SinShady / **phase-1-project-west-ds-082420**

forked from learn-co-students/phase-1-project-west-ds-082420

SinShady with final PDFs History

1 contributor

Raw Blame

1276 lines (1276 sloc) 443 KB

Oppurtunity Youth in King County

For this project we will be looking at data in South King County, Washington at the specific demographic of Opportunity Youth. We hope to find trends that shed light on the systematic difficulties Oppurtunity Youths face as barriers to education and employment.

In [1]:
```
#import libraries and data
import psycopg2
import pandas as pd
import numpy as np
import geopandas as gpd
import matplotlib.pyplot as plt
import os
```

Creating a map of King County

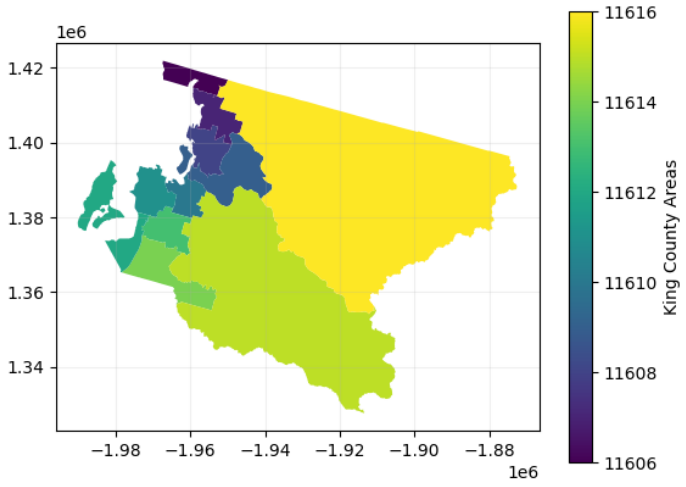
In [2]:
```
#WARNING: RUN THIS LINE ONLY ONCE. MOVES UP 2 PARENT DIRECTORIES TO IMPORT SOURCE DATA
#Note: Tried the method given in the data download and exploration file but could not get it to work
os.chdir("..\\..\\")
```

In [3]:
```
#Read in shape file for geo data
gdf = gpd.read_file("src/data/shapefiles/ipums_puma_2010.shp")
```

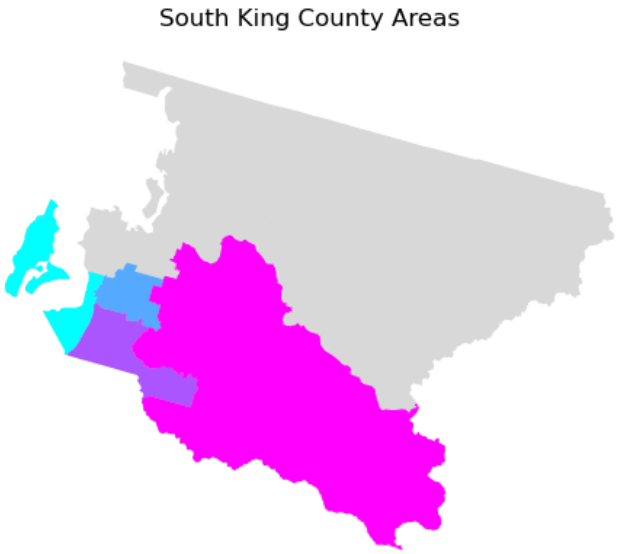
In [4]:
```
#Extract Washington, King County, and South King County
gdf["PUMA"] = gdf["PUMA"].astype(str).astype(int)
washington_map = gdf[(gdf['State']=='Washington')]
greater_king_co_map= gdf[(gdf.PUMA >= 11606) & (gdf.PUMA <= 11616)]
south_king_co_map=gdf[(gdf.PUMA >= 11612) & (gdf.PUMA <= 11615)]
```

