

1 - Population pyramids

November 29, 2023

In this notebook I am going to load and clean the census data. I will remove any area which are not at the level of a Local Authority and indicate which region each Local Authority is in.

I will then reorganise the data into a form in which it is possible to produce population pyramids for each region in order to address the question: how much do the age profiles of different regions of England and Wales differ?

```
[1]: #Import statements
import pandas as pd
import re
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[2]: #Generate pandas dataframe from excel sheet
census_demographic_df = pd.read_excel('Data/census2021firstresultsenglandwales1.
↳xlsx', sheet_name='P03', skiprows=lambda x: x in [0,1,2,3,4,5,6])
census_demographic_df.head()
```

```
[2]:
```

	Area code [note 2]	Area name	All persons \
0	K04000001	England and Wales	59597300
1	E92000001	England	56489800
2	E12000001	North East	2647100
3	E06000047	County Durham	522100
4	E06000005	Darlington	107800

```
Females:\nAged 4 years and under\n[note 12] \
```

0	1577300
1	1501600
2	65600
3	12000
4	2700

```
Females:\nAged 5 to 9 years\n[note 12] \
```

0	1720400
1	1634400
2	73000
3	13800
4	3000

	Females:\nAged 10 to 14 years\n[note 12]	\
0	1753100	
1	1664100	
2	75300	
3	14400	
4	3300	

	Females:\nAged 15 to 19 years\n[note 12]	\
0	1653100	
1	1568100	
2	73600	
3	15300	
4	2800	

	Females:\nAged 20 to 24 years\n[note 12]	\
0	1793300	
1	1701500	
2	80500	
3	16400	
4	2800	

	Females:\nAged 25 to 29 years\n[note 12]	\
0	1991900	
1	1897400	
2	81800	
3	15000	
4	3300	

	Females:\nAged 30 to 34 years\n[note 12]	...	\
0	2145200	...	
1	2044400	...	
2	87100	...	
3	16000	...	
4	3600	...	

	Males:\nAged 45 to 49 years\n[note 12]	\
0	1867900	
1	1777100	
2	77800	
3	15400	
4	3300	

	Males:\nAged 50 to 54 years\n[note 12]	\
0	2027500	
1	1922800	
2	89600	

3	18600
4	3800
Males:\nAged 55 to 59 years\n[note 12] \	
0	1978300
1	1869600
2	94100
3	19500
4	3900
Males:\nAged 60 to 64 years\n[note 12] \	
0	1699600
1	1602000
2	85600
3	17200
4	3400
Males:\nAged 65 to 69 years\n[note 12] \	
0	1428200
1	1341900
2	74200
3	15400
4	3000
Males:\nAged 70 to 74 years\n[note 12] \	
0	1419500
1	1331800
2	71100
3	14800
4	2800
Males:\nAged 75 to 79 years\n[note 12] \	
0	1008900
1	947200
2	47300
3	10300
4	2000
Males:\nAged 80 to 84 years\n[note 12] \	
0	668200
1	628600
2	31800
3	6700
4	1400
Males:\nAged 85 to 89 years\n[note 12] \	
0	372700

1	351400
2	17100
3	3400
4	800

Males:\nAged 90 years and over\n[note 12]

0	169200
1	159800
2	7100
3	1400
4	300

[5 rows x 41 columns]

```
[3]: #Use regexes to clean column names
pattern = '\\n'
census_demographic_df.rename(columns=(lambda x : re.sub(pattern, '', x)),
    inplace=True)
pattern2 = '\\[note [1-9]+\\]'
census_demographic_df.rename(columns=(lambda x : re.sub(pattern2, '', x)),
    inplace=True)
census_demographic_df
```

```
[3]:
```

	Area code	Area name	All persons \
0	K04000001	England and Wales	59597300
1	E92000001	England	56489800
2	E12000001	North East	2647100
3	E06000047	County Durham	522100
4	E06000005	Darlington	107800
..
370	W06000018	Caerphilly	175900
371	W06000019	Blaenau Gwent	66900
372	W06000020	Torfaen	92300
373	W06000021	Monmouthshire	93000
374	W06000022	Newport	159600

	Females:Aged 4 years and under	Females:Aged 5 to 9 years \
0	1577300	1720400
1	1501600	1634400
2	65600	73000
3	12000	13800
4	2700	3000
..
370	4600	5100
371	1700	1800
372	2400	2600
373	1900	2200

374	4800	5000
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	Females:Aged 10 to 14 years	Females:Aged 15 to 19 years \
0	1753100	1653100
1	1664100	1568100
2	75300	73600
3	14400	15300
4	3300	2800
..
370	5300	4500
371	1900	1600
372	2700	2300
373	2500	2200
374	5000	4200

	Females:Aged 20 to 24 years	Females:Aged 25 to 29 years \
0	1793300	1991900
1	1701500	1897400
2	80500	81800
3	16400	15000
4	2800	3300
..
370	4400	5600
371	1800	2300
372	2300	3000
373	1800	2200
374	4300	5700

	Females:Aged 30 to 34 years	... Males:Aged 45 to 49 years \
0	2145200	... 1867900
1	2044400	... 1777100
2	87100	... 77800
3	16000	... 15400
4	3600	... 3300
..
370	6000	... 5400
371	2400	... 2100
372	3300	... 2600
373	2400	... 2800
374	6500	... 4800

	Males:Aged 50 to 54 years	Males:Aged 55 to 59 years \
0	2027500	1978300
1	1922800	1869600
2	89600	94100
3	18600	19500
4	3800	3900

..
370	6200	6300
371	2400	2600
372	3100	3300
373	3500	3800
374	5300	5200

	Males:Aged 60 to 64 years	Males:Aged 65 to 69 years	\
0	1699600	1428200	
1	1602000	1341900	
2	85600	74200	
3	17200	15400	
4	3400	3000	
..	
370	5400	4700	
371	2100	1900	
372	2900	2500	
373	3300	3000	
374	4500	3600	

	Males:Aged 70 to 74 years	Males:Aged 75 to 79 years	\
0	1419500	1008900	
1	1331800	947200	
2	71100	47300	
3	14800	10300	
4	2800	2000	
..	
370	4800	3300	
371	1900	1300	
372	2500	1800	
373	3100	2300	
374	3500	2500	

	Males:Aged 80 to 84 years	Males:Aged 85 to 89 years	\
0	668200	372700	
1	628600	351400	
2	31800	17100	
3	6700	3400	
4	1400	800	
..	
370	2100	1000	
371	800	400	
372	1100	600	
373	1500	900	
374	1700	900	

Males:Aged 90 years and over

0	169200
1	159800
2	7100
3	1400
4	300
..	...
370	400
371	100
372	300
373	400
374	400

[375 rows x 41 columns]

```
[4]: #Delete whitespace in 'Area code' column name
census_demographic_df=census_demographic_df.rename(columns = {'Area code ':
↳'Area code'})
```

In order to keep only a list of Local Authorities and remove the other categories which aggregate the Local Authorities (LAs) (so removing the Metropolitan Counties (E11), Counties (E10), Regions (E12), Inner and Outer London (E13), and countries) I shall keep only those rows where the Area Code is E06-E09 or W06 (the Welsh LAs). Before I do this I am going to extract the regions and Wales into a separate dataframe to use it to check totals later.

```
[5]: #Keep only the regions of England and Wales as a whole and put in regions_
↳dataframe
patternKeep = "^E12|W92"
keep = census_demographic_df['Area code'].str.contains(patternKeep)
regions_demographic_df = census_demographic_df[keep]

#Keep only the local authorities of England and Wales
patternKeep = "^E0[6-9]|W06"
keep = census_demographic_df['Area code'].str.contains(patternKeep)
census_demographic_df = census_demographic_df[keep]
```

```
[6]: #Use regex to remove notes from Local Authority names
pattern_note = '\s\[note\s[0-9]+\]'
census_demographic_df['Area name'] = [re.sub(pattern_note, '', str(x)) for x in_
↳census_demographic_df['Area name']]
```

I would like to indicate which region each Local Authority is in as I want to look at the data at both a regional and Local Authority level. For Wales there doesn't appear to be any regions but as the overall population of Wales (approximately 3 million) is on a similar par to the regions I am going to group the Welsh authorities together and indicate they are in Wales.

```
[7]: #Reset index
census_demographic_df.reset_index(level=None, inplace=True)
census_demographic_df.drop ('index', axis=1, inplace = True)
```

```
#Include a column with the region each local authority is in
census_demographic_df.loc[range(0,12),'Region'] = 'North East'
census_demographic_df.loc[range(12,51),'Region'] = 'North West'
census_demographic_df.loc[range(51,72),'Region'] = 'Yorkshire and the Humber'
census_demographic_df.loc[range(72,107),'Region'] = 'East Midlands'
census_demographic_df.loc[range(107,137),'Region'] = 'West Midlands'
census_demographic_df.loc[range(137,182),'Region'] = 'East of England'
census_demographic_df.loc[range(182,215),'Region'] = 'London'
census_demographic_df.loc[range(215,279),'Region'] = 'South East'
census_demographic_df.loc[range(279,309),'Region'] = 'South West'
census_demographic_df.loc[range(309,331),'Region'] = 'Wales'
```

