Future Insights: Trends in US Computer, Engineering, & Science Occupations

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Dataset Source

https://usa.ipums.org/usa/index.shtml - Customize the dataset



Select year: 5-year (2017-2021)

USA SAMPLES

USA FULL COUNT

PUERTO RICO

Default sample from each year

2022

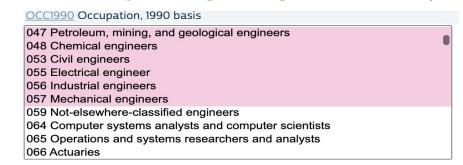
ACS
2021

ACS
2020

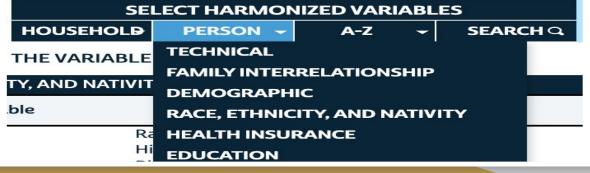
ACS
1

ACS 5yr
ACS 5yr
ACS 5yr

Select computer, engineering, & science occupations

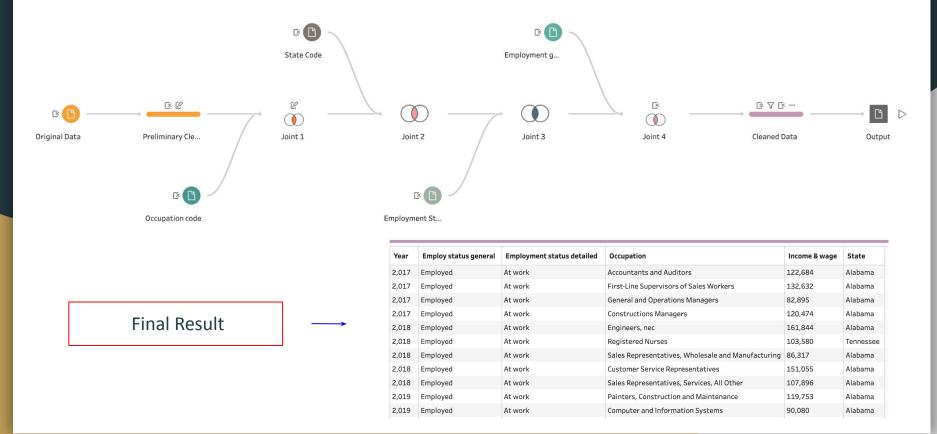


Select variables



Summary Statistics

Employment Trends	Diversity correlation	Educ	ation & skills	Indust	ry & sector
Employment Status	Year	Year	Year of college	Year	Class of worker
Occupation	Gender	Gender	Degree level	Sample	Worked last year
Income & wage	State	Age	State	Serial	Occupation
State	Race	Birth year	Field of degree (General)	Pernum	Industry category
Year	Occupation	Race	Field of degree (Detailed)	Employment status label	Industry subcategory
	Age	Ethnicity	Occupation	Labor force label	State



1) Original Data

Year: 2017-2021

EMPSTAT = Employment Status:

whether the respondent was a part of the labor force,-- working or seeking work -- and unemployed.

OCC = Occupation:

a harmonized occupation coding scheme based on the Census Bureau's ACS occupation classification scheme.

INCWAGW = Income & wage:

each respondent's total pre-tax wage and salary income

PWSTATE2 = State:

the state in which the respondent's primary workplace was located.

YEAR	EMPSTAT	EMPSTATD	OCC	INCWAGE	PWSTATE2
2017	1	10	2310	42000	1
2017	1	10	4220	18789	1
2017	0	0	9920	999999	0
2017	1	10	800	122684	1
2017	1	10	3130	57474	1
2017	1	10	4000	12158	1
2017	3	30	7750	0	0
2017	3	30	5700	0	0
2017	1	10	20	45647	1
2017	1	10	2430	50842	1
2017	0	0	9920	999999	0
2017	0	0	9920	999999	0
2017	2	20	4050	1216	0
2017	3	30	9920	0	0
2017	1	10	4510	0	1
2017	0	0	9920	999999	0
2017	0	0	9920	999999	0
2017	3	30	9920	0	0
2017	3	30	9920	0	0
2017	3	30	9920	0	0
2017	3	30	9920	0	0
2017	1	10	5860	25863	1
2017	2	20	310	14368	0
2017	3	30	9920	0	0

2) Code-to-label

YEAR	EMPSTAT	EMPSTATD	occ	INCWAGE	PWSTATE2	1
2017	1	10	2310	42000	1	l
2017	1	10	4220	18789	1] [



State code	State name
0	N/A
1	Alabama
2	Alaska
4	Arizona
5	Arkansas
6	California

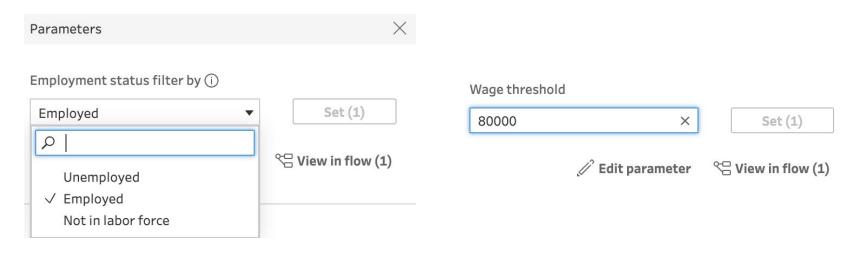


Year	Employ status general	Employment status detailed	Occupation	Income & wage	State
2,017	Employed	At work	Accountants and Auditors	122,684	Alabama
2,017	Employed	At work	First-Line Supervisors of Sales Workers	132,632	Alabama
2,017	Employed	At work	General and Operations Managers	82,895	Alabama

 Using join to connect the code file with original data to transfer the number to text

• Remove the N/A value

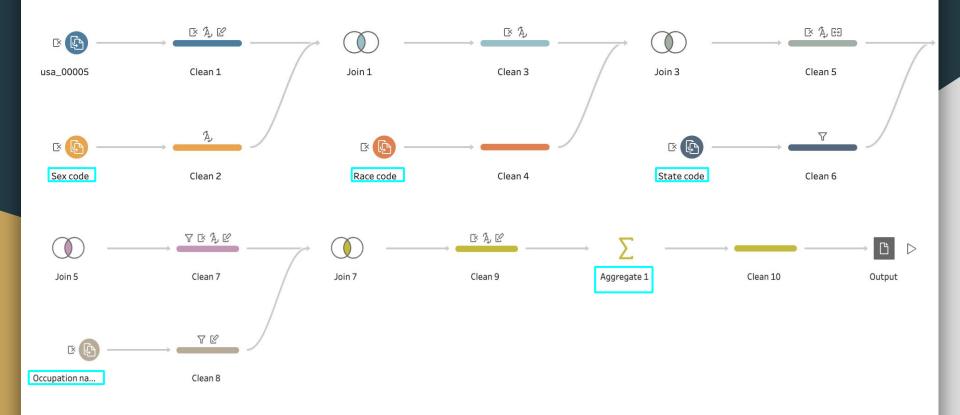
3) Create Parameters



• Filter the employment status: Employed

Filter the wage range

Data Cleaning and Preparation Operations - Diversity and Inclusion



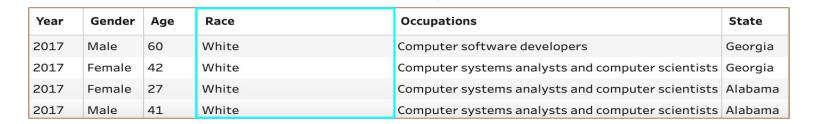
Data Cleaning and Preparation Operations - Diversity and Inclusion

- 1) Clean and Prepare the Data
- Delete null values, filtering out irrelevant data, correcting data types
- Use Join to turn all the numbers to words

Age Code	Age	Sex Code	Sex	Race Code	Race
0	Less than 1 year old	1	Male	1	White
1	1	2	Female	2	Black/African American
2	2			3	American Indian or Alaska Native
3	3			4	Chinese
4	4			5	Japanese
5	5			6	Other Asian or Pacific Islander
6	6			7	Other race, nec
7	7			8	Two major races
8	8			9	Three or more major races

					Applied Join Clauses
MULTYEAR	SEX	AGE	RACE	Occupations	Clean 3 Clean 4 RACE = Race Code
2017	1	60	1	229	
2017	2	64	1	22	Join Type: Inner Click the graphic to change the join type.
2017	1	64	1	4	Clean 3 Clean 4
2017	2	62	1	999	Summary of Join Results
2017	2	63	1	999	Click the bar segments to view the included and excluded values.
2017	1	56	1	999	/// Mismatched values
2017	2	81	1	999	Clean 3 5,456
2017	2	83	2	999	Clean 4 9
2017	2	51	1	678	
					Join Result 5,456

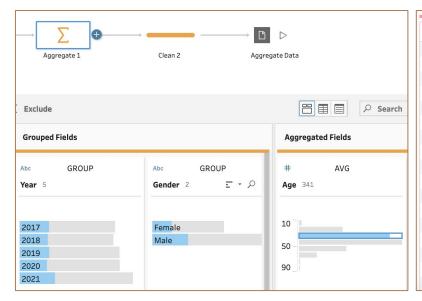
Applied Jain Clauses



Data Cleaning and Preparation Operations - Diversity and Inclusion

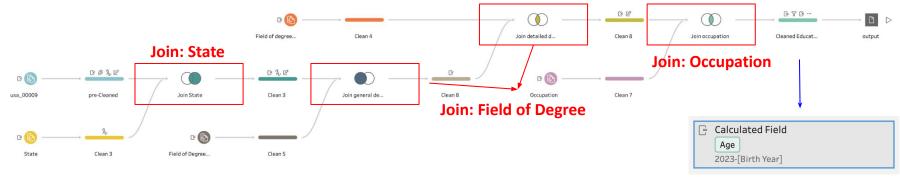
- 2) Create an Aggregation Step
 - Find the average age for each year

- 3) Final result
 - Output data and run the flow



		•			
Year	Gender	State	Race	Occupation	AGE
2020	Male	North Carolina	Other Asian or Pacific Islander	Not-elsewhere-classified engineers	47
2021	Male	Florida	Other Asian or Pacific Islander	Computer systems analysts and computer scientists	35.5
2019	Male	Florida	Other race, nec	Computer systems analysts and computer scientists	22
2021	Female	New Jersey	Two major races	Electrical engineer	55
2021	Female	Texas	Two major races	Computer systems analysts and computer scientists	45.6666666666666
2021	Male	Massachusetts	White	Computer systems analysts and computer scientists	43.875
2017	Male	N/A	White	Industrial engineers	85
2020	Male	California	Black/African American	Computer systems analysts and computer scientists	40.6666666666666
2021	Female	Maryland	Chinese	Computer systems analysts and computer scientists	45
2019	Female	Massachusetts	White	Not-elsewhere-classified engineers	25.5
2019	Female	North Carolina	Other Asian or Pacific Islander	Computer systems analysts and computer scientists	32
2021	Male	New York	White	Not-elsewhere-classified engineers	49.6666666666666
2020	Male	Vermont	White	Computer systems analysts and computer scientists	35
2018	Male	Michigan	White	Computer systems analysts and computer scientists	40.54545454545455
2020	Male	Pennsylvania	White	Aerospace engineer	59
2019	Female	Georgia	White	Industrial engineers	25
2021	Female	N/A	Chinese	Computer systems analysts and computer scientists	22

Data Cleaning and Preparation Operations - Education and Skill

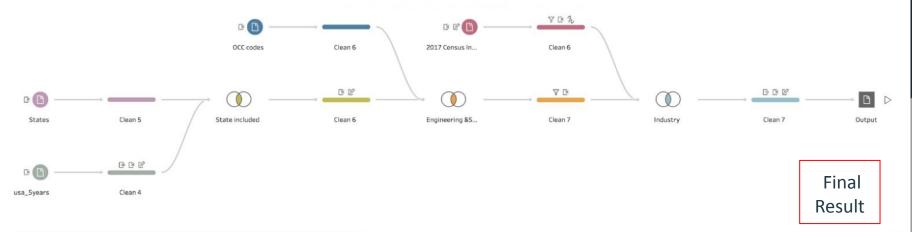


Calculate: Age

Et a a l	2017	Fema
Final	2017	Male
Result	2017	Male
	2017	Eom

	Year	Gender	Birth Year	Age	Race	Ethnicity	Year of College	Degree Level	State	Field of Degree(General)	Field of degree(Detailed)	Occupation
	2017	Male	1,991	32	Other Asian or Pacific Islander	Not Hispanic or Latino	5+ years of college	Master's degree	California	Business	Accounting	Computer software developers
ا_	2017	Female	1,974	49	White	Not Hispanic or Latino	4 years of college	Bachelor's degree	California	Business	General Business	Computer software developers
	2017	Female	1,993	30	Other Asian or Pacific Islander	Not Hispanic or Latino	5+ years of college	Master's degree	California	Engineering	Computer Engineering	Computer systems analysts and computer scientists
										3	Computer Science	Computer systems analysts and computer scientists
	2017	Male	1,961	62	White	Not Hispanic or Latino	4 years of college	Bachelor's degree	California	Computer and Informat	Business Management and Administration	Computer systems analysts and computer scientists
	2017	Male	1,968	55	White	Not Hispanic or Latino	4 years of college	Bachelor's degree	California	Business	Computer Science	Computer systems analysts and computer scientists
ᆈ	2017	Female	1,969	54	Two major races	Hispanic or Latino	4 years of college	Bachelor's degree	California	Computer and Informat	Mechanical Engineering	Aerospace engineer
	2017	Male	1,976	47	Black/African American	Not Hispanic or Latino	5+ years of college	Master's degree	California	Engineering	N/A	Drafters
	2017	Male	1,965	58	White	Not Hispanic or Latino	2 years of college	Associate's degree	California	N/A	Drama and Theater Arts	Computer systems analysts and computer scientists
	2017		1,975	48		Not Hispanic or Latino	, ,		California	,	Civil Engineering	Civil engineers
	201/	Female	1,5/3	40	Diacky African Affierican	NOT HISPAUL OF LATINO	5+ years of college	Doctor al degree	Camorilla	FILLE ALLS	Business Management and Administration	Not-elsewhere-classified engineers

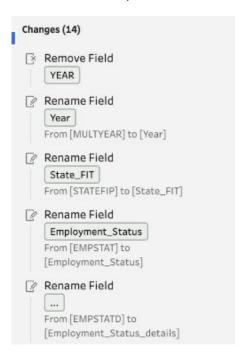
Data Cleaning and Preparation Operations - Industry and Sector



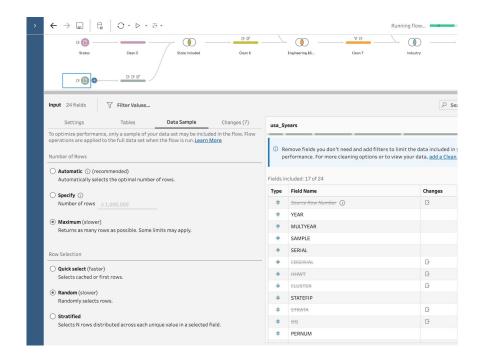
Year	SAMPLE	SERIAL	PERNUM	Employment_Status_Label	Labor_Force_Label	Occupation	Industry Category	Industry Subcategories
2,018	202,103	1,790,892	2	Employed	Yes, in the labor force	Other life, physical, and social science technicians	Educational Services, and Health Care and Social Assist	Colleges, universities, and professional schools, ir
2,018	202,103	1,818,306	1	Employed	Yes, in the labor force	Computer occupations, all other	Public Administration	Executive offices and legislative bodies
2,018	202,103	1,796,288	1	Employed	Yes, in the labor force	Computer occupations, all other	Public Administration	Executive offices and legislative bodies
2,018	202,103	1,818,728	2	Employed	Yes, in the labor force	Computer occupations, all other	Educational Services, and Health Care and Social Assist	Elementary and secondary schools
2,018	202,103	1,819,794	2	Employed	Yes, in the labor force	Computer programmers	Educational Services, and Health Care and Social Assist	Elementary and secondary schools
2,018	202,103	1,818,100	2	Employed	Yes, in the labor force	Architects, except landscape and naval	Public Administration	Executive offices and legislative bodies
2,018	202,103	1,799,860	1	Employed	Yes, in the labor force	Computer programmers	Educational Services, and Health Care and Social Assist	Home health care services
2,018	202,103	1,799,860	2	Employed	Yes, in the labor force	Industrial engineers, including health and safety	Transportation and Warehousing	Air transportation
2,018	202,103	1,821,653	1	Employed	Yes, in the labor force	Computer occupations, all other	Military	U. S. Army
2,019	202,103	1,855,481	1	Employed	Yes, in the labor force	Computer programmers	Finance and Insurance	Insurance carriers
2,019	202,103	1,837,223	1	Employed	Yes, in the labor force	Software developers	Real Estate and Rental and Leasing	Lessors of real estate, and offices of real estate $\boldsymbol{a}_{\boldsymbol{\xi}}$

Data Cleaning and Preparation Operations - Industry and Sector

1) Remove unnecessary columns

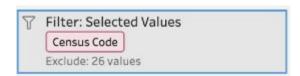


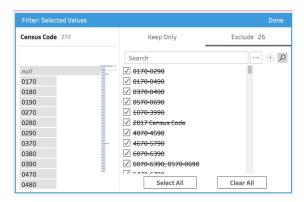
2) Sampling data



Data Cleaning and Preparation Operations - Industry and Sector

3) Filtering rows





4) Create a Calculated field "Industry Category" to replace codes with labels



Calculation is valid ^

SERIAL is an 8-digit numeric variable which assigns a unique identification number to each household record in a given sample (See PERNUM for the analogous person record identifier). A combination of SAMPLE and SERIAL provides a unique identifier for every household in the IPUMS; the combination of SAMPLE, SERIAL, and PERNUM uniquely identifies every person in the database. SERIAL specific variable codes for missing, edited, or unidentified observations, observations not applicable (N/A), observations not in universe (NIU), top and bottom value coding, etc. are provided below if applicable by Census year (and data sample if specified).

SAMPLE	SERIAL	PERNUM
202,103	1,790,892	2
202,103	1,818,306	1
202,103	1,796,288	1
202,103	1,818,728	2
202,103	1,819,794	2
202,103	1,818,100	2
202,103	1,799,860	1
202,103	1,799,860	2
202,103	1,821,653	1
202,103	1,855,481	1
202,103	1,837,223	1
202,103	1,835,499	1
202,103	1,849,553	1
202,103	1,848,136	1
202,103	1,837,649	1
202,103	1,831,714	1
202,103	1,840,256	1
202,103	1,845,952	1

Is it possible to merge files or add columns if future?

Challenges

- 1. Data Selection
 - a. Inflexible government data that are cleaned and well-structured
 - b. Detailed data needed for in-depth analysis

- 2. Data Cleaning
 - a. Large data- Filtering and sampling
 - b. Data with confusing code
 - i. Join
 - ii. define parameters
 - c. Files with different sample size



Insights from data

Employment

- Trend Over Time
- Geographical Analysis
- Wage Distribution

Industry

- Occupation Distribution
- Demands Over Time

Diversity

- Gender & Age
- Race & Ethnicity

Education

- Common majors
- Education Levels



