

```
In [1]: import wbgapi as wb
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In [3]: #Data Frame-1
#Climate change indicators
indicator_ids1 = ['EG.ELC.ACCS.ZS', 'ER.H2O.FWTL.ZS', 'AG.LND.FRST.ZS', 'AG.LND.IRIG.AG
country_codes = ['USA', 'CAN', 'IND', 'JPN', 'CHN', 'THA', 'CRI', 'BAN', 'AUS', 'AFG']
my_dataframe1 = wb.data.DataFrame(indicator_ids1, country_codes, mrv=10)
print(my_dataframe1)
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		YR2011	YR2012	YR2013	YR2014 \
economy series					
AFG	AG.LND.FRST.ZS	1.850994	1.850994	1.850994	1.850994
	AG.LND.IRIG.AG.ZS	5.391717	5.465576	5.518333	5.742548
	EG.ELC.ACCS.ZS	43.222019	69.100000	68.982941	89.500000
	ER.H2O.FWTL.ZS	NaN	43.015907	NaN	NaN
AUS	AG.LND.FRST.ZS	16.955310	17.047689	17.140067	17.232446
	AG.LND.IRIG.AG.ZS	0.495556	0.553121	0.639634	0.630650
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	3.133875	NaN	NaN
CAN	AG.LND.FRST.ZS	38.734864	38.730258	38.725651	38.721045
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	1.278596	NaN	NaN
CHN	AG.LND.FRST.ZS	21.491096	21.696596	21.902095	22.107595
	AG.LND.IRIG.AG.ZS	10.206754	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	99.857224	99.964874	99.997009	100.000000
	ER.H2O.FWTL.ZS	NaN	21.286882	NaN	NaN
CRI	AG.LND.FRST.ZS	56.552409	56.872934	57.193459	57.513984
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	99.232348	99.503298	99.563517	99.359291
	ER.H2O.FWTL.ZS	NaN	2.069912	NaN	NaN
IND	AG.LND.FRST.ZS	23.463822	23.553422	23.643023	23.732624
	AG.LND.IRIG.AG.ZS	35.398787	36.329478	36.785607	37.900969
	EG.ELC.ACCS.ZS	67.600000	79.900000	80.738045	83.585213
	ER.H2O.FWTL.ZS	NaN	44.778700	NaN	NaN
JPN	AG.LND.FRST.ZS	68.481756	68.469684	68.457613	68.445542
	AG.LND.IRIG.AG.ZS	34.509976	34.710925	35.191714	34.816290
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	18.888837	NaN	NaN
THA	AG.LND.FRST.ZS	39.285560	39.280863	39.276165	39.271467
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	99.359283	99.108624	99.427063	99.483643
	ER.H2O.FWTL.ZS	NaN	25.526503	NaN	NaN
USA	AG.LND.FRST.ZS	33.779470	33.809533	33.839596	33.869659
	AG.LND.IRIG.AG.ZS	NaN	5.604332	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	15.224927	NaN	NaN

		YR2015	YR2016	YR2017	YR2018 \
economy series					
AFG	AG.LND.FRST.ZS	1.850994	1.850994	1.850994	1.850994
	AG.LND.IRIG.AG.ZS	5.710894	6.481140	5.990504	5.134629
	EG.ELC.ACCS.ZS	71.500000	97.700000	97.700000	98.715622
	ER.H2O.FWTL.ZS	NaN	NaN	43.015907	NaN
AUS	AG.LND.FRST.ZS	17.324825	17.425488	17.422914	17.421315
	AG.LND.IRIG.AG.ZS	0.617346	0.626675	0.603490	0.640299
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	NaN	3.240854	NaN
CAN	AG.LND.FRST.ZS	38.716438	38.712013	38.707888	38.703763
	AG.LND.IRIG.AG.ZS	1.559886	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000

	ER.H2O.FWTL.ZS	NaN	NaN	1.253684	NaN
CHN	AG.LND.FRST.ZS	22.313094	22.542877	22.742310	22.941735
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	NaN	21.038038	NaN
CRI	AG.LND.FRST.ZS	57.834508	58.155112	58.475715	58.796122
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	99.409804	99.500000	99.600000	99.700000
	ER.H2O.FWTL.ZS	NaN	NaN	2.826549	NaN
IND	AG.LND.FRST.ZS	23.822225	23.911825	24.001426	24.091027
	AG.LND.IRIG.AG.ZS	38.059486	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	88.000000	89.534882	92.456833	95.193298
	ER.H2O.FWTL.ZS	NaN	NaN	44.778700	NaN
JPN	AG.LND.FRST.ZS	68.433471	68.422497	68.408779	68.408779
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	NaN	18.888837	NaN
THA	AG.LND.FRST.ZS	39.266770	39.180646	39.110180	39.039715
	AG.LND.IRIG.AG.ZS	NaN	NaN	NaN	NaN
	EG.ELC.ACCS.ZS	99.600000	99.852127	99.900000	99.820000
	ER.H2O.FWTL.ZS	NaN	NaN	25.526503	NaN
USA	AG.LND.FRST.ZS	33.899723	33.899723	33.866926	33.866926
	AG.LND.IRIG.AG.ZS	NaN	NaN	5.785517	NaN
	EG.ELC.ACCS.ZS	100.000000	100.000000	100.000000	100.000000
	ER.H2O.FWTL.ZS	NaN	NaN	15.769912	NaN

		YR2019	YR2020
economy series			
AFG	AG.LND.FRST.ZS	1.850994	1.850994
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	97.700000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
AUS	AG.LND.FRST.ZS	17.421315	17.421315
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
CAN	AG.LND.FRST.ZS	38.699637	38.695513
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
CHN	AG.LND.FRST.ZS	23.141166	23.340596
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
CRI	AG.LND.FRST.ZS	59.116725	59.437329
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	99.710000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
IND	AG.LND.FRST.ZS	24.180628	24.270228
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	97.815285	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
JPN	AG.LND.FRST.ZS	68.408779	68.408779
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
THA	AG.LND.FRST.ZS	38.969250	38.898784
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	99.900000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN
USA	AG.LND.FRST.ZS	33.866926	33.866926
	AG.LND.IRIG.AG.ZS	NaN	NaN
	EG.ELC.ACCS.ZS	100.000000	NaN
	ER.H2O.FWTL.ZS	NaN	NaN

In [4]:

```
#Data Frame-2
#Agricultural and Rural Development indicators
indicator_ids1 = ['AG.LND.ARBL.ZS', 'AG.LND.AGRI.ZS', 'AG.LND.FRST.ZS', 'AG.LND.CROP.ZS']
country_codes = ['USA', 'CAN', 'IND', 'JPN', 'CHN', 'THA', 'CRI', 'BAN', 'AUS', 'AFG']
my_dataframe2 = wb.data.DataFrame(indicator_ids1, country_codes, mrv=10)
print(my_dataframe2)
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		YR2011	YR2012	YR2013	YR2014	YR2015 \
economy	series					
AFG	AG.LND.AGRI.ZS	58.067580	58.067580	58.067580	58.067580	58.067580
	AG.LND.ARBL.ZS	11.933646	11.932114	11.924455	11.903011	11.893821
	AG.LND.CROP.ZS	0.182275	0.183807	0.191465	0.212909	0.222100
	AG.LND.FRST.ZS	1.850994	1.850994	1.850994	1.850994	1.850994
AUS	AG.LND.AGRI.ZS	51.562813	50.385431	48.393710	48.738399	45.314424
	AG.LND.ARBL.ZS	4.125705	4.118220	4.063991	4.160251	4.047056
	AG.LND.CROP.ZS	0.049868	0.046288	0.050792	0.045130	0.039871
	AG.LND.FRST.ZS	16.955310	17.047689	17.140067	17.232446	17.324825
CAN	AG.LND.AGRI.ZS	6.435684	6.443636	6.451555	6.459475	6.467505
	AG.LND.ARBL.ZS	4.158678	4.186451	4.214223	4.241996	4.269881
	AG.LND.CROP.ZS	0.019932	0.019965	0.019965	0.019965	0.019965
	AG.LND.FRST.ZS	38.734864	38.730258	38.725651	38.721045	38.716438
CHN	AG.LND.AGRI.ZS	56.116911	56.107828	56.108038	56.096738	56.090201
	AG.LND.ARBL.ZS	12.800417	12.728202	12.707193	12.695893	12.689357
	AG.LND.CROP.ZS	1.635171	1.698303	1.719524	1.719524	1.719524
	AG.LND.FRST.ZS	21.491096	21.696596	21.902095	22.107595	22.313094
CRI	AG.LND.AGRI.ZS	35.624755	35.487662	35.272229	34.821778	34.811986
	AG.LND.ARBL.ZS	4.543674	4.543674	4.543674	4.406580	4.964747
	AG.LND.CROP.ZS	6.169213	6.169213	6.247552	6.247552	5.973365
	AG.LND.FRST.ZS	56.552409	56.872934	57.193459	57.513984	57.834508
IND	AG.LND.AGRI.ZS	60.430043	60.420626	60.439461	60.447196	60.431389
	AG.LND.ARBL.ZS	52.798173	52.652538	52.617559	52.624622	52.608814
	AG.LND.CROP.ZS	4.167241	4.305140	4.372408	4.372408	4.372408
	AG.LND.FRST.ZS	23.463822	23.553422	23.643023	23.732624	23.822225
JPN	AG.LND.AGRI.ZS	12.513032	12.480110	12.449931	12.395062	12.334705
	AG.LND.ARBL.ZS	11.670782	11.648834	11.626886	11.582990	11.536351
	AG.LND.CROP.ZS	0.842250	0.831276	0.823045	0.812071	0.798354
	AG.LND.FRST.ZS	68.481756	68.469684	68.457613	68.445542	68.433471
THA	AG.LND.AGRI.ZS	41.222181	42.788076	43.277418	43.277418	43.277418
	AG.LND.ARBL.ZS	30.848128	32.414023	32.903365	32.903365	32.903365
	AG.LND.CROP.ZS	8.808158	8.808158	8.808158	8.808158	8.808158
	AG.LND.FRST.ZS	39.285560	39.280863	39.276165	39.271467	39.266770
USA	AG.LND.AGRI.ZS	44.278758	44.064999	44.124682	44.184360	44.244027
	AG.LND.ARBL.ZS	17.093563	16.945488	17.005166	17.064845	17.124512
	AG.LND.CROP.ZS	0.295165	0.295165	0.295165	0.295165	0.295165
	AG.LND.FRST.ZS	33.779470	33.809533	33.839596	33.869659	33.899723
		YR2016	YR2017	YR2018	YR2019	YR2020
economy	series					
AFG	AG.LND.AGRI.ZS	58.067580	58.067580	58.081365	NaN	NaN
	AG.LND.ARBL.ZS	11.838679	11.792727	11.798854	NaN	NaN
	AG.LND.CROP.ZS	0.277242	0.323193	0.330852	NaN	NaN
	AG.LND.FRST.ZS	1.850994	1.850994	1.850994	1.850994	1.850994
AUS	AG.LND.AGRI.ZS	44.539926	48.340618	46.658095	NaN	NaN
	AG.LND.ARBL.ZS	3.907556	3.997910	4.026771	NaN	NaN
	AG.LND.CROP.ZS	0.041732	0.041862	0.043162	NaN	NaN
	AG.LND.FRST.ZS	17.425488	17.422914	17.421315	17.421315	17.421315
CAN	AG.LND.AGRI.ZS	6.474755	6.471855	6.491374	NaN	NaN
	AG.LND.ARBL.ZS	4.297542	4.295200	4.315053	NaN	NaN
	AG.LND.CROP.ZS	0.019854	0.019296	0.018961	NaN	NaN
	AG.LND.FRST.ZS	38.712013	38.707888	38.703763	38.699637	38.695513
CHN	AG.LND.AGRI.ZS	56.081665	56.079351	56.079083	NaN	NaN
	AG.LND.ARBL.ZS	12.680817	12.678503	12.678246	NaN	NaN
	AG.LND.CROP.ZS	1.719524	1.719524	1.719524	NaN	NaN