Now to check if I have correctly assigned all the regions/Wales. I will do this by grouping according to the region column and then summing the 'All persons' column and manually checking this against the totals on the excel sheet.

```
[8]: census_demographic_df.to_csv('cleaned_census_df.csv')
```

```
[9]: #Groupby region and then sum the columns
regions_sum_df = census_demographic_df.groupby(['Region']).sum(numeric_only =
↳ True)
regions_sum_df
```

```
[9]:
```

	All persons	Females:Aged 4 years and under \
Region		
East Midlands	4879800	123400
East of England	6334500	170000
London	8800000	258900
North East	2647100	65600
North West	7417300	199000
South East	9278400	241600
South West	5701300	133800
Wales	3107700	75800
West Midlands	5950600	163700
Yorkshire and the Humber	5480800	145800

	Females:Aged 5 to 9 years \
Region	
East Midlands	138800
East of England	185900
London	260200
North East	72900
North West	217600
South East	269200
South West	151800
Wales	86100
West Midlands	179600

Yorkshire and the Humber	159500
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Females:Aged 10 to 14 years \

Region	
East Midlands	141500
East of England	187700
London	262000
North East	75300
North West	219700
South East	277200
South West	156400
Wales	88900
West Midlands	182300
Yorkshire and the Humber	162000

Females:Aged 15 to 19 years \

Region	
East Midlands	140000
East of England	165900
London	240300
North East	73500
North West	211100
South East	253600
South West	152900
Wales	85200
West Midlands	173600
Yorkshire and the Humber	157700

Females:Aged 20 to 24 years \

Region	
East Midlands	151400
East of England	167300
London	304700
North East	80500
North West	227400
South East	254100
South West	162700
Wales	91700
West Midlands	178800
Yorkshire and the Humber	174600

Females:Aged 25 to 29 years \

Region	
East Midlands	152400
East of England	196800
London	412200
North East	82000

North West	243100
South East	277200
South West	166900
Wales	94500
West Midlands	189200
Yorkshire and the Humber	178300

Females:Aged 30 to 34 years \

Region	
East Midlands	163600
East of England	219300
London	422500
North East	86900
North West	261000
South East	312900
South West	182400
Wales	101000
West Midlands	206400
Yorkshire and the Humber	189600

Females:Aged 35 to 39 years \

Region	
East Midlands	157400
East of England	215800
London	382700
North East	84300
North West	248600
South East	314000
South West	175200
Wales	95900
West Midlands	197000
Yorkshire and the Humber	178900

Females:Aged 40 to 44 years ... \

Region		...
East Midlands	147900	...
East of England	206900	...
London	343000	...
North East	77900	...
North West	226200	...
South East	309300	...
South West	168200	...
Wales	89600	...
West Midlands	183800	...
Yorkshire and the Humber	164100	...

Males:Aged 45 to 49 years \

Region	
East Midlands	152100
East of England	202500
London	292800
North East	77600
North West	227000
South East	302200
South West	171000
Wales	91000
West Midlands	184400
Yorkshire and the Humber	167200

Males:Aged 50 to 54 years \

Region	
East Midlands	171200
East of England	218300
London	280000
North East	89500
North West	254500
South East	322800
South West	194400
Wales	104700
West Midlands	205100
Yorkshire and the Humber	187700

Males:Aged 55 to 59 years \

Region	
East Midlands	167800
East of England	212800
London	247800
North East	94100
North West	252800
South East	315800
South West	199400
Wales	108600
West Midlands	195100
Yorkshire and the Humber	184500

Males:Aged 60 to 64 years \

Region	
East Midlands	144600
East of England	182500
London	197000
North East	85600
North West	219000
South East	266700
South West	176700

Wales	97600
West Midlands	168300
Yorkshire and the Humber	161600

Males:Aged 65 to 69 years \

Region	
East Midlands	123500
East of England	154200
London	145600
North East	74200
North West	184500
South East	223300
South West	155700
Wales	86400
West Midlands	143700
Yorkshire and the Humber	137700

Males:Aged 70 to 74 years \

Region	
East Midlands	125800
East of England	159600
London	124500
North East	71100
North West	182600
South East	227900
South West	164100
Wales	87800
West Midlands	141000
Yorkshire and the Humber	136300

Males:Aged 75 to 79 years \

Region	
East Midlands	90300
East of England	115100
London	83900
North East	47400
North West	125500
South East	165200
South West	120700
Wales	61700
West Midlands	105300
Yorkshire and the Humber	93800

Males:Aged 80 to 84 years \

Region	
East Midlands	57000
East of England	76700

London	60000
North East	32100
North West	83000
South East	110100
South West	78400
Wales	39600
West Midlands	69600
Yorkshire and the Humber	62300

Males:Aged 85 to 89 years \

Region	
East Midlands	31300
East of England	44300
London	34700
North East	17100
North West	44000
South East	63700
South West	44700
Wales	21500
West Midlands	37700
Yorkshire and the Humber	34300

Males:Aged 90 years and over

Region	
East Midlands	13500
East of England	20600
London	16000
North East	7100
North West	18600
South East	30700
South West	21600
Wales	9600
West Midlands	16800
Yorkshire and the Humber	14400

[10 rows x 39 columns]

Now I will produce a dataframe which only includes the data for females and the total population for each region in order to get a dataframe in the right format to produce a population pyramid.

```
[10]: #Create list of columns with data on females only
females_cols = [col for col in regions_sum_df.columns if 'Females' in col]

#Append 'All persons' column
females_cols.append('All persons')

#Create new database with only female columns and 'All persons' column
```

```
female_regions_census_df = regions_sum_df[females_cols]
female_regions_census_df.head()
```

```
[10]:
```

	Females:Aged 4 years and under	Females:Aged 5 to 9 years \
Region		
East Midlands	123400	138800
East of England	170000	185900
London	258900	260200
North East	65600	72900
North West	199000	217600

	Females:Aged 10 to 14 years	Females:Aged 15 to 19 years \
Region		
East Midlands	141500	140000
East of England	187700	165900
London	262000	240300
North East	75300	73500
North West	219700	211100

	Females:Aged 20 to 24 years	Females:Aged 25 to 29 years \
Region		
East Midlands	151400	152400
East of England	167300	196800
London	304700	412200
North East	80500	82000
North West	227400	243100

	Females:Aged 30 to 34 years	Females:Aged 35 to 39 years \
Region		
East Midlands	163600	157400
East of England	219300	215800
London	422500	382700
North East	86900	84300
North West	261000	248600

	Females:Aged 40 to 44 years	Females:Aged 45 to 49 years \
Region		
East Midlands	147900	156400
East of England	206900	208500
London	343000	301400
North East	77900	81200
North West	226200	231400

	Females:Aged 50 to 54 years	Females:Aged 55 to 59 years \
Region		
East Midlands	175400	172300
East of England	224600	219900

London	293800	263700
North East	93700	98400
North West	259300	260100

	Females:Aged 60 to 64 years	Females:Aged 65 to 69 years \
Region		
East Midlands	147700	130200
East of England	187000	165500
London	207700	160700
North East	89600	77500
North West	224800	191700

	Females:Aged 70 to 74 years	Females:Aged 75 to 79 years \
Region		
East Midlands	135400	100900
East of England	176800	131000
London	144600	105500
North East	77300	53900
North West	197200	144400

	Females:Aged 80 to 84 years	Females:Aged 85 to 89 years \
Region		
East Midlands	70400	45200
East of England	95000	63700
London	81200	52300
North East	41700	26300
North West	107500	67100

	Females:Aged 90 years and over	All persons
Region		
East Midlands	28800	4879800
East of England	42100	6334500
London	34500	8800000
North East	15500	2647100
North West	40800	7417300

```
[11]: #Create new dataframe to store percentages of females in each age group by
      ↪region
female_regions_proportions_df = pd.DataFrame()
for i in females_cols:
    female_regions_proportions_df['%' + i] = female_regions_census_df [i]/
    ↪female_regions_census_df ['All persons']*100

#Create list of columns of female percentages to sum
cols_to_sum = female_regions_proportions_df.columns[ :
    ↪female_regions_proportions_df.shape[1]-1]
```

```

#Create new column with sum of female percentages
female_regions_proportions_df['Total'] =
    ↳female_regions_proportions_df[cols_to_sum].sum(axis=1)

#Remove '%All persons' and 'Total' columns
female_regions_proportions_minus_totals_df = female_regions_proportions_df.
    ↳copy()
female_regions_proportions_minus_totals_df .drop(['%All persons', 'Total'],
    ↳axis=1, inplace = True)

#Remove % Females: from column names
female_regions_proportions_minus_totals_df.columns =
    ↳female_regions_proportions_minus_totals_df.columns.map(lambda x: x.
    ↳removeprefix("%Females:"))

female_regions_proportions_df.head()