In [5]:
```
#Create a cloropeth of pumas in King County
plt.style.use('default')
fig, ax=plt.subplots()
plt.grid(alpha = 0.2)
greater_king_co_map.plot(column='PUMA',ax=ax,
```

```
legend = True, legend_kwds={'label': "King Co
unty Areas"})
plt.savefig('./reports/figures/project_one_ma
p_of_king_county.png')
plt.show()
```



In [6]:
```
#Create a cloropeth of pumas in South King Co
unty
fig, ax=plt.subplots()
# washington_map.plot(ax=ax, color='grey',alp
ha = .1, zorder=1)
greater_king_co_map.plot(ax=ax, color='grey',
alpha = .3, zorder=1);
south_king_co_map.plot(ax=ax, column = 'PUMA'
, cmap='cool', zorder= 2)
plt.title('South King County Areas')
plt.axis('off')
plt.savefig('./reports/figures/project_one_ma
p_of_south_king_county.png')
plt.show()
```



Connecting to the Opportunity

Connecting to the Oppurtunity Youth Database

In [7]:
```
#connect to database oppurtunity_youth and cr
eate cursor. We are using psycopg2 and postgr
es to obtain our data
DBNAME = "opportunity_youth"
conn = psycopg2.connect(dbname=DBNAME)
cursor = conn.cursor()
```

In [8]:
```
#create a list of the table names
cursor.execute("""SELECT table_name FROM info
rmation_schema.tables
       WHERE table_schema = 'public'""")
tables = []
for table in cursor.fetchall():
    tables.append(table[0])
tables
```

Out[8]:
```
['pums_2017',
 'puma_names_2010',
 'wa_jobs_2017',
 'wa_geo_xwalk',
 'ct_puma_xwalk']
```

Querying the Data

In [9]:
```
# Creating a filtered data fram for youth in
 South King County Washington
# Filter puma_names_2010 to King County in Wa
shington
# Filter pums_2017 for ages between 16 and 25
# Join pums_2017 with filtered puma_names_201
0
# SQL query with info for puma, person, age,
 education, and work
# Column info:
    # PUMA          Public use microdata area
code
    # puma_name     County, city, location
    # SERIALNO      Housing unit/GQ person se
rial number
    # sporder       Which person in housing u
nit
    # agep          Age
    # sch           school enrollment (1 = ha
s not attended in last 3 months)
    # schl          Education level
    # esr           Employment status

df weight = pd.read sql(""" SELECT puma, puma
```

```
_name, serialno, sporder, agep, sch, schl, es
r, pwgtp
                    FROM puma_names_2010 pn
                    JOIN pums_2017 pms
                    USING (puma)
                    WHERE state_name LIKE 'Wa
shington%'
                    AND puma_name LIKE 'Kin
g%'
                     AND puma_name LIKE '%Sou
th%'
                    AND agep BETWEEN 15.9 AND
25.0
                    ;""", conn)

df_weight
```

Out[9]:

	puma	puma_name	serialno	sporder
0	11606	King County (Northwest)--Shoreline, Kenmore & ...	2013000003218	01
1	11606	King County (Northwest)--Shoreline, Kenmore & ...	2013000003218	02
2	11612	King County (Far Southwest)--Federal Way, Des ...	2013000007063	02
3	11613	King County (Southwest Central)--Kent City ...	2013000008046	02
4	11614	King County (Southwest)--Auburn City & Lakelan...	2013000011255	02
...	...	...	...	...
3177	11606	King County (Northwest)--Shoreline, Kenmore & ...	2017001491175	01
		King County		

3178	11606	(Northwest)--Shoreline, Kenmore & ...	2017001511157	01
3179	11606	King County (Northwest)--Shoreline, Kenmore & ...	2017001526134	01
3180	11606	King County (Northwest)--Shoreline, Kenmore & ...	2017001530240	01
3181	11613	King County (Southwest Central)--Kent City ...	2017001530818	01

