DR. ALVIN'S PUBLICATIONS

DATA WRANGLING A Population of Countries dataset

WITH PYTHON By Dr. Alvin Ang



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I. STEP 1: IMPORT THE DATASET

- You may find the file here: <u>https://www.alvinang.sg/s/Population-of-Countries-in-2000.csv</u>
- <u>https://www.alvinang.sg/s/Data Wrangling a Population of Countries Dataset by Dr</u> <u>Alvin Ang.ipynb</u>

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A. TAKE A PEEK AT THE DATASET

	A B	C	D	E	F	G	Н	I	
1	country	country_isocode	year	рор	xrat	tcgdp	СС	cg	
2	0 Argentina	ARG	2000	37335.653	0.9995	295072.21869	75.716805379	5.5788042896	
3	1 Australia	AUS	2000	19053.186	1.72483	541804.6521	67.759025993	6.7200975332	
4	2 India	IND	2000	1006300.297	44.9416	1728144.3748	64.575551328	14.072205773	
5	3 Israel	ISR	2000	6114.57	4.07733	129253.89423	64.436450847	10.266688415	
6	4 Malawi	MWI	2000	11801.505	59.543808333	5026.2217836	74.707624181	11.658954494	
7	5 South Africa	ZAF	2000	45064.098	6.93983	227242.36949	72.718710427	5.7265463933	
8	6 United States	USA	2000	282171.957	1	9898700	72.347054303	6.0324539789	
9	7 Uruguay	URY	2000	3219.793	12.099591667	25255.961693	78.978740282	5.108067988	
10									

• Legend:

- o country: country name
- o country iso code: Country code
- o year: 2000
- o pop: population in thousands
- o xrat: exchange rate to US dollar (national currency units per US dollar)
- o tcgdp: total PPP converted GDP, Geary-Khamis method, at current prices (in milion International dollar)
- o cc: consumption share of PPP converted GDP per capita at current prices
- cg: government consumption share of PPP converted GDP per capita at current prices

•	Ste	p 1: Imp	or	t the Da	ataset							
> 1s	D	import p	and	las as po	d							
		df = pd.	rea	nd_csv('	https://www.	alvi	.nang.sg	/s/Pop	ulation-of	-Count	ries-in	-2000.csv')
		df.sampl	e()									
	C→	Unnamed:	0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg	
		5		South Africa	ZAF	2000	45064.098	6.93983	227242.36949	72.71871	5.726546	

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II. STEP 2: PREVIEW THE DATASET

	 Step 2: Preview the Dataset 															
C		D	df.	shap	е											
		₽	(8,9	(8, 9)												
c		[6]	df.	df.info()												
			<clas Range Data # Ø 1 2 3 4 5 6 7 8 dtype memor</clas 	s 'pa Index colum Colum Unnam count count year pop xrat tcgdp cc cg s: fl y usa	ndas.core : 8 entri ns (total n - ed: 0 ry ry_isocod oat64(5), ge: 704.0	.frame. es, 0 t 9 colu Non- 8 no 8 no 8 no 8 no 8 no 8 no 8 no 8 no	DataFrame o 7 mns): Null Coun n-null n-null n-null n-null n-null n-null n-null 2), objec	'> t Dtyp int6 obje obje int6 floa floa floa floa tloa	e - 4 ct ct 4 t64 t64 t64 t64							
	df	.desc	ribe()													
<u>`</u>		Un	named: 0	year	рор	xrat	tcgdp	cc	cg							
	CO	unt	8.00000	8.0	8.000000e+00	8.000000	8.000000e+00	8.000000	8.000000							
	me	an	3.50000	2000.0	1.763826e+05	16.415811	1.606312e+06	71.404995	8.145477							
	s	td	2.44949	0.0	3.479223e+05	22.758175	3.397025e+06	5.318015	3.383397							
	m	in	0.00000	2000.0	3.219793e+03	0.999500	5.026222e+03	64.436451	5.108068							
	25	6%	1.75000	2000.0	1.037977e+04	1.543623	1.032544e+05	66.963157	5.689611							
	50	%	3.50000	2000.0	2.819442e+04	5.508580	2.611573e+05	72.532882	6.376276							
	75	6%	5.25000	2000.0	1.043411e+05	20.310094	8.383896e+05	74.959919	10.614755							
	m	ax	7.00000	2000.0	1.006300e+06	59.543808	9.898700e+06	78.978740	14.072206							