```

```

[11]:          %Females:Aged 4 years and under  %Females:Aged 5 to 9 years  \
Region
East Midlands                2.528792                2.844379
East of England              2.683716                2.934723
London                      2.942045                2.956818
North East                  2.478184                2.753957
North West                  2.682917                2.933682

          %Females:Aged 10 to 14 years  %Females:Aged 15 to 19 years  \
Region
East Midlands                2.899709                2.868970
East of England              2.963138                2.618991
London                      2.977273                2.730682
North East                  2.844622                2.776623
North West                  2.961994                2.846049

          %Females:Aged 20 to 24 years  %Females:Aged 25 to 29 years  \
Region
East Midlands                3.102586                3.123079
East of England              2.641092                3.106796
London                      3.462500                4.684091
North East                  3.041064                3.097730
North West                  3.065806                3.277473

          %Females:Aged 30 to 34 years  %Females:Aged 35 to 39 years  \
Region

```


East Midlands	3.352596	3.225542
East of England	3.461994	3.406741
London	4.801136	4.348864
North East	3.282838	3.184617
North West	3.518801	3.351624

	%Females:Aged 40 to 44 years	%Females:Aged 45 to 49 years \
Region		
East Midlands	3.030862	3.205049
East of England	3.266240	3.291499
London	3.897727	3.425000
North East	2.942843	3.067508
North West	3.049627	3.119734

	... %Females:Aged 55 to 59 years \
Region	...
East Midlands	... 3.530882
East of England	... 3.471466
London	... 2.996591
North East	... 3.717276
North West	... 3.506667

	%Females:Aged 60 to 64 years	%Females:Aged 65 to 69 years \
Region		
East Midlands	3.026763	2.668142
East of England	2.952088	2.612677
London	2.360227	1.826136
North East	3.384836	2.927732
North West	3.030752	2.584498

	%Females:Aged 70 to 74 years	%Females:Aged 75 to 79 years \
Region		
East Midlands	2.774704	2.067708
East of England	2.791065	2.068040
London	1.643182	1.198864
North East	2.920177	2.036191
North West	2.658649	1.946800

	%Females:Aged 80 to 84 years	%Females:Aged 85 to 89 years \
Region		
East Midlands	1.442682	0.926267
East of England	1.499724	1.005604
London	0.922727	0.594318
North East	1.575309	0.993540
North West	1.449314	0.904642

%Females:Aged 90 years and over	%All persons	Total
---------------------------------	--------------	-------

Region			
East Midlands	0.590188	100.0	50.803312
East of England	0.664614	100.0	50.985871
London	0.392045	100.0	51.498864
North East	0.585546	100.0	51.150315
North West	0.550065	100.0	50.934976

[5 rows x 21 columns]

```
[12]: #Create list of columns with data on males only
males_cols = [col for col in regions_sum_df.columns if 'Males' in col]

#Append 'All persons' column
males_cols.append('All persons')

#Create new database with only male columns and 'All persons' column
male_regions_census_df = regions_sum_df[males_cols]
male_regions_census_df.head()
```

```
[12]: Males:Aged 4 years and under  Males:Aged 5 to 9 years  \
```

Region		
East Midlands	129700	144600
East of England	179000	195000
London	270000	271600
North East	68700	77700
North West	208700	227700

```
Males:Aged 10 to 14 years  Males:Aged 15 to 19 years  \
```

Region		
East Midlands	148700	147700
East of England	197400	177300
London	273700	249100
North East	79200	76400
North West	232500	219900

```
Males:Aged 20 to 24 years  Males:Aged 25 to 29 years  \
```

Region		
East Midlands	155000	149300
East of England	175300	191300
London	284800	374300
North East	82300	78700
North West	226100	233800

```
Males:Aged 30 to 34 years  Males:Aged 35 to 39 years  \
```

Region		
East Midlands	155400	150300
East of England	205100	202400

London	385800	353000
North East	80900	79600
North West	245400	236900

	Males:Aged 40 to 44 years	Males:Aged 45 to 49 years	\
Region			
East Midlands	144000	152100	
East of England	199700	202500	
London	324000	292800	
North East	74300	77600	
North West	219100	227000	

	Males:Aged 50 to 54 years	Males:Aged 55 to 59 years	\
Region			
East Midlands	171200	167800	
East of England	218300	212800	
London	280000	247800	
North East	89500	94100	
North West	254500	252800	

	Males:Aged 60 to 64 years	Males:Aged 65 to 69 years	\
Region			
East Midlands	144600	123500	
East of England	182500	154200	
London	197000	145600	
North East	85600	74200	
North West	219000	184500	

	Males:Aged 70 to 74 years	Males:Aged 75 to 79 years	\
Region			
East Midlands	125800	90300	
East of England	159600	115100	
London	124500	83900	
North East	71100	47400	
North West	182600	125500	

	Males:Aged 80 to 84 years	Males:Aged 85 to 89 years	\
Region			
East Midlands	57000	31300	
East of England	76700	44300	
London	60000	34700	
North East	32100	17100	
North West	83000	44000	

	Males:Aged 90 years and over	All persons	
Region			
East Midlands	13500	4879800	

East of England	20600	6334500
London	16000	8800000
North East	7100	2647100
North West	18600	7417300

```
[13]: #Create new dataframe to store percentages of males in each age group by region
male_regions_proportions_df = pd.DataFrame()
for i in males_cols:
    male_regions_proportions_df['%'+ i] = male_regions_census_df [i]/
    ↪ male_regions_census_df ['All persons']*100

#Create list of columns of male percentages to sum
cols_to_sum = male_regions_proportions_df.columns[ :
    ↪ male_regions_proportions_df.shape[1]-1]

#Create new column with sum of male percentages
male_regions_proportions_df['Total'] = male_regions_proportions_df[cols_to_sum].
    ↪ sum(axis=1)

#Remove '%All persons' and 'Total' columns
male_regions_proportions_minus_totals_df = male_regions_proportions_df.copy()
male_regions_proportions_minus_totals_df.drop(['%All persons', 'Total'],
    ↪ axis=1, inplace = True)

#Remove % Males: from column names
male_regions_proportions_minus_totals_df.columns =
    ↪ male_regions_proportions_minus_totals_df.columns.map(lambda x: x.
    ↪ removeprefix("%Males:"))

male_regions_proportions_df.head()
```

```
[13]: %Males:Aged 4 years and under %Males:Aged 5 to 9 years \
Region
East Midlands 2.657896 2.963236
East of England 2.825795 3.078380
London 3.068182 3.086364
North East 2.595293 2.935288
North West 2.813692 3.069850

%Males:Aged 10 to 14 years %Males:Aged 15 to 19 years \
Region
East Midlands 3.047256 3.026763
East of England 3.116268 2.798958
London 3.110227 2.830682
```

North East	2.991953	2.886177
North West	3.134564	2.964691

	%Males:Aged 20 to 24 years	%Males:Aged 25 to 29 years	\
Region			
East Midlands	3.176360	3.059552	
East of England	2.767385	3.019970	
London	3.236364	4.253409	
North East	3.109063	2.973065	
North West	3.048279	3.152090	

	%Males:Aged 30 to 34 years	%Males:Aged 35 to 39 years	\
Region			
East Midlands	3.184557	3.080044	
East of England	3.237825	3.195201	
London	4.384091	4.011364	
North East	3.056175	3.007064	
North West	3.308482	3.193885	

	%Males:Aged 40 to 44 years	%Males:Aged 45 to 49 years	...	\
Region			...	
East Midlands	2.950941	3.116931	...	
East of England	3.152577	3.196780	...	
London	3.681818	3.327273	...	
North East	2.806845	2.931510	...	
North West	2.953905	3.060413	...	