	AG.LND.FRST.ZS	22.542877	22.742310	22.941735	23.141166	23.340596
CRI	AG.LND.AGRI.ZS	34.811986	34.851156	34.909910	NaN	NaN
	AG.LND.ARBL.ZS	5.238935	4.925578	4.945163	NaN	NaN
	AG.LND.CROP.ZS	6.071289	6.423815	6.462985	NaN	NaN
	AG.LND.FRST.ZS	58.155112	58.475715	58.796122	59.116725	59.437329
IND	AG.LND.AGRI.ZS	60.431389	60.431389	60.431389	NaN	NaN
	AG.LND.ARBL.ZS	52.608814	52.608814	52.608814	NaN	NaN
	AG.LND.CROP.ZS	4.372408	4.372408	4.372408	NaN	NaN
	AG.LND.FRST.ZS	23.911825	24.001426	24.091027	24.180628	24.270228
JPN	AG.LND.AGRI.ZS	12.266118	12.192044	12.126200	NaN	NaN
	AG.LND.ARBL.ZS	11.478738	11.415638	11.363512	NaN	NaN
	AG.LND.CROP.ZS	0.787380	0.776406	0.762689	NaN	NaN
	AG.LND.FRST.ZS	68.422497	68.408779	68.408779	68.408779	68.408779
THA	AG.LND.AGRI.ZS	43.277418	43.277418	43.277418	NaN	NaN
	AG.LND.ARBL.ZS	32.903365	32.903365	32.903365	NaN	NaN
	AG.LND.CROP.ZS	8.808158	8.808158	8.808158	NaN	NaN
	AG.LND.FRST.ZS	39.180646	39.110180	39.039715	38.969250	38.898784
USA	AG.LND.AGRI.ZS	44.303705	44.363367	44.363367	NaN	NaN
	AG.LND.ARBL.ZS	17.184190	17.243857	17.243857	NaN	NaN
	AG.LND.CROP.ZS	0.295165	0.295165	0.295165	NaN	NaN
	AG.LND.FRST.ZS	33.899723	33.866926	33.866926	33.866926	33.866926

