public policy embodied in the 1997 OMB standard mandates that self-identification should be the standard used for the collection and recording of race and ethnicity information (17).

The NLMS-based classification ratios discussed above are used to adjust the age-specific number of deaths for ages 1–95 and over for the total Hispanic, non-Hispanic white, and non-Hispanic black populations, and by sex for each group, as follows:

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Preview & Export Extracted Data

Preview of Extracted Tabular Data

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<u>Revise</u> Selections to go back and adjust;

Note the fused tables;

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Data has been extracted rom the cells you selected in		All races and origins		White			Black			Hispan 21		Non-Hispanic white1	Non-Hispanic black1
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Shoose Alternate	5	74.3 71.9 76.6	74.4	72.1	76.7	71.5	68.3	74.4	77.4	74 3	79.7	74.1 71.8 76.4	71.1 67.9 74.0
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	65	19.3 18.0 20.5	19.3	18.0	20.5	18.2	16.4	19.6	21.4	19.7	22.0	19.3 18.0 20.4	18.1 16.2 19.5
	70	15.6 14.4 16.6	15.6	14.4	16.5	14.9	13.3	16.0	17.5	16.0	18.5	15.5 14.4 16.5	14.8 13.2 15 9
	75	12.2 11.2 13.0	12.1	11.2	12.9	11.9	10.6	12.7	13.9	12.6	14.6	12 11.1 12.9	11.8 12.5 12.7 27

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Tabula confused by merged cells;

All tables are on one sheet in a pile;

Far columns fused;