	%Males:Aged 55 to 59 years	%Males:Aged 60 to 64 years	\
Region			
East Midlands	3.438666	2.963236	
East of England	3.359381	2.881048	
London	2.815909	2.238636	
North East	3.554834	3.233727	
North West	3.408248	2.952557	

	%Males:Aged 65 to 69 years	%Males:Aged 70 to 74 years	\
Region			
East Midlands	2.530841	2.577975	
East of England	2.434288	2.519536	
London	1.654545	1.414773	
North East	2.803068	2.685958	
North West	2.487428	2.461812	

	%Males:Aged 75 to 79 years	%Males:Aged 80 to 84 years	\
Region			
East Midlands	1.850486	1.168081	
East of England	1.817034	1.210830	

London	0.953409	0.681818
North East	1.790639	1.212648
North West	1.691990	1.119006

	%Males:Aged 85 to 89 years	%Males:Aged 90 years and over \
Region		
East Midlands	0.641420	0.276651
East of England	0.699345	0.325203
London	0.394318	0.181818
North East	0.645990	0.268218
North West	0.593208	0.250765

	%All persons	Total
Region		
East Midlands	100.0	49.219230
East of England	100.0	49.082011
London	100.0	48.506818
North East	100.0	48.868573
North West	100.0	49.096032

[5 rows x 21 columns]

```
[14]: #Extract index values to list
index_list = female_regions_proportions_df.index.values.tolist()

index_list
```

```
[14]: ['East Midlands',
      'East of England',
      'London',
      'North East',
      'North West',
      'South East',
      'South West',
      'Wales',
      'West Midlands',
      'Yorkshire and the Humber']
```

```
[15]: def create_dfs(regions):
        """Creates a dictionary of empty dataframes from list of regions in_
        ↪ index_list
        with region as key"""
        dfs = {}
        for x in regions:
            dfs[x] = pd.DataFrame()
        return dfs
```

```
dfs = create_dfs(index_list)
```

```
[16]: def generate_dataframes(dfs):
        """Iterates through items in dfs dictionary to generate dataframes with
        ↪ females and males columns by age group as
        % of overall population for each region as dictionary value."""
        for key, val in dfs.items():
            dfs[key] = pd.concat([female_regions_proportions_minus_totals_df.
            ↪loc[key], male_regions_proportions_minus_totals_df.loc[key]], axis = 1)
            dfs[key] = dfs[key].reset_index()
            dfs[key].columns = ['Ages', 'Females', 'Males']
            dfs[key]['Males'] = dfs[key]['Males'] * -1
            dfs[key] = dfs[key].iloc[::-1]

        return dfs

dfs_dict = generate_dataframes(dfs)

print(dfs_dict.values())
```

		Ages	Females	Males
18	Aged 90 years and over	0.590188	-0.276651	
17	Aged 85 to 89 years	0.926267	-0.641420	
16	Aged 80 to 84 years	1.442682	-1.168081	
15	Aged 75 to 79 years	2.067708	-1.850486	
14	Aged 70 to 74 years	2.774704	-2.577975	
13	Aged 65 to 69 years	2.668142	-2.530841	
12	Aged 60 to 64 years	3.026763	-2.963236	
11	Aged 55 to 59 years	3.530882	-3.438666	
10	Aged 50 to 54 years	3.594410	-3.508341	
9	Aged 45 to 49 years	3.205049	-3.116931	
8	Aged 40 to 44 years	3.030862	-2.950941	
7	Aged 35 to 39 years	3.225542	-3.080044	
6	Aged 30 to 34 years	3.352596	-3.184557	
5	Aged 25 to 29 years	3.123079	-3.059552	
4	Aged 20 to 24 years	3.102586	-3.176360	
3	Aged 15 to 19 years	2.868970	-3.026763	
2	Aged 10 to 14 years	2.899709	-3.047256	
1	Aged 5 to 9 years	2.844379	-2.963236	
0	Aged 4 years and under	2.528792	-2.657896,	
	Females		Males	
18	Aged 90 years and over	0.664614	-0.325203	
17	Aged 85 to 89 years	1.005604	-0.699345	
16	Aged 80 to 84 years	1.499724	-1.210830	
15	Aged 75 to 79 years	2.068040	-1.817034	

14	Aged 70 to 74 years	2.791065	-2.519536	
13	Aged 65 to 69 years	2.612677	-2.434288	
12	Aged 60 to 64 years	2.952088	-2.881048	
11	Aged 55 to 59 years	3.471466	-3.359381	
10	Aged 50 to 54 years	3.545663	-3.446207	
9	Aged 45 to 49 years	3.291499	-3.196780	
8	Aged 40 to 44 years	3.266240	-3.152577	
7	Aged 35 to 39 years	3.406741	-3.195201	
6	Aged 30 to 34 years	3.461994	-3.237825	
5	Aged 25 to 29 years	3.106796	-3.019970	
4	Aged 20 to 24 years	2.641092	-2.767385	
3	Aged 15 to 19 years	2.618991	-2.798958	
2	Aged 10 to 14 years	2.963138	-3.116268	
1	Aged 5 to 9 years	2.934723	-3.078380	
0	Aged 4 years and under	2.683716	-2.825795,	Ages
	Females		Males	
18	Aged 90 years and over	0.392045	-0.181818	
17	Aged 85 to 89 years	0.594318	-0.394318	
16	Aged 80 to 84 years	0.922727	-0.681818	
15	Aged 75 to 79 years	1.198864	-0.953409	
14	Aged 70 to 74 years	1.643182	-1.414773	
13	Aged 65 to 69 years	1.826136	-1.654545	
12	Aged 60 to 64 years	2.360227	-2.238636	
11	Aged 55 to 59 years	2.996591	-2.815909	
10	Aged 50 to 54 years	3.338636	-3.181818	
9	Aged 45 to 49 years	3.425000	-3.327273	
8	Aged 40 to 44 years	3.897727	-3.681818	
7	Aged 35 to 39 years	4.348864	-4.011364	
6	Aged 30 to 34 years	4.801136	-4.384091	
5	Aged 25 to 29 years	4.684091	-4.253409	
4	Aged 20 to 24 years	3.462500	-3.236364	
3	Aged 15 to 19 years	2.730682	-2.830682	
2	Aged 10 to 14 years	2.977273	-3.110227	
1	Aged 5 to 9 years	2.956818	-3.086364	
0	Aged 4 years and under	2.942045	-3.068182,	Ages
	Females		Males	
18	Aged 90 years and over	0.585546	-0.268218	
17	Aged 85 to 89 years	0.993540	-0.645990	
16	Aged 80 to 84 years	1.575309	-1.212648	
15	Aged 75 to 79 years	2.036191	-1.790639	
14	Aged 70 to 74 years	2.920177	-2.685958	
13	Aged 65 to 69 years	2.927732	-2.803068	
12	Aged 60 to 64 years	3.384836	-3.233727	
11	Aged 55 to 59 years	3.717276	-3.554834	
10	Aged 50 to 54 years	3.539723	-3.381059	
9	Aged 45 to 49 years	3.067508	-2.931510	
8	Aged 40 to 44 years	2.942843	-2.806845	
7	Aged 35 to 39 years	3.184617	-3.007064	

6	Aged 30 to 34 years	3.282838	-3.056175	
5	Aged 25 to 29 years	3.097730	-2.973065	
4	Aged 20 to 24 years	3.041064	-3.109063	
3	Aged 15 to 19 years	2.776623	-2.886177	
2	Aged 10 to 14 years	2.844622	-2.991953	
1	Aged 5 to 9 years	2.753957	-2.935288	
0	Aged 4 years and under	2.478184	-2.595293,	Ages
	Females		Males	
18	Aged 90 years and over	0.550065	-0.250765	
17	Aged 85 to 89 years	0.904642	-0.593208	
16	Aged 80 to 84 years	1.449314	-1.119006	
15	Aged 75 to 79 years	1.946800	-1.691990	
14	Aged 70 to 74 years	2.658649	-2.461812	
13	Aged 65 to 69 years	2.584498	-2.487428	
12	Aged 60 to 64 years	3.030752	-2.952557	
11	Aged 55 to 59 years	3.506667	-3.408248	
10	Aged 50 to 54 years	3.495881	-3.431168	
9	Aged 45 to 49 years	3.119734	-3.060413	
8	Aged 40 to 44 years	3.049627	-2.953905	
7	Aged 35 to 39 years	3.351624	-3.193885	
6	Aged 30 to 34 years	3.518801	-3.308482	
5	Aged 25 to 29 years	3.277473	-3.152090	
4	Aged 20 to 24 years	3.065806	-3.048279	
3	Aged 15 to 19 years	2.846049	-2.964691	
2	Aged 10 to 14 years	2.961994	-3.134564	
1	Aged 5 to 9 years	2.933682	-3.069850	
0	Aged 4 years and under	2.682917	-2.813692,	Ages
	Females		Males	
18	Aged 90 years and over	0.696241	-0.330876	
17	Aged 85 to 89 years	1.014183	-0.686541	
16	Aged 80 to 84 years	1.501336	-1.186627	
15	Aged 75 to 79 years	2.075789	-1.780479	
14	Aged 70 to 74 years	2.741852	-2.456242	
13	Aged 65 to 69 years	2.573720	-2.406665	
12	Aged 60 to 64 years	2.969262	-2.874418	
11	Aged 55 to 59 years	3.500604	-3.403604	
10	Aged 50 to 54 years	3.612692	-3.479048	
9	Aged 45 to 49 years	3.370193	-3.257027	
8	Aged 40 to 44 years	3.333549	-3.179427	
7	Aged 35 to 39 years	3.384204	-3.166494	
6	Aged 30 to 34 years	3.372349	-3.132006	
5	Aged 25 to 29 years	2.987584	-2.897051	
4	Aged 20 to 24 years	2.738619	-2.822685	
3	Aged 15 to 19 years	2.733230	-2.912140	
2	Aged 10 to 14 years	2.987584	-3.151405	
1	Aged 5 to 9 years	2.901362	-3.054406	
0	Aged 4 years and under	2.603897	-2.742930,	Ages
	Females		Males	