3182 rows × 9 columns

Accounting for the weights in the data

In [10]:
```
#Function that expands the table from 2878 rows to 68347 using pwgtp to account for the weights
def duplicate_rows(df, countcol):
    for _, row in df.iterrows():
        for i in range(int(row[countcol])-1):
            # Append this row at the end of the DataFrame
            df = df.append(row)

    # Remove countcol (could do a drop too to do that...)
    notcountcols = [x for x in df.columns if x != countcol]
    df = df[notcountcols]
    # optional: sort it by index
    df.sort_index(inplace=True)
    return df
```

In [11]:
```
#**WARNING: THIS CELL MAY TAKE 15 MIN TO RUN**
#Runs the function that expands the table
```

```
df_dup = duplicate_rows(df_weight, 'pwgtp')
df_dup = df_dup.reset_index()
```

Filtering and Grouping the Data

In [12]:
```
#Code to filter Educational Groups
def school_range (schl):
    if int(schl) < 15: return "No diploma"
    elif int(schl) < 17: return "HS diploma o
r GED"
    elif int(schl) < 19: return "Some Colleg
e, no degree"
    elif int(schl) < 25: return "Degree (Asso
ciate or higher)"
    else: return "Unknown"
df_dup["School_Level"]= df_dup.schl.apply(sch
ool_range)
```

In [13]:
```
#Code to filter Oppurtunity Youth Status
def Y_Status (esr, sch):
    if (int(esr) == 3 or int(esr) == 6) and i
nt(sch) == 1: return "Opportunity Youth"
    elif (int(esr) == 1 or int(esr) == 2 or i
nt(esr) == 4 or int(esr) == 5) and int(sch) <
= 15: return "Working without Diploma"
    else: return "Not Opportunity Youth"
df_dup["OY_Status"] = df_dup.apply(lambda x:
Y_Status(x["esr"], x["sch"]), axis=1)
```

In [14]:
```
#Code to filter Opportunity Youth Status
def Y_Status (esr, sch):
    if (int(esr) == 3 or int(esr) == 6) and i
nt(sch) == 1: return "Opportunity Youth"
    else: return "Not Opportunity Youth"
df_dup["Is_OY"] = df_dup.apply(lambda x: Y_St
atus(x["esr"], x["sch"]), axis=1)
```

In [15]:
```
#Code that adds Age_Group and groups ages int
o bins
df_dup['Age_Group'] = pd.cut(x=df_dup['agep'
], bins=[16, 18, 21, 24], labels=['16-18', '1
9-21', '21-24'])
```

In [16]:
```
#adds Total_Population column that takes the
 total population of each age group
df_dup["Total_Populations"]=df_dup.groupby("A
ge_Group")["Age_Group"].transform("count")
df_dup["OY_Status_Counts"]=df_dup.groupby("OY
_Status")["OY_Status"].transform("count")
```

In [17]:
```
#Sets up a new dataframe to filter for Oppurt
inuty Youth
df_chart = df_dup
df_chart['schl'] = df_chart['schl'].astype(fl
oat)

#Creates a dataframe of df_dup that is only O
pportunity Youth
oy_chart = df_dup.loc[df_dup["Is_OY"] == "Opp
ortunity Youth"]

#Creates a dataframe of df_dup that is everyo
ne not an Opportunity Youth
noy_chart = df_dup.loc[df_dup["Is_OY"] == "No
t Opportunity Youth"]
```

In [18]:
```
#function that adds percentage of total to co
lumn
def add_perc_to_column(df, column_n, total):
    df[column_n] = (round(df[column_n]/total*
100)).astype(str) + "% : " + (df[column_n]).a
stype(str)
```

In [19]:
```
#Level of Education by Age
#Group by and count by age group and educatio
n level
oy_chart.sort_values(["School_Level"])
totals = oy_chart.groupby(["School_Level"])[
"Total_Populations"].count()
grouper = oy_chart.groupby(["Age_Group", "Sch
ool_Level"], as_index=False).count()
total_pop = totals.sum()
```

Finding meaning in the data

In [20]:
```
#Creates School Level Pivot
#Cannot find a way to add percentage of popul
ation without doing pivot table in excel
sch_piv= pd.pivot_table(data = grouper, index
="School_Level", columns="Age_Group", values=
"Total_Populations",
                        aggfunc = "sum", margi
ns_name="Total", margins=True)
sch_piv
```

Out[20]:

Age_Group	16-18	19-21	21-24	Total
School_Level				
Degree (Associate or higher)	19	341	812	1172

HS diploma or GED	621	1737	1461	3819
No diploma	464	560	531	1555
Some College, no degree	13	131	567	711
Total	1117	2769	3371	7257

In [21]:
```
#Converts to Data Frame and adds percent of p
opulation
sch_piv_df = pd.DataFrame(sch_piv.to_records
()).set_index('School_Level')
sch_piv_perc_df = sch_piv_df.copy()
[add_perc_to_column(sch_piv_perc_df, c, total
_pop) for c in sch_piv_perc_df.columns]
sch_piv_perc_df
```

Out[21]:

	16-18	19-21	21-24	Total
School_Level				
Degree (Associate or higher)	0.0% : 19	5.0% : 341	11.0% : 812	16.0% : 1172
HS diploma or GED	9.0% : 621	24.0% : 1737	20.0% : 1461	53.0% : 3819
No diploma	6.0% : 464	8.0% : 560	7.0% : 531	21.0% : 1555
Some College, no degree	0.0% : 13	2.0% : 131	8.0% : 567	10.0% : 711
Total	15.0% : 1117	38.0% : 2769	46.0% : 3371	100.0% : 7257

In [22]:
```
##Opportunity Youth Status by Age
#Group by and count by age group and OY Statu
s
df_dup.sort_values(["Age_Group"])
totals_by_group = df_dup.groupby(["OY_Status"
])["Total_Populations"].count()
grouper_oy = df_dup.groupby(["Age_Group", "OY
_Status"], as_index=False).count()
total_pop = totals_by_group.sum()

#Creates OY_Status Pivot Table with Totals
#Cannot find a way to add percentage of popul
ation without doing pivot table in excel
oy_piv=pd.pivot_table(data=grouper_oy, index=
"OY_Status", columns="Age_Group", values="Tot
al_Populations",
                      aggfunc = "sum", margin
s_name="Total", margins=True)
oy_piv
```

```
oy_piv
```

Out[22]:

Age_Group	16-18	19-21	21-24	Total
OY_Status				
Not Opportunity Youth	10298	5592	2363	18253
Opportunity Youth	1117	2769	3371	7257
Working without Diploma	4895	12349	17253	34497
Total	16310	20710	22987	60007

In [23]:
```
#Converts to Data Frame and adds percent of p
opulation
oy_piv_df = pd.DataFrame(oy_piv.to_records())
.set_index('OY_Status')
oy_piv_perc_df = oy_piv_df.copy()
[add_perc_to_column(oy_piv_perc_df, c, total_
pop) for c in oy_piv_perc_df.columns]
oy_piv_perc_df
```

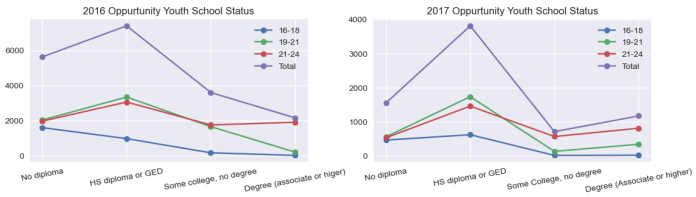
Out[23]:

	16-18	19-21	21-24	Total
OY_Status				
Not Opportunity Youth	17.0% : 10298	9.0% : 5592	4.0% : 2363	30.0% : 18253
Opportunity Youth	2.0% : 1117	5.0% : 2769	6.0% : 3371	12.0% : 7257
Working without Diploma	8.0% : 4895	21.0% : 12349	29.0% : 17253	57.0% : 34497
Total	27.0% : 16310	35.0% : 20710	38.0% : 22987	100.0% : 60007

In [24]:
```
#Create data fram from csv of 2016 data
#Add Opportunity row
old_oy = pd.read_csv('src/data/csv_files/oy_c
sv.csv')
old_oy = pd.DataFrame(old_oy)
old_oy = old_oy.set_index('op_youth', drop=Tr
ue)

#Prepping the the 2017 data for comparison
sch_oy_df=sch_piv_df.append(oy_piv_df[1:2]).r
eindex(["Opportunity Youth", "No diploma", "H
S diploma or GED", "Some College, no degree",
"Degree (Associate or higher)"])
```

Visuals Comparing 2016 to 2017

In [25]:
```
#A plot to compare the 2016 vs 2017 data
#Note: We can see that our pupulation sample
 is significantly smaller. This is because we
did not include Renton City
#From the data we can see a greater percent o
f people have finished highschool 2017 but no
t beyond that
plt.style.use('seaborn')
fig, ax = plt.subplots(1,2, figsize = (12,4))
fig.tight_layout(pad=5)
old_oy2=old_oy[1:]
ax[0].plot(old_oy2.index, old_oy2.values, mar
ker = 'o')
ax[0].set_title('2016 Oppurtunity Youth Schoo
l Status');
sch_oy2=sch_oy_df[1:]
ax[1].plot(sch_oy2.index, sch_oy2.values, mar
ker = 'o')
ax[1].set_title('2017 Oppurtunity Youth Schoo
l Status');

for ax in fig.axes:
    plt.sca(ax)
    plt.xticks(rotation=10)
    plt.legend(['16-18', '19-21', '21-24', 'T
otal'])

plt.savefig('./reports/figures/project_2016vs
2017_oy_education_level.png', dpi=300, bbox_i
nches='tight')
```



Graphs that further demonstrate educational trend in Oppurtunity Youth

In [26]:
```
#Sets up a new dataframe to draw our graphs d
rom
df_chart = df_dup
```

```
df_chart['schl'] = df_chart['schl'].astype(float
oat)
#
#Creates a dataframe of df_dup2 that is only
 Opportunity Youth
oy_chart = df_dup.loc[df_dup["Is_OY"] == "Opp
ortunity Youth"]
#Creates a dataframe of df_dp2 that is everyo
ne not an Opportunity You
noy_chart = df_dup.loc[df_dup["Is_OY"] == "No
t Opportunity Youth"]
```

In [27]:
```
#Builds, labels, titles, the scatter
plt.style.use("default")
fig, ax = plt.subplots(figsize=(10, 6))
ax_scatter = plt.scatter(df_dup.agep, df_dup.
schl, alpha=.01, c='green')
plt.title('Level of Education by Age For Yout
h in South King County', fontsize=16, y=1.05)
plt.xlabel('Age (in years)', fontsize=12)
plt.ylabel('Level of Education', fontsize=12)
plt.yticks([1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11
, 12, 13, 14, 15, 16, 17, 18, 19, 20, 21, 22,
23], [1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12,
13, 14, 'High School Degree',16, 17, 18, 19,
'Associates', 'Bachelors', 22, 23])
plt.grid(alpha=.2)

#best fit line
plt.plot(noy_chart.agep, np.poly1d(np.polyfit
(noy_chart.agep, noy_chart.schl, 1))(noy_char
t.agep))

#Finding best fit M and B
acx = df_dup.agep
acy = df_dup.schl
acy.to_numpy(dtype="float32")
acx.to_numpy(dtype="float32")
m, b = np.polyfit(acx, acy, 1)

#Saves Graph
plt.savefig('./reports/figures/project_one_sc
atter_all.png', dpi=300, bbox_inches='tight')

#Prints m, then b, then plots the graph
print(m)
print(b)
plt.show()
```

```
0.5578685984084439
5.036902861584741
```