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2	Age (year	Total Male	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total Male Fem re	Total Male Female	
3	0	78.7 76.3 8	78.9	76.6	81.3	75.5	72.2	78.5	81.9	79.3	84.3	78.7 76.3 81.0	75.1 71.9 78.1	
4	1	78.2 75.8 8	78.3	76	80.6	75.3	72.1	78.3	81.3	78.7	83.7	78.1 75.7 80 4	75.0 71.8 77.9	
5	5	74.3 71.9 7	74.4	72.1	76.7	71.5	68.3	74.4	77.4	74.8	79.7	74.1 71.8 7 6.4	71.1 67.9 74.0	
6	10	69.3 66.9 7	69.4	67.1	71.7	66.5	63.3	69.4	72.4	69.8	74.8	69.2 66.9 71.5	66.1 62.9 69.1	
7	15	64.4 62.0 6	64.5	62.2	66.8	61.6	58.4	64.5	67.5	64.9	69.8	64.2 61 9 66.5	61.2 58.0 64.1	
8	20	59.5 57.2 6	59.6	57.3	61.9	56.8	53.7	59.6	62.6	60	64.9	59.4 5 .1 61.6	56.4 53.3 59.3	
9	25	54.8 52.5 5	54.8	52.7	57	52.1	49.2	54.7	57.8	55.3	60	54.6 2.4 56.7	51.8 48.9 54.4	
10	30	50.0 47.9 5	50.1	48	52.2	47.5	44.7	49.9	53	50.6	55.1	49. 47.8 51.9	47.2 44.4 49.6	
11	35	45.3 43.3 4	45.4	43.4	47.4	42.9	40.2	45.2	48.2	45.9	50.3	45 2 43.2 47.2	42.6 39.9 44.9	
12	40	40.7 38.7 4	40.7	38.8	42.6	38.3	35.8	40.5	43.5	41.2	45.4	40 6 38.7 42.4	38.1 35.5 40.3	
13	45	36.1 34.2 3	36.1	34.3	37.9	33.8	31.4	36	38.8	36.6	40.6	3 .0 34.1 37.8	33.6 31.1 35.7	
14	50	31.6 29.8 3	31.6	29.9	33.3	29.5	27.1	31.6	34.2	32	35.9	35 29.8 33.2	29.3 26.9 31.3	
15	55	27.3 25.6 2	27.3	25.7	28.9	25.4	23.2	27.3	29.7	27.7	31.3	7.3 25.6 28.8	25.3 23.0 27.2	
16	60	23.2 21.7 2	23.2	21.7	24.6	21.7	19.6	23.4	25.5	23.6	26.9	3.2 21.7 24.5	21.5 19.4 23.2	
17	65	19.3 18.0 2	19.3	18	20.5	18.2	16.4	19.6	21.4	19.7	22.6	9.3 18.0 20.4	18.1 16.2 19.5	
18	70	15.6 14.4 1	15.6	14.4	16.5	14.9	13.3	16	17.5	16	18.5	.5.5 14.4 16.5	14.8 13.2 15.9	
19	75	12.2 11.2 1	12.1	11.2	12.9	11.9	10.6	12.7	13.9	12.6	14.6	2.1 11.1 12.9	11.8 10.5 12.7	
20	80	9.1 8.3 9.7	9.1	8.3	9.6	9.2	8.2	9.7	10.5	9.5	11.1	0.1 8.3 9.6	9.1 8.1 9.7	
21	85	6.6 5.9 7.0	6.5	5.9	6.9	6.9	6.1	7.2	7.7	6.8	8	.5 5.9 6.9	6.8 6.1 7.2	
22	90	4.64.14.8	4.5	4	4.7	5	4.5	5.2	5.4	4.7	5.5	.5 4.0 4.7	5.0 4.5 5.2	
23	95	3.2 2.8 3.3	3.1	2.7	3.2	3.7	3.3	3.8	3.7	3.3	3.8	1 2.7 3.2	3.7 3.3 3.8	
24	100	2.2 2.0 2.3	2.2	2	2.2	2.7	2.4	2.7	2.7	2.3	2.6	2 2.0 2.2	2.7 2.5 2.7	
25		All races ar	nd origins	White			Black			Hispanic1		Non-Hispanic white1	Non-Hispanic black1	
26	Age (year	Total Male	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total Male Female	Total Male Female	
27	0	100,000 10	100,000	100,000	100,000	100,000	100,000	100,000	100,000	100,000	100,000	10 000 100,000 100,000	100,000 100,000 100,0	00
28	1	99,411 99,	99,508	99,467	99,551	98,861	98,760	98,966	99,503	99,465	99,543	99, 10 99,467 99,556	98,875 98,783 98,971	
29	5	99,312 99,	99,419	99,368	99,473	98,708	98,580	98,840	99,426	99,381	99,474	99,4 0 99,358 99,486	98,707 98,602 98,837	
30	10	99,254 99,	99,365	99,307	99,426	98,626	98,488	98,767	99,379	99,327	99,436	99,36 99,293 99,443	98,617 98,510 98,761	
31	15	99,181 99,	99,296	99,228	99,367	98,529	98,373	98,691	99,322	99,263	99,387	99,294 9,204 99,389	98,513 98,395 98,680	
32	20	98,943 98,	99,072	98,928	99,222	98,194	97,868	98,532	99,132	99,006	99,267	99,066 9 903 99,239	98,149 97,848 98,506	
33	25	98,503 98,	98,652	98,328	98,995	97,574	96,926	98,247	98,785	98,509	99,088	98,637 98, 93 99,001	97,482 96,838 98,198	
34	30	97,980 97,	98,137	97,612	98,693	96,863	95,878	97,871	98,403	97,954	98,898	98,087 97,533 98,669	96,742 95,754 97,800	
35	35	97,357 96,	97,518	96,794	98,284	96,021	94,720	97,329	97,970	97,369	98,634	97,417 96,647 8,222	95,860 94,552 97,221	
36	40	96,609 95,	96,782	95,862	97,754	94,972	93,348	96,579	97,465	96,686	98,318	96,618 95,640 97,538	94,739 93,086 96,419	
37	45	95,619 94,	95,808	94,674	96,999	93,595	91,674	95,473	96,782	95,785	97,860	95,571 94,377 96,81	93,285 91,318 95,251	
	-													

Tabula Java Program



- 1. Workaround:
- 2. Create two columns;
- Data → Text to Columns;
- 4. Space-Separated;
- 5. Repeat as needed.