18	Aged 90 years and over	0.791048	-0.378861	
17	Aged 85 to 89 years	1.127813	-0.784032	
16	Aged 80 to 84 years	1.683826	-1.375125	
15	Aged 75 to 79 years	2.383667	-2.117061	
14	Aged 70 to 74 years	3.137881	-2.878291	
13	Aged 65 to 69 years	2.948450	-2.730956	
12	Aged 60 to 64 years	3.246628	-3.099293	
11	Aged 55 to 59 years	3.646537	-3.497448	
10	Aged 50 to 54 years	3.564099	-3.409749	
9	Aged 45 to 49 years	3.127357	-2.999316	
8	Aged 40 to 44 years	2.950204	-2.837949	
7	Aged 35 to 39 years	3.072983	-2.962482	
6	Aged 30 to 34 years	3.199270	-3.048428	
5	Aged 25 to 29 years	2.927403	-2.902847	
4	Aged 20 to 24 years	2.853735	-2.974760	
3	Aged 15 to 19 years	2.681844	-2.816901	
2	Aged 10 to 14 years	2.743234	-2.862505	
1	Aged 5 to 9 years	2.662551	-2.795854	
0	Aged 4 years and under	2.346833	-2.469612,	Ages
Females Males				
18	Aged 90 years and over	0.649998	-0.308910	
17	Aged 85 to 89 years	1.023265	-0.691830	
16	Aged 80 to 84 years	1.579947	-1.274254	
15	Aged 75 to 79 years	2.249252	-1.985391	
14	Aged 70 to 74 years	3.008656	-2.825241	
13	Aged 65 to 69 years	2.941082	-2.780191	
12	Aged 60 to 64 years	3.285388	-3.140586	
11	Aged 55 to 59 years	3.677961	-3.494546	
10	Aged 50 to 54 years	3.565338	-3.369051	
9	Aged 45 to 49 years	3.066577	-2.928211	
8	Aged 40 to 44 years	2.883161	-2.770538	
7	Aged 35 to 39 years	3.085883	-2.905686	
6	Aged 30 to 34 years	3.249992	-3.073012	
5	Aged 25 to 29 years	3.040834	-2.963607	
4	Aged 20 to 24 years	2.950735	-3.092319	
3	Aged 15 to 19 years	2.741577	-2.921775	
2	Aged 10 to 14 years	2.860636	-3.018309	
1	Aged 5 to 9 years	2.770538	-2.896032	
0	Aged 4 years and under	2.439103	-2.561380,	Ages
Females Males				
18	Aged 90 years and over	0.603301	-0.282324	
17	Aged 85 to 89 years	0.946123	-0.633550	
16	Aged 80 to 84 years	1.468759	-1.169630	
15	Aged 75 to 79 years	2.023325	-1.769569	
14	Aged 70 to 74 years	2.576211	-2.369509	
13	Aged 65 to 69 years	2.524115	-2.414883	
12	Aged 60 to 64 years	2.877021	-2.828286	
11	Aged 55 to 59 years	3.366047	-3.278661	

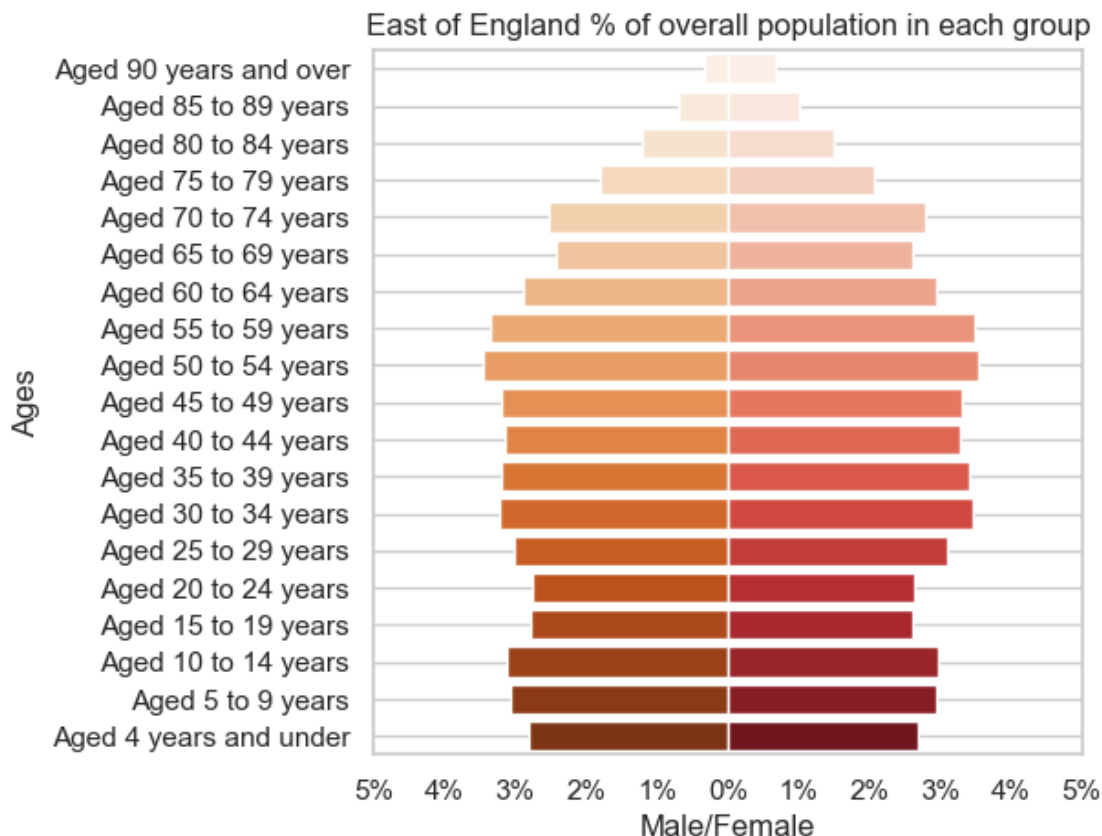
10	Aged 50 to 54 years	3.488724	-3.446711	
9	Aged 45 to 49 years	3.120694	-3.098847	
8	Aged 40 to 44 years	3.088764	-2.976170	
7	Aged 35 to 39 years	3.310591	-3.125735	
6	Aged 30 to 34 years	3.468558	-3.256814	
5	Aged 25 to 29 years	3.179511	-3.092125	
4	Aged 20 to 24 years	3.004739	-3.098847	
3	Aged 15 to 19 years	2.917353	-3.090445	
2	Aged 10 to 14 years	3.063557	-3.228246	
1	Aged 5 to 9 years	3.018183	-3.174470	
0	Aged 4 years and under	2.750983	-2.885423,	Ages
	Females		Males	
18	Aged 90 years and over	0.576558	-0.262735	
17	Aged 85 to 89 years	0.941468	-0.625821	
16	Aged 80 to 84 years	1.466939	-1.136695	
15	Aged 75 to 79 years	1.957743	-1.711429	
14	Aged 70 to 74 years	2.705809	-2.486863	
13	Aged 65 to 69 years	2.603635	-2.512407	
12	Aged 60 to 64 years	3.028755	-2.948475	
11	Aged 55 to 59 years	3.461174	-3.366297	
10	Aged 50 to 54 years	3.519559	-3.424683	
9	Aged 45 to 49 years	3.114509	-3.050650	
8	Aged 40 to 44 years	2.994088	-2.913808	
7	Aged 35 to 39 years	3.264122	-3.119982	
6	Aged 30 to 34 years	3.459349	-3.242227	
5	Aged 25 to 29 years	3.253175	-3.145526	
4	Aged 20 to 24 years	3.185666	-3.156474	
3	Aged 15 to 19 years	2.877317	-3.017808	
2	Aged 10 to 14 years	2.955773	-3.112684	
1	Aged 5 to 9 years	2.910159	-3.045176	
0	Aged 4 years and under	2.660196	-2.786090])	

```
[17]: east_of_england = dfs_dict['East of England']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=east_of_england, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=east_of_england, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('East of England % of overall population in each group');
```