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Ste	ep 3:	Check f	or Miss	sing Da	ta
D	df.i	.sna().any	/()		
	#the	ere's no m	nissing	data!	
¢	Unname countr year pop xrat tcgdp cc cg dtype:	ed: 0 ry ry_isocode : bool	False False False False False False False False False		
	D	df.isnu #there'	11 ().s s no N	aum() IaNs!	
	Ċ	Unnamed: 0 country country_is year pop xrat tcgdp cc cg dtype: int	ocode 64	0 0 0 0 0 0	

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IV. STEP 4: DEALING WITH MISSING DATA

Step 4: Deal with Missing Data

4a) Drop NaNs

4a)(i) Drop off Columns that have NaNs

A. DROP NANS

df	df.dropna(axis = 'columns')												
#c	#drop off columns that have NaNs												
	Unnamed:	0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg			
0		0	Argentina	ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804			
1		1	Australia	AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098			
2		2	India	IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206			
3		3	Israel	ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688			
4		4	Malawi	MWI	2000	11801.505	59.543808	5.026222e+03	74.707624	11.658954			
5		5	South Africa	ZAF	2000	45064.098	6.939830	2.272424e+05	72.718710	5.726546			
6		6	United States	USA	2000	282171.957	1.000000	9.898700e+06	72.347054	6.032454			
7		7	Uruguay	URY	2000	3219.793	12.099592	2.525596e+04	78.978740	5.108068			

1. DROP OFF COLUMNS THAT HAVE NANS

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2. I	DROP	OFF	ROWS	THAT	HAVE NANS	,
------	------	-----	------	------	-----------	---

4a)(a)(ii) Drop off Rows that have NaNs											
[]	df.dropn	a(a	axis = 'r	ows')								
	#drop of	f١	rows that	have NaNs								
	Unnamed:	0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg		
	0	0	Argentina	ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804		
	1	1	Australia	AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098		
	2	2	India	IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206		
	3	3	Israel	ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688		
	4	4	Malawi	MWI	2000	11801.505	59.543808	5.026222e+03	74.707624	11.658954		
	5	5	South Africa	ZAF	2000	45064.098	6.939830	2.272424e+05	72.718710	5.726546		
	6	6	United States	USA	2000	282171.957	1.000000	9.898700e+06	72.347054	6.032454		
	7	7	Uruguay	URY	2000	3219.793	12.099592	2.525596e+04	78.978740	5.108068		

B. FILL UP NANS WITH OTHER VALUES

1. FILL UP NANS WITH 0

4b)	b) Fill Up NaNs with Other Values											
4b)(b)(i) Fill Up NaNs with 0											
D	df.filln	a(0)									
	#fill up		ll NaNs w	ith 0								
C⇒	Unnamed:	0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg		
	0		Argentina	ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804		
	1	1	Australia	AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098		
	2	2	India	IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206		
	3	3	Israel	ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688		
	4	4	Malawi	MWI	2000	11801.505	59.543808	5.026222e+03	74.707624	11.658954		
	5		South Africa	ZAF	2000	45064.098	6.939830	2.272424e+05	72.718710	5.726546		
	6		United States	USA	2000	282171.957	1.000000	9.898700e+06	72.347054	6.032454		
	7	7	Uruguay	URY	2000	3219.793	12.099592	2.525596e+04	78.978740	5.108068		

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2. FORWARD FILL

4b)	4b)(ii) Forward Fill														
0	df.fillna(method = 'ffill')														
C→	Unname	d: 0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg					
	0		Argentina	ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804					
	1	1	Australia	AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098					
	2	2	India	IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206					
	3	3	Israel	ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688					
	4	4	Malawi	MWI	2000	11801.505	59.543808	5.026222e+03	74.707624	11.658954					
	5		South Africa	ZAF	2000	45064.098	6.939830	2.272424e+05	72.718710	5.726546					
	6		United States	USA	2000	282171.957	1.000000	9.898700e+06	72.347054	6.032454					
	7	7	Uruguay	URY	2000	3219.793	12.099592	2.525596e+04	78.978740	5.108068					

3. BACKWARD FILL

4b)	4b)(iii) Backward Fill														
D	df.filln	a(method =	'bfill')											
C→	Unnamed:	0	country	country_isocode	year	рор	xrat	tcgdp	cc	cg					
	0		Argentina	ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804					
	1	1	Australia	AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098					
	2	2	India	IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206					
	3	3	Israel	ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688					
	4	4	Malawi	MWI	2000	11801.505	59.543808	5.026222e+03	74.707624	11.658954					
	5		South Africa	ZAF	2000	45064.098	6.939830	2.272424e+05	72.718710	5.726546					
	6		United States	USA	2000	282171.957	1.000000	9.898700e+06	72.347054	6.032454					
	7	7	Uruguay	URY	2000	3219.793	12.099592	2.525596e+04	78.978740	5.108068					

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∕ 0s	[6]	df	.col	umns									
		Ind	lex(<mark>[' </mark> dt	Unnamed: 0' tcgap', 'cc ype='object	'count ', 'cg'] ')	ry', 'cou ,	untry_	isocode', 'y	year', 'po	p', 'xrat',	^ ↓ @		
✓ 0s	Þ	df df	.col	umns =	'S/N'	.'coun	ıtry'	,'countr	ry_isoc	ode','yea	r','pop	o','xra	t','t(
			S/N	country	country	_isocode	year	рор	xrat	tcgdp	cc	cg	<i>7</i> .
		0		Argentina		ARG	2000	37335.653	0.999500	2.950722e+05	75.716805	5.578804	
		1	1	Australia		AUS	2000	19053.186	1.724830	5.418047e+05	67.759026	6.720098	
		2		India		IND	2000	1006300.297	44.941600	1.728144e+06	64.575551	14.072206	
		3	3	Israel		ISR	2000	6114.570	4.077330	1.292539e+05	64.436451	10.266688	
_													

V. STEP 5: RENAMING COLUMNS

A. USING .RENAME

<pre>[12] df = df.rename(columns={"country_isocode": "blablabla"})</pre>									
[13]	df.columns								
	<pre>Index(['S/N', 'country',</pre>	'blablabla'	'year', 'pop', 'xrat',	'tcgdp', 'cc',					

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VI. STEP 6: FILTERING A ROW

Step 6: Filtering a Row													
	[] #Filtering out United States												
			-	· .					-				
	\$	S/N	count	ry bla	blabla	year		рор	xrat	tcgdp	cc	cg	
	6	6	United Sta	tes	USA	2000	282171	1.957	1.0	9898700.0	72.347054	6.032454	
Ø	#Fi	lte	ering o	out Au	stral	ia							
	df[df	['count	ry']	== 'A	ustr	ralia	']					
C→	\$	S/N	country	blablab	la yea	r	рор	xr	at	tcgdp	cc	cg	
	1	1	Australia	Al	JS 200	0 190)53.186	1.724	83 5	641804.6521	67.759026	6.720098	

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VII. STEP 7: FILTERING A COLUMN



0	#OR										
	df	:[['у	ear', 'co	ountry']]							
C⇒	C→ year country										
	0	2000	Argentina								
	1	2000	Australia								
	2	2000	India								
	3	2000	Israel								
	4	2000	Malawi								
	5	2000	South Africa								
	6	2000	United States								
	7	2000	Uruguay								

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Step 8: Searching Out Using REGEX

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D	df #c	.filter(r	egex	x = 'y\$	', axis	s = 1)		1.71
	#5	earching	out	COTUMU	names	епатив	WICH	У
C→		country						
	0	Argentina						
	1	Australia						
	2	India						
	3	Israel						
	4	Malawi						
	5	South Africa						
	6	United States						
	7	Uruguay						

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Step 9: Searching Out Using LIKE												
0	df.filter(like = 'p', axis = 1)											
	#1	.ike 'p' n	neans to s	earch	for any	column	name	that	has	'p'	in	it
C→		рор	tcgdp									
	0	37335.653	2.950722e+05									
	1	19053.186	5.418047e+05									
	2	1006300.297	1.728144e+06									
	3	6114.570	1.292539e+05									
	4	11801.505	5.026222e+03									
	5	45064.098	2.272424e+05									
	6	282171.957	9.898700e+06									
	7	3219.793	2.525596e+04									
TH	THE END											

IX. STEP 9: SEARCHING USING LIKE

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ABOUT DR. ALVIN ANG

Dr. Alvin Ang earned his Ph.D., Masters and Bachelor degrees from NTU, Singapore. He is a scientist, entrepreneur, as well as a personal/business advisor. More about him at <u>www.AlvinAng.sg</u>.

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