	А	В	с	D	E	F	G	н	I.	J	К	L	м	N	0	Р	Q
1		All races a	nd origins	White			Black			Hispanic1		Non-Hispanic	whit 1		Non-His	panic black	1
2	Age (year	Total Male	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total	Male	Female	Total	Male	Femal
3	0	78.7 76.3 8	78.9	76.6	81.3	75.5	72.2	78.5	81.9	79.3	84.3	78.7	76.3	81	75.1	71.9	78.
4	1	78.2 75.8 8	78.3	76	80.6	75.3	72.1	78.3	81.3	78.7	83.7	75 1	75.7	80.4	75	71.8	77.9
5	5	74.3 71.9 7	74.4	72.1	76.7	71.5	68.3	74.4	77.4	74.8	79.7	4.1	71.8	76.4	71.1	67.9	74
6	10	69.3 66.9 7	69.4	67.1	71.7	66.5	63.3	69.4	72.4	69.8	74.8	69.2	66.9	71.5	66.1	62.9	69.1
7	15	64.4 62.0 6	64.5	62.2	66.8	61.6	58.4	64.5	67.5	64.9	69.8	64.2	61.9	66.5	61.2	58	64.1
8	20	59.5 57.2 6	59.6	57.3	61.9	56.8	53.7	59.6	62.6	60	64.9	59.4	57.1	61.6	56.4	53.3	59.3
9	25	54.8 52.5 5	54.8	52.7	57	52.1	49.2	54.7	57.8	55.3	60	54.6	52.4	56.7	51.8	48.9	54.4
10	30	50.0 47.9 5	50.1	48	52.2	47.5	44.7	49.9	53	50.6	55.1	49.9	47.8	51.9	47.2	44.4	49.6
11	35	45.3 43.3 4	45.4	43.4	47.4	42.9	40.2	45.2	48.2	45.9	50.3	45.2	43.2	47.2	42.6	39.9	44.9
12	40	40.7 38.7 4	40.7	38.8	42.6	38.3	35.8	40.5	43.5	41.2	45.4	40.6	38.7	42.4	38.1	35.5	40.3
13	45	36.1 34.2 3	36.1	34.3	37.9	33.8	31.4	36	38.8	36.6	40.6	36	34.1	37.8	33.6	31.1	35.7
14	50	31.6 29.8 3	31.6	29.9	33.3	29.5	27.1	31.6	34.2	32	35.9	31.5	29.8	33.2	29.3	26.9	31.3
15	55	27.3 25.6 2	27.3	25.7	28.9	25.4	23.2	27.3	29.7	27.7	31.3	27.3	25.6	28.8	25.3	23	27.2
16	60	23.2 21.7 2	23.2	21.7	24.6	21.7	19.6	23.4	25.5	23.6	26.9	23.2	21.7	24.5	21.5	19.4	23.2
17	65	. 19.3 18.0 2	19.3	18	20.5	18.2	16.4	19.6	21.4	19.7	22.6	19.3	18	20.4	18.1	16.2	19.5
18	70	15.6 14.4 1	15.6	14.4	16.5	14.9	13.3	16	17.5	16	18.5	15.5	14.4	16.5	14.8	13.2	15.9
19	75	. 12.2 11.2 1	12.1	11.2	12.9	11.9	10.6	12.7	13.9	12.6	14.6	12.1	11.1	12.9	11.8	10.5	12.7
20	80	9.1 8.3 9.7	9.1	8.3	9.6	9.2	8.2	9.7	10.5	9.5	11.1	9.1	8.3	9.6	9.1	8.1	9.7
21	85	6.6 5.9 7.0	6.5	5.9	6.9	6.9	6.1	7.2	7.7	6.8	8	6.5	5.9	6.9	6.8	6.1	7.2
22	90	4.64.14.8	4.5	4	4.7	5	4.5	5.2	5.4	4.7	5.5	4.5	4	4.7	5	4.5	5.2
23	95	3.2 2.8 3.3	3.1	2.7	3.2	3.7	3.3	3.8	3.7	3.3	3.8	3.1	2.7	3.2			3.8
24	100	2.2 2.0 2.3	2.2	2	2.2	2.7	2.4	2.7	2.7	2.3	2.6	2.2	2	2.2	2.7	2.5	2.7
25		All races a	<u> </u>	White			Black			Hispanic1		Non-Hispanic	white1		Non-His	panic black	1
26	Age (year	Total Male	Total	Male	Female	Total	Male	Female	Total	Male	Female	Tetal Male Fe	male		Total Ma	le Female	
27	0	100,000 10	100,000	100,000	100,000	100,000	100,000	100,000	100,000	100,000		10,000 100,00		00	100,000	100,000 100	,000
28	1	99,411 99,	99,508	99,467	99,551	98,861	98,760	98,966	99,503	99,465	99,543	99, 10 99,467	99,556		98,875 9	8,783 98,971	L
29	5	99,312 99,	99,419	99,368	99,473	98,708	98,580	98,840	99,426	99,381	99,474	99,4 0 99,358	99,486		98,707 9	8,602 98,837	1
30	10	99,254 99,	99,365	99,307	99,426	98,626	98,488	98,767	99,379	99,327	99,436	99,36, 99,293	99,443		98,617 9	8,510 98,761	L
		99,181 99,	99,296	99,228	99,367	98,529	98,373	98,691	99,322	99,263	99,387	99,294 99,204	99,389		98,513 9	8,395 98,680)
		98,943 98,	99,072	98,928	99,222	98,194	97,868	98,532	99,132	99,006	99,267	99,066 93,903	99,239		98,149 9	7,848 98,506	i i
33	25	98,503 98,	98,652	98,328	98,995	97,574	96,926	98,247	98,785	98,509		98,637 98,293	99,001		97,482 9	6,838 98,198	3
34	30	97,980 97,	98,137	97,612	98,693	96,863	95,878	97,871	98,403	97,954	98,898	98,087 97,5 3	98,669		96,742 9	5,754 97,800	נ
35	35	97,357 96,	97,518	96,794	98,284	96,021	94,720	97,329	97,970	97,369	98,634	97,417 96,647	98,222		95,860 94	4,552 97,221	L
36	40	96,609 95,	96,782	95,862	97,754	94,972	93,348	96,579	97,465	96,686		96,618 95,640				3,086 96,419	
37	45	95,619 94,	95,808	94,674	96,999	93,595	91,674	95,473	96,782	95,785	97,860	95,571 94,377	96,813		93,285 93	1,318 95,251	L
38	50	94,158 92,	94,374	92,952	95,864	91,622	89,341	93,833	95,733	94,481	97,083	94,065 92,586	95,601		91,228 8	8,889 93,52	