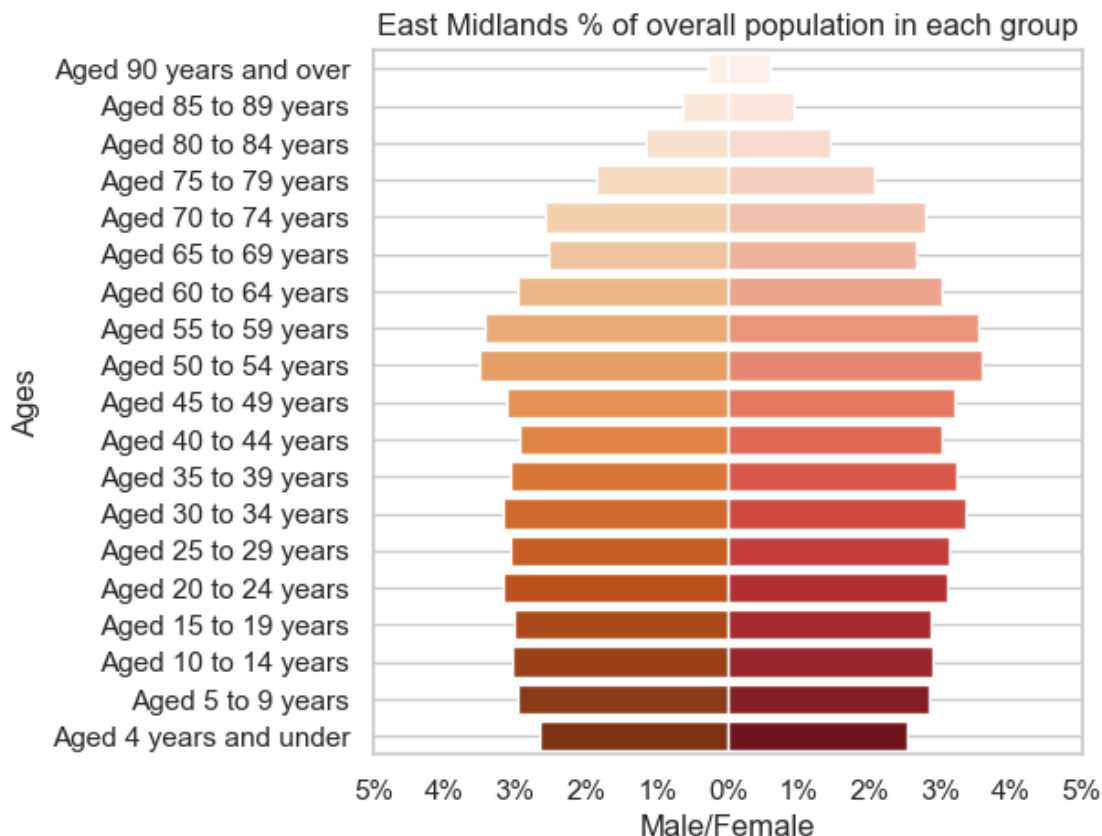


```
[18]: east_midlands = dfs_dict['East Midlands']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=east_midlands, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=east_midlands, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('East Midlands % of overall population in each group');
```



```
[19]: london = dfs_dict['London']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=london, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=london, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('London % of overall population in each group');
```

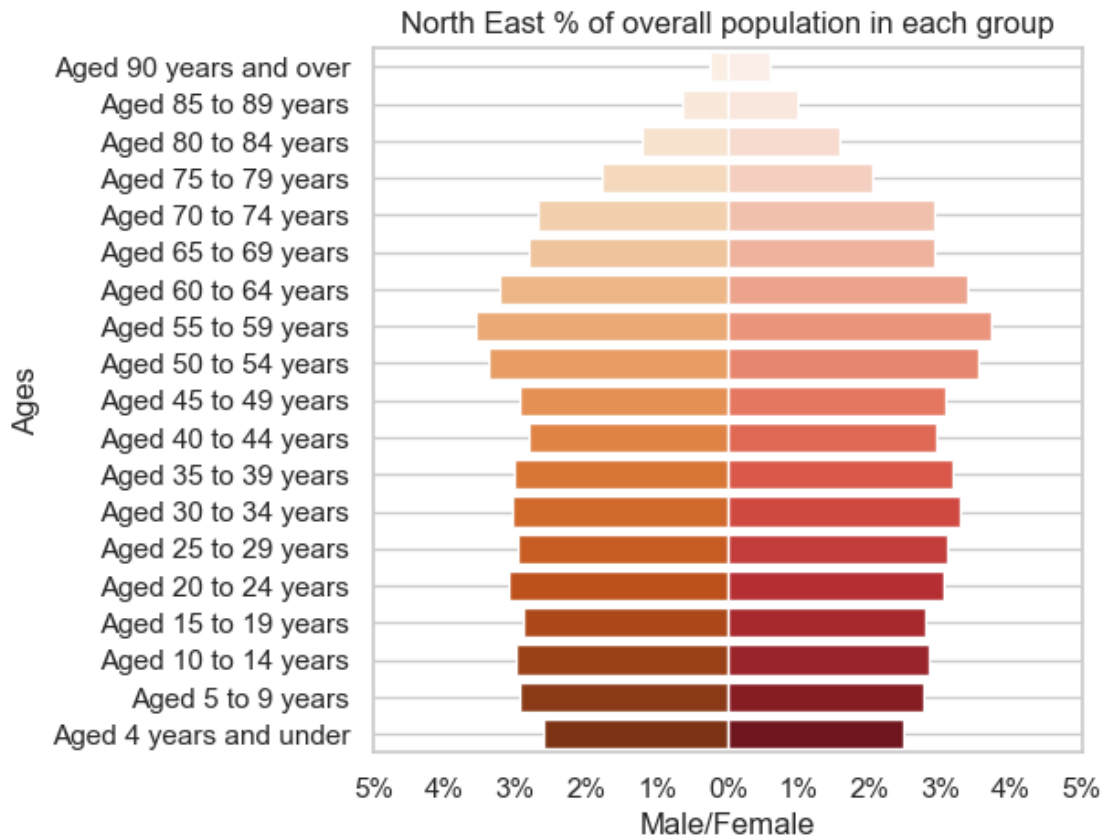


```
[20]: north_east = dfs_dict['North East']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=north_east, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=north_east, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('North East % of overall population in each group');
```

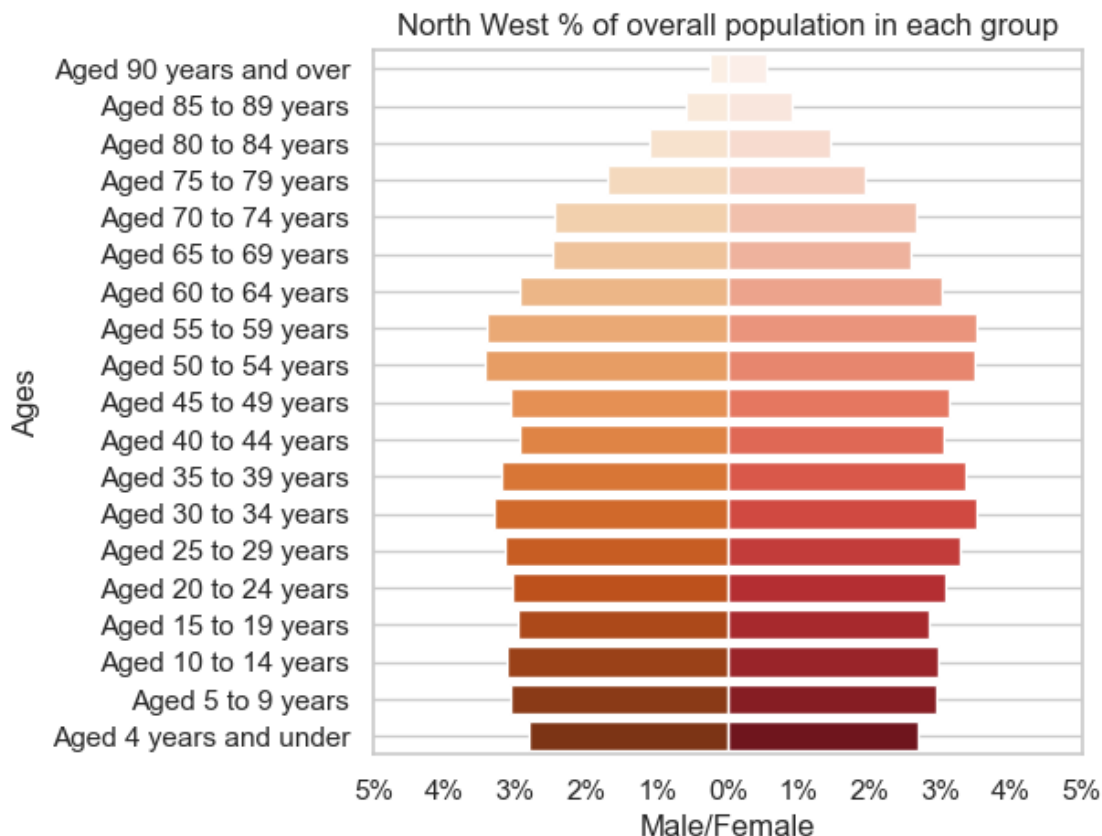


```
[21]: north_west = dfs_dict['North West']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=north_west, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=north_west, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('North West % of overall population in each group');
```