In [5]:

```
#Computing statistics
#Transforming the data
my_dataframe1.groupby('economy').transform(lambda x : x.median())
```

Out[5]:

		YR2011	YR2012	YR2013	YR2014	YR2015	YR2016	Y
economy	series							
AFG	AG.LND.FRST.ZS	5.391717	24.240742	5.518333	5.742548	5.710894	6.481140	24.!
	AG.LND.IRIG.AG.ZS	5.391717	24.240742	5.518333	5.742548	5.710894	6.481140	24.!
	EG.ELC.ACCS.ZS	5.391717	24.240742	5.518333	5.742548	5.710894	6.481140	24.!
	ER.H2O.FWTL.ZS	5.391717	24.240742	5.518333	5.742548	5.710894	6.481140	24.!
AUS	AG.LND.FRST.ZS	16.955310	10.090782	17.140067	17.232446	17.324825	17.425488	10.!
	AG.LND.IRIG.AG.ZS	16.955310	10.090782	17.140067	17.232446	17.324825	17.425488	10.!
	EG.ELC.ACCS.ZS	16.955310	10.090782	17.140067	17.232446	17.324825	17.425488	10.!
	ER.H2O.FWTL.ZS	16.955310	10.090782	17.140067	17.232446	17.324825	17.425488	10.!
CAN	AG.LND.FRST.ZS	69.367432	38.730258	69.362826	69.360522	38.716438	69.356007	38.!
	AG.LND.IRIG.AG.ZS	69.367432	38.730258	69.362826	69.360522	38.716438	69.356007	38.!
	EG.ELC.ACCS.ZS	69.367432	38.730258	69.362826	69.360522	38.716438	69.356007	38.!
	ER.H2O.FWTL.ZS	69.367432	38.730258	69.362826	69.360522	38.716438	69.356007	38.!
CHN	AG.LND.FRST.ZS	21.491096	21.696596	60.949552	61.053797	61.156547	61.271439	22.!
	AG.LND.IRIG.AG.ZS	21.491096	21.696596	60.949552	61.053797	61.156547	61.271439	22.!
	EG.ELC.ACCS.ZS	21.491096	21.696596	60.949552	61.053797	61.156547	61.271439	22.!
	ER.H2O.FWTL.ZS	21.491096	21.696596	60.949552	61.053797	61.156547	61.271439	22.!
CRI	AG.LND.FRST.ZS	77.892378	56.872934	78.378488	78.436637	78.622156	78.827556	58.!
	AG.LND.IRIG.AG.ZS	77.892378	56.872934	78.378488	78.436637	78.622156	78.827556	58.!
	EG.ELC.ACCS.ZS	77.892378	56.872934	78.378488	78.436637	78.622156	78.827556	58.!

		YR2011	YR2012	YR2013	YR2014	YR2015	YR2016	Y
economy	series							
IND	ER.H2O.FWTL.ZS	77.892378	56.872934	78.378488	78.436637	78.622156	78.827556	58.4
	AG.LND.FRST.ZS	35.398787	40.554089	36.785607	37.900969	38.059486	56.723353	44.7
	AG.LND.IRIG.AG.ZS	35.398787	40.554089	36.785607	37.900969	38.059486	56.723353	44.7
	EG.ELC.ACCS.ZS	35.398787	40.554089	36.785607	37.900969	38.059486	56.723353	44.7
JPN	ER.H2O.FWTL.ZS	35.398787	40.554089	36.785607	37.900969	38.059486	56.723353	44.7
	AG.LND.FRST.ZS	68.481756	51.590305	68.457613	68.445542	84.216735	84.211248	68.4
	AG.LND.IRIG.AG.ZS	68.481756	51.590305	68.457613	68.445542	84.216735	84.211248	68.4
	EG.ELC.ACCS.ZS	68.481756	51.590305	68.457613	68.445542	84.216735	84.211248	68.4
THA	ER.H2O.FWTL.ZS	68.481756	51.590305	68.457613	68.445542	84.216735	84.211248	68.4
	AG.LND.FRST.ZS	69.322422	39.280863	69.351614	69.377555	69.433385	69.516386	39.7
	AG.LND.IRIG.AG.ZS	69.322422	39.280863	69.351614	69.377555	69.433385	69.516386	39.7
	EG.ELC.ACCS.ZS	69.322422	39.280863	69.351614	69.377555	69.433385	69.516386	39.7
USA	ER.H2O.FWTL.ZS	69.322422	39.280863	69.351614	69.377555	69.433385	69.516386	39.7
	AG.LND.FRST.ZS	66.889735	24.517230	66.919798	66.934830	66.949861	66.949861	24.8
	AG.LND.IRIG.AG.ZS	66.889735	24.517230	66.919798	66.934830	66.949861	66.949861	24.8
	EG.ELC.ACCS.ZS	66.889735	24.517230	66.919798	66.934830	66.949861	66.949861	24.8
	ER.H2O.FWTL.ZS	66.889735	24.517230	66.919798	66.934830	66.949861	66.949861	24.8

In [6]:

```
#Apply function to have idea about the range of values
my_dataframe1.groupby('economy').apply(lambda x: x.astype(float).max() - x.min())
```

Out[6]:

	YR2011	YR2012	YR2013	YR2014	YR2015	YR2016	YR2017	YR2018
economy								
AFG	41.371025	67.249006	67.131947	87.649006	69.649006	95.849006	95.849006	96.864628
AUS	99.504444	99.446879	99.360366	99.369350	99.382654	99.373325	99.396510	99.359701
CAN	61.265136	98.721404	61.274349	61.278955	98.440114	61.287987	98.746316	61.296237
CHN	89.650470	78.677992	78.094914	77.892405	77.686906	77.457123	78.961962	77.058265
CRI	42.679939	97.433386	42.370058	41.845307	41.575296	41.344888	96.773451	40.903878
IND	44.136178	56.346578	57.095022	59.852589	64.177775	65.623056	68.455407	71.102272
JPN	65.490024	81.111163	64.808286	65.183710	31.566529	31.577503	81.111163	31.591221
THA	60.073723	73.582120	60.150898	60.212175	60.333230	60.671482	74.373497	60.780285
USA	66.220530	94.395668	66.160404	66.130341	66.100277	66.100277	94.214483	66.133074

In [7]:

```
x = my_dataframe1.groupby('economy')['YR2016'].median()
y = my_dataframe1.groupby('economy')['YR2017'].median()
```

```
z = my_dataframe1.groupby('economy')['YR2018'].median()
print(x,y,z)
```

```
economy
AFG      6.481140
AUS     17.425488
CAN     69.356007
CHN     61.271439
CRI     78.827556
IND     56.723353
JPN     84.211248
THA     69.516386
USA     66.949861
Name: YR2016, dtype: float64 economy
AFG     24.503205
AUS     10.331884
CAN     38.707888
CHN     22.742310
CRI     58.475715
IND     44.778700
JPN     68.408779
THA     39.110180
USA     24.818419
Name: YR2017, dtype: float64 economy
AFG      5.134629
AUS     17.421315
CAN     69.351881
CHN     61.470868
CRI     79.248061
IND     59.642163
JPN     84.204390
THA     69.429858
USA     66.933463
Name: YR2018, dtype: float64
```

```
In [8]: #Filtering the data
my_dataframe2.groupby('economy').filter(lambda x : x['YR2014'].median() > 25)
```

```
Out[8]:
```

		YR2011	YR2012	YR2013	YR2014	YR2015	YR2016	YR2017
economy	series							
IND	AG.LND.AGRI.ZS	60.430043	60.420626	60.439461	60.447196	60.431389	60.431389	60.431389
	AG.LND.ARBL.ZS	52.798173	52.652538	52.617559	52.624622	52.608814	52.608814	52.608814
	AG.LND.CROP.ZS	4.167241	4.305140	4.372408	4.372408	4.372408	4.372408	4.372408
	AG.LND.FRST.ZS	23.463822	23.553422	23.643023	23.732624	23.822225	23.911825	24.001825
THA	AG.LND.AGRI.ZS	41.222181	42.788076	43.277418	43.277418	43.277418	43.277418	43.277418
	AG.LND.ARBL.ZS	30.848128	32.414023	32.903365	32.903365	32.903365	32.903365	32.903365
	AG.LND.CROP.ZS	8.808158	8.808158	8.808158	8.808158	8.808158	8.808158	8.808158
	AG.LND.FRST.ZS	39.285560	39.280863	39.276165	39.271467	39.266770	39.180646	39.110646
USA	AG.LND.AGRI.ZS	44.278758	44.064999	44.124682	44.184360	44.244027	44.303705	44.363705
	AG.LND.ARBL.ZS	17.093563	16.945488	17.005166	17.064845	17.124512	17.184190	17.243190
	AG.LND.CROP.ZS	0.295165	0.295165	0.295165	0.295165	0.295165	0.295165	0.295165
	AG.LND.FRST.ZS	33.779470	33.809533	33.839596	33.869659	33.899723	33.899723	33.869723

In [9]:

```
#Pivot Table
import numpy as np
x = my_dataframe1.pivot_table(index=['series'],aggfunc={'YR2015':np.mean})
y = my_dataframe2.pivot_table(index=['series'],aggfunc={'YR2015':np.mean})
print(x,y)
```

	YR2015	
series		
AG.LND.FRST.ZS	33.718005	
AG.LND.IRIG.AG.ZS	11.486903	
EG.ELC.ACCS.ZS	95.389978	YR2015
series		
AG.LND.AGRI.ZS	40.115471	
AG.LND.ARBL.ZS	16.893100	
AG.LND.CROP.ZS	2.472101	
AG.LND.FRST.ZS	33.718005	

STORY OF THE DATA

The primary objective of this assignment is to load the world bank data directly into Python IDE whereas the secondary objective is to use different python operations or functions that will help to compute the results or values. We have used four indicators each from climate change category and Agricultural and Rural development category. The time frame from which we have collected data is from 2011 to 2020. I have selected 10 countries. These 10 countries are Afghansitan, Australia, Canada, China, Costa Rica, India, Japan, Thailand, USA, and Bangladesh I have performed filtering based on the median condition. I have created pivot table to get the mean values of year 2015 for both of the data frames. I have used apply function to measure the spread of values of each year present in the data. I have also performed data transformation in terms of median. Similarly, groupby function alone has also been used to compute the median values of year 2016 to 2018 of data frame 1.