Star Attraction: Google Sheet Hack

- Plenty of data or directories are still in static HTML tables;
- Lena Groeger, ProPublica.org, has a Google Sheet hack;
- Populates a Google sheet with a static HTML table in a single formula:





NATIONAL INTERAGENCY FIRE GENTER

otal Wildland Fires and Acres (1926-2017)

The National Interagency Coordination Center at NIFC compiles annual wildland fire statistics for federal and state agencies. This information is provided through Situation Reports, which have been in use for several decades. Prior to 1983, sources of these figures are not known, or cannot be confirmed, and were not derived from the current situation reporting process. As a result the figures prior to 1983 should not be compared to later data.

Source: National Interagency Coordination Center

Year	Fires	Acres	
2017	71,499	10,026,086	
2016	67,743	5,509,995	
2015	68,151	10,125,149	
2014	63,312	3,595,613	
2013	47,579	4,319,546	
2012	67,774	9,326,238	
2011	74,126	8,711,367	
2010	71,971	3,422,724	
2009	78,792	5,921,786	
2008	78,979	5,292,468	
2007	85,705	9,328,045	
2006	96,385	9,873,745	
2005	66,753	8,689,389	
2004	65,461	*8,097,880	

National Interagency Fire Center 🙀 🖿

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Google Sheet Hack



• Step One

- Gather these data elements;
- URL;
 type of element;
 first data element.

The target page
Table
0 (starts at the top)

https://www.nifc.gov/fireInfo/fireInfo_stats_totalFires.html



- "table"
- **`**"0"