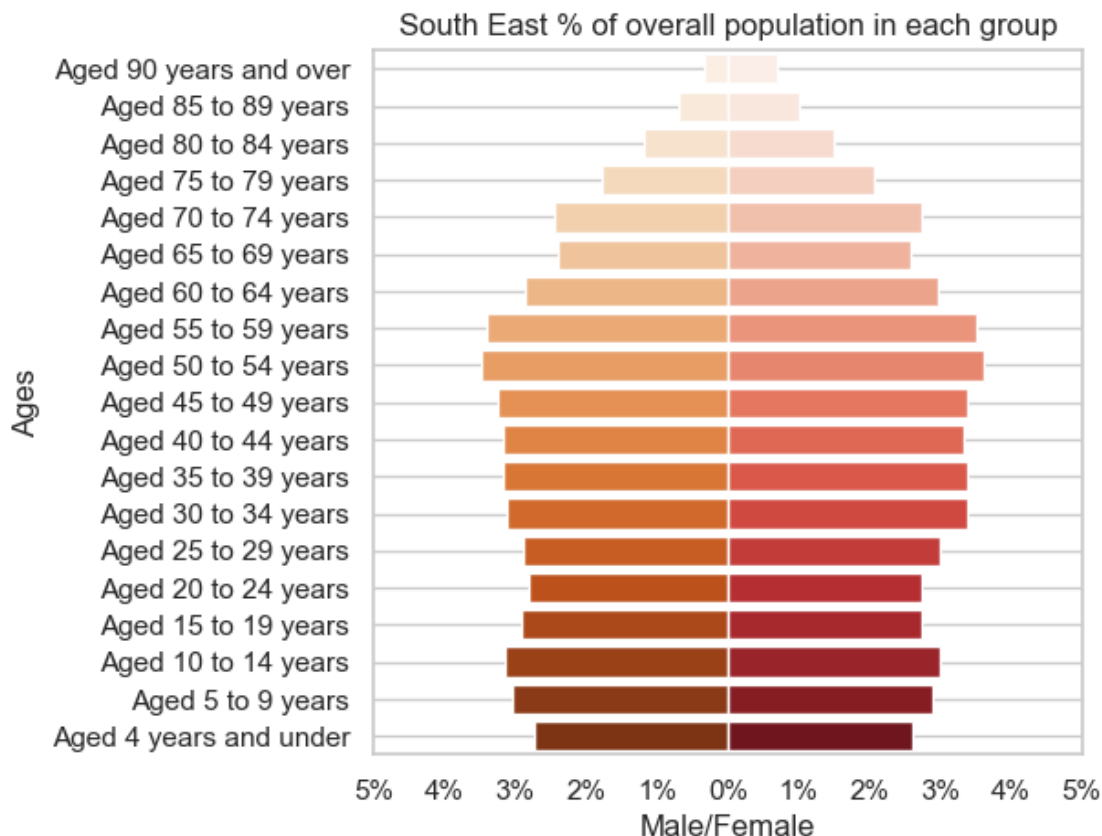


```
[22]: south_east = dfs_dict['South East']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=south_east, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=south_east, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('South East % of overall population in each group');
```

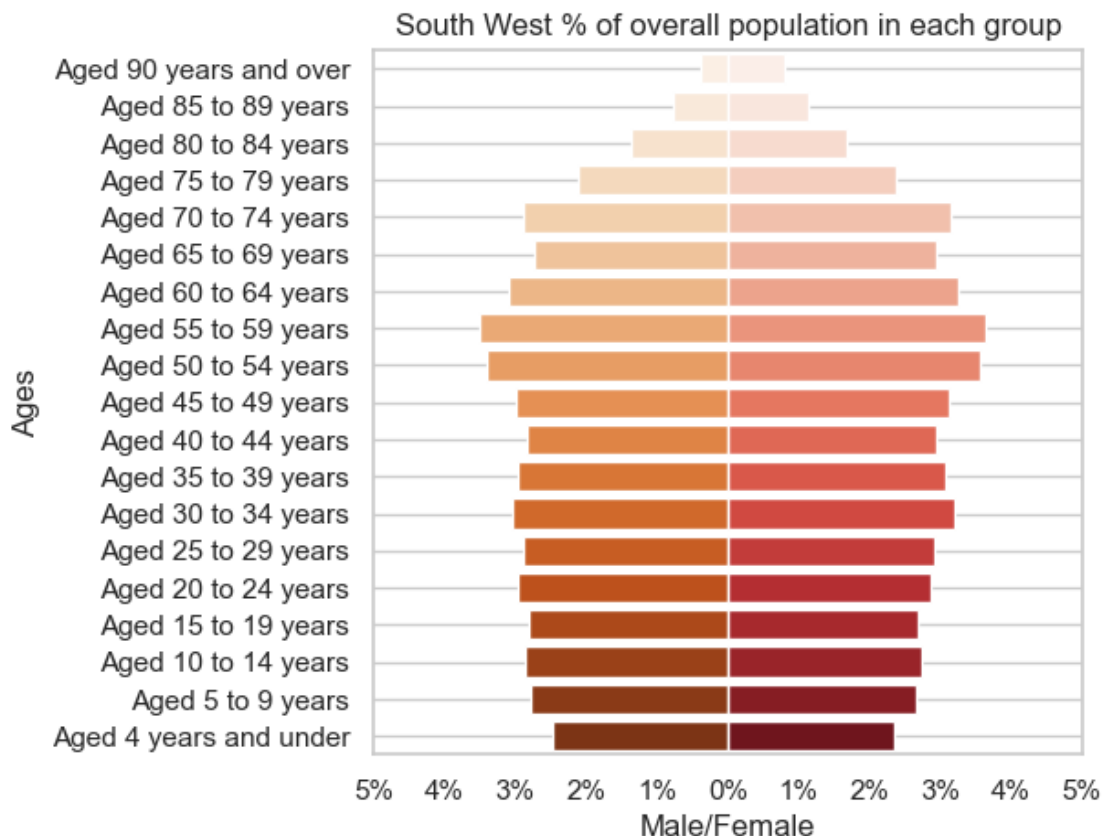



```
[23]: south_west = dfs_dict['South West']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=south_west, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=south_west, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('South West % of overall population in each group');
```

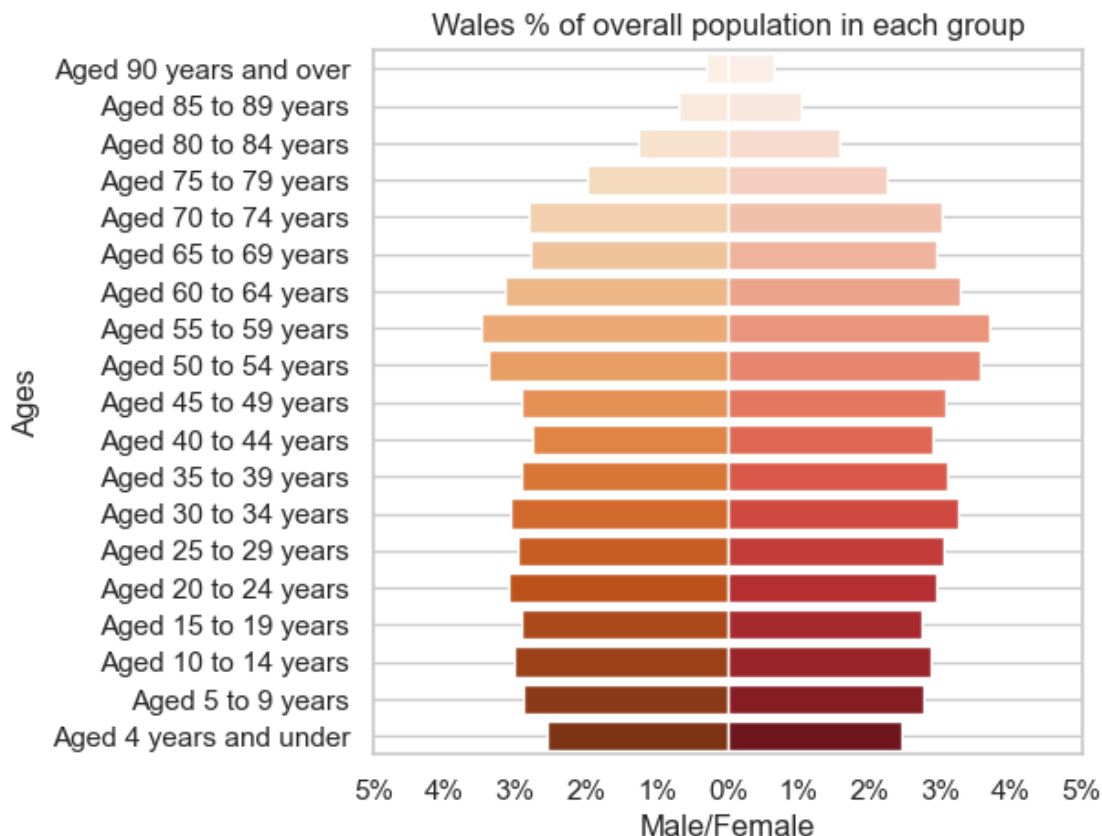


```
[24]: wales = dfs_dict['Wales']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=wales, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=wales, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('Wales % of overall population in each group');
```