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fx									fx								
_	A	В	С	D		5	~ 8 7			A	В	с					
1	Total Wildland Fires and					6.				~	В	c					
	The National Interagene wildland fire statistics for is provided through Situ decades. Prior to 1983,	or federal and ation Reports	state agencies. , which have be	This information en in use for sev		fx	A		1	Total Wildland Fires and Acres (1926-2017							
2	be confirmed, and were process. As a result the later data.					1	#N/A			The National Interagency Coordination Center at NIFC compiles annual wildland fire statistics for federal and state agencies. This information is provided through Situation Reports, which have been in use for several							
	Source: National Intera	gency Coordi	nation Center														
3	Year Fires		cres			3					decades. Prior to 1983, sources of th						
4	2018	58,083	8,767,492			4			2		be confirmed, and were not derived						
5	2017	71,499	10,026,086			5					process. As a result the figures prior later						
6	2016	67,743	5,509,995		-						later	uala.					
7	2015	68,151	10,125,149		-	т					Source: National Interage	ancy Coordination Contor					
8 9	2014 2013	63,312 47,579	3,595,613 4,319,546			Error					ŭ						
10	2013	67.774	9.326.238						3	Year	Fires	Acres					
11	2011	74,126	8,711,367						4	2017	71,49	9	10,026,086				
12	2010	71,971	3,422,724						5	2016	67,74	3	5,509,995				
13	2009	78,792	5,921,786						6	2015	68,15	1	10,125,149				
14	2008	78,979	5,292,468														
15	2007	85,705	9,328,045						7	2014	63,31		3,595,613				
16 17	2006	96,385 66,753	9,873,745 8.689.389						8	2013	47,57	9	4,319,546				
18	2003	65,461 *8							9	2012	67,77	4	9,326,238				
19	2003	63,629	3,960,842						10	2011	74.12		8,711,367				
20	2002	73,457	7,184,712		7						,						
21	2001	84,079	3,570,911	/	-	1	р		11	2010	71,97		3,422,724				
22	2000	92,250	7,393,493		F	Rough	Pre	etty	2	2009	78,79	2	5,921,786				
23 24	1999 1998	92,487 81.043	5,626,093 1,329,704		_				13	2008	78,97	9	5,292,468				
24	1998	66,196	2.856.959						14	2007	85,70		9,328,045				
26	1996	96,363	6,065,998														
27	1995	82,234	1,840,546						15	2006	96,38		9,873,745				
28	1994	79,107	4,073,579						16	2005	66,75	3	8,689,389				

Lena Groeger, ProPublica

Intro to Data & Code

LENA GROEGER, PROPUBLICA, SEPTEMBER 2015

I.Data Journalism: What is it & Why Should I Care?

2. How to Get Data From the Web

3. What to Do With Your Data

https://bit.ly/1Kn6Eav



Getting Data Without (Much) Code

LENA GROEGER, PROPUBLICA, SEPTEMBER 2015

Tools You'll Need

Google Chrome »

Firefox and Safari are OK, but all of our examples and tools will be in Chrome. Please don't use Internet Explorer, I beg you.

Google Spreadsheets »

We'll learn a pretty neat trick that let's you grab data with Google Spreadsheets.

Example's We'll Use

Failed Banks: https://www.fdic.gov/bank/individual/failed/banklist.html

School Zone Clusters: http://www.atlanta.k12.ga.us/Page/832

FDA Directory: http://dslo.afdo.org/results/?q=Georgia&unifyfda=1&bystate=1&selected_facets=area_exact:%22100%22

Ready? Let's Get Some Data

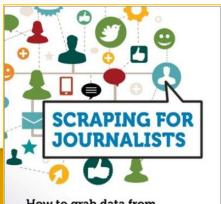
Try a Blank Search. First things first. Often you can just try to search with nothing in the search field, and return ALL the data in a database. Let's try it with this example: http://www.asias.faa.gov/pls/apex/

Look for the Download Button. Often websites with data will have CSV, Excel, or other data download options: http://www.oecd.org/gender/data/employmentandunemploymentratebysexandagegroupquarterlydata.htm

Try Google Spreadsheets. Did you know that you can use Google spreadsheets to pull down an html table? You can using a simple formula: =ImportHTML(&@curl&@l, &@celementtype&@l, numberElement on page 33

Further Information

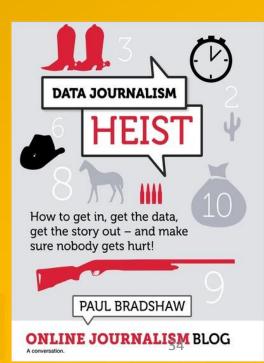
- Paul Bradshaw,
- Master's Program, Birmingham City University,
- Online Journalism Blog: https://onlinejournalismblog.com/;
- Ebooks for sale from LeanPub:
 - Scraping for Journalists (\$20.01) https://leanpub.com/scrapingforjournalists
 - Data Journalism Heist (\$9.99)
 - https://leanpub.com/DataJournalismHeist





How to grab data from hundreds of sources, put it in a form you can interrogate – and still hit deadlines

ONLINE JOURNALISM BLOG





Data Scraping for the Coding-Challenged

Carl P. Olson colson@towson.edu

Thank You!