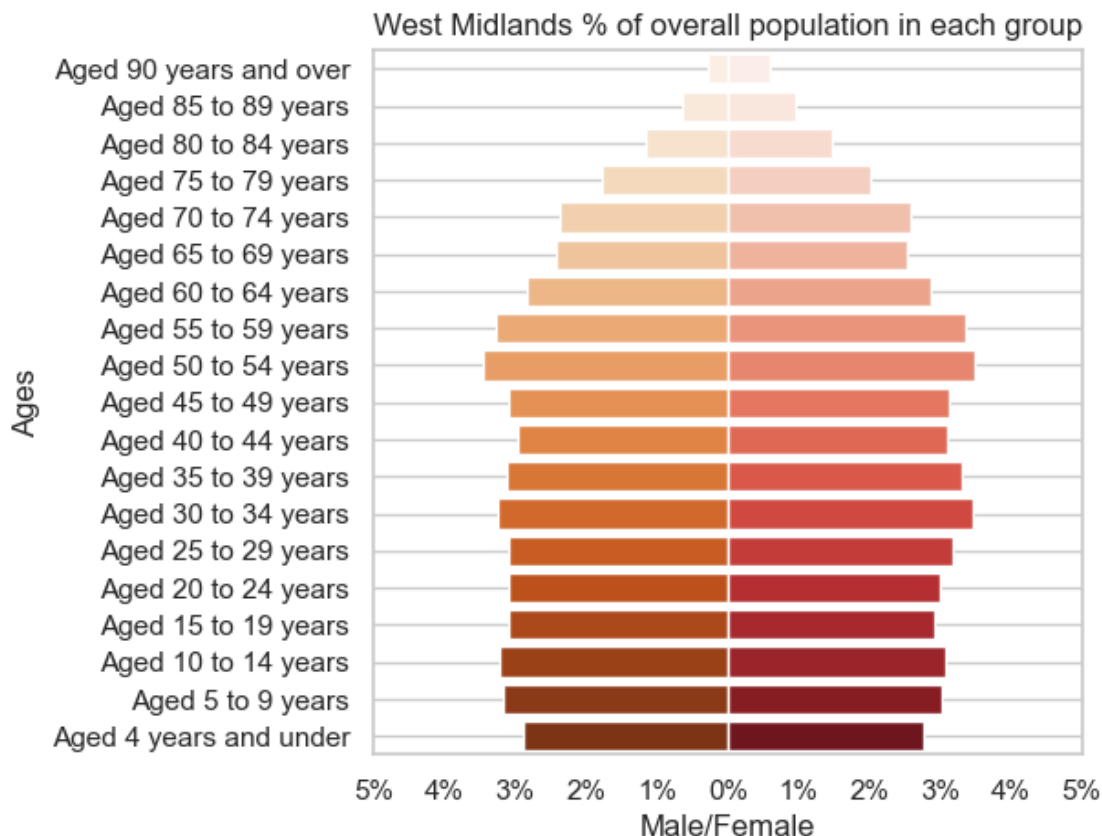


```
[25]: west_midlands = dfs_dict['West Midlands']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=west_midlands, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=west_midlands, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('West Midlands % of overall population in each group');
```

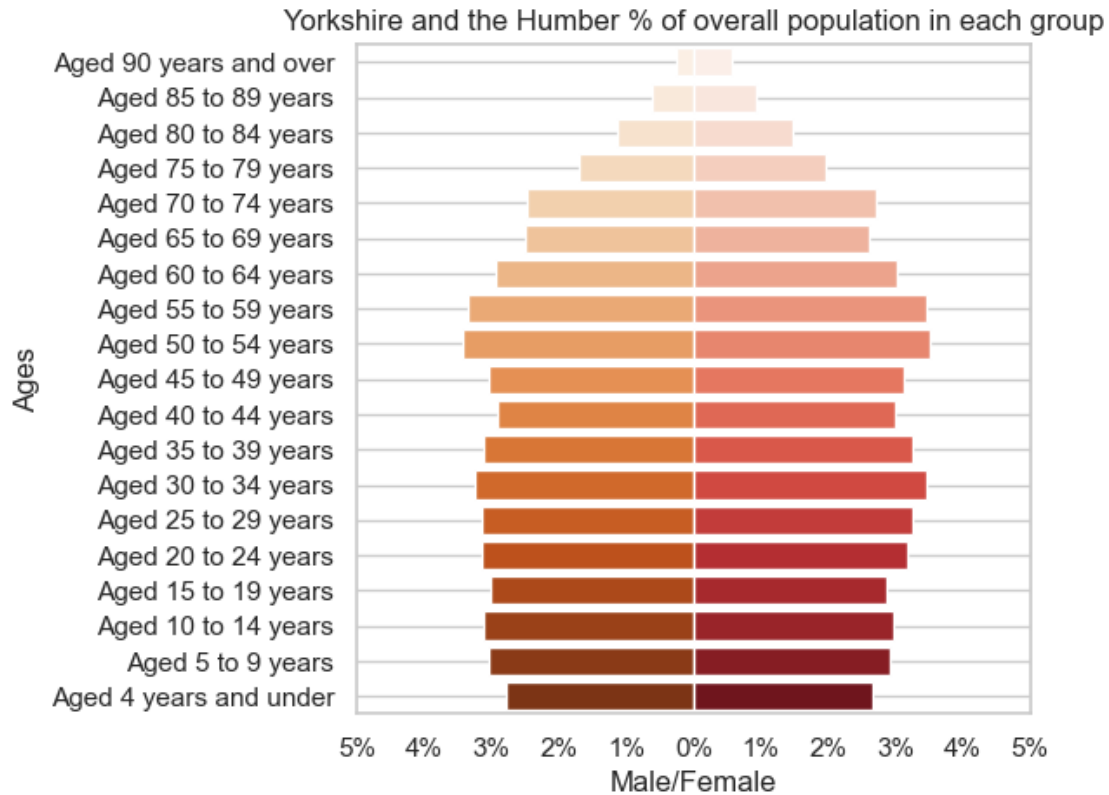


```
[26]: yorkshire = dfs_dict['Yorkshire and the Humber']

# Reset the chart plot size
sns.set(rc={"figure.figsize":(5, 5)},
        style='whitegrid')

ax1 = sns.barplot(x='Males', y='Ages', data=yorkshire, palette="Oranges")
ax2 = sns.barplot(x='Females', y='Ages', data=yorkshire, palette="Reds")

plt.xlabel("Male/Female")
plt.grid()
plt.xticks(ticks=[ -5, -4, -3, -2, -1 , 0, 1, 2, 3, 4,5],
labels=['5%', '4%', '3%', '2%', '1%', '0%', '1%', '2%', '3%', '4%', '5%'])
plt.title('Yorkshire and the Humber % of overall population in each group');
```



The population pyramids for much of the regions are quite similar apart from London. They show the proportions of children seems to be between 4 and 6% of the overall population. There is often a small bulge around ages 10-14 presumably reflecting a small “baby boom”. Some regions show a small dip in proportion of people aged 40-49. There is a bulge for all regions except London at either age 50-54 or 55-59 before the proportions steadily decline as the age groups increase.

The population pyramids for the East Midlands, North East, South West and Wales all show a more pronounced “arrowhead” shape at the top of the population pyramids for ages 50-90+. This shows they have a larger proportion of the population over 50 compared with the other regions.

London with its “Christmas tree” shape is different from the other regions. There is fairly large proportion of children then a big jump in the proportions of 25-34 years old (presumably reflecting people moving to London for work or studying) which then declines quite quickly as the age groups increase. Thus, London looks to have a younger population on average compared to the other regions.

Overall it seems there is some difference between the age profiles in different regions of England and Wales, however there is a big difference between the age profile of London and the other regions of England and Wales. If I had more time and different data it would be interesting to see if this difference was also reflected in other large UK cities.

[]: