Testing a Potential Moderator (Confounder) in between the response variable life expectancy and the explanatory variable internet user rate for different countries from the GapMinder dataset

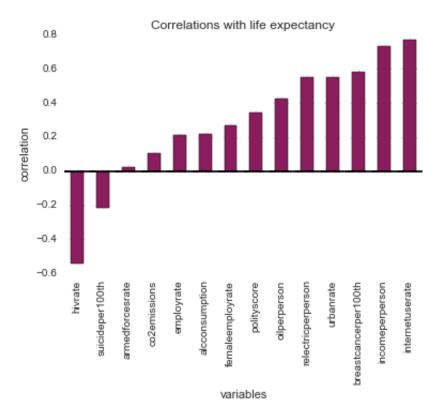
First, a *correlation analysis* was conducted on the *GapMinder* dataset to understand the association of 14 explanatory variables (including income per person, alcohol consumption, armed forces rate, breast cancer per 100th, co2 emissions, female employment rate, hiv rate, internet use rate, oil per person, polity score, relectric per person, suicide per 100th, employment rate, urbanization rate) with the variable *life expectancy*.

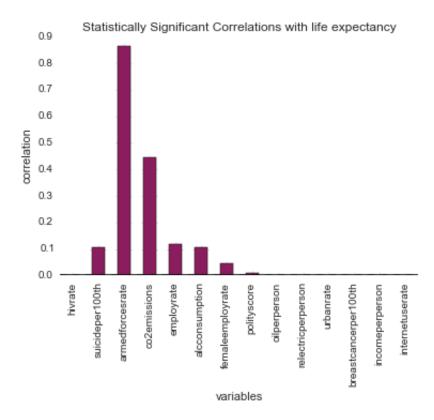
After removing the obeservations with missing values the pearson correlation coefficient is computed. As can be seen from the below results, the variable *internetuserate* has a strong positive correlation (with correlation coefficient ~ 0.77 and significant at 1% level, with a very low p-value) with the variable *life expectancy*.

Out[17]:

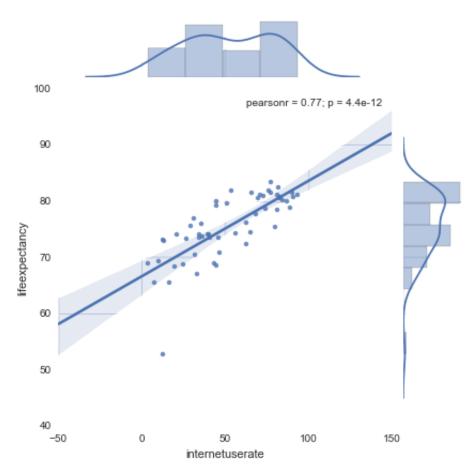
	variables	pearson-r	p-value
0	hivrate	-0.542506	1.566318e-05
0	suicideper100th	-0.218335	1.059663e-01
0	armedforcesrate	0.023648	8.626540e-01
0	co2emissions	0.103990	4.456349e-01
0	employrate	0.210334	1.197189e-01
0	alcconsumption	0.218541	1.056298e-01
0	femaleemployrate	0.268129	4.571763e-02
0	polityscore	0.344843	9.248381e-03
0	oilperperson	0.422911	1.165352e-03
0	relectricperperson	0.551581	1.052532e-05
0	urbanrate	0.552084	1.029253e-05
0	breastcancerper100th	0.580247	2.769328e-06
0	incomeperperson	0.732452	1.400123e-10
0	internetuserate	0.769160	4.381504e-12

<matplotlib.figure.Figure at 0x61e85d0>





Out[5]:
 <seaborn.axisgrid.JointGrid at 0x301ee70>



Analysis 1 (with correlation coefficient)

We want to answer the following question: does the variable *income per person* **moderate** the relationship between the qunatitative explanatory variable *internet use rate* and the quantitative response variable *life expectancy*? To answer this, the income per person variable was partitioned into **4 groups** (LOW, MID, HIGH, VERY HIGH) using the quartiles as shown below, after removing the NA values.

```
56,000000
count
mean
         12982.654643
         12712.681024
std
           558.062877
min
          2532.598585
25%
50%
          6219.692968
75%
         25373.478550
         39972.352768
max
```

Name: incomeperperson, dtype: float64

association between internetuserate and lifeexpectancy for LOW income countries (0.19318865715608682, 0.50814385277502172)

association between internetuserate and lifeexpectancy for MIDDLE income countries

(0.49179543859770558, 0.07407099884980306)

association between internetuserate and lifeexpectancy for HIGH income countrie \boldsymbol{s}

(0.26833836982846088, 0.35361855231884515)

association between internetuserate and lifeexpectancy for VERY HIGH income countries

C:\Python27\lib\site-packages\IPython\kernel__main__.py:13: SettingWithCopyWarnin
g:

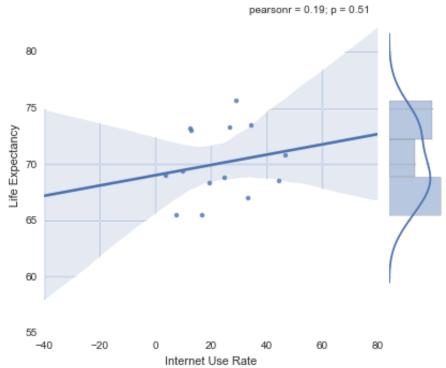
A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead

See the the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

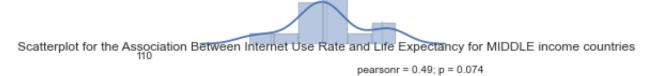
As can be seen from the below scatterplots for different income groups, the income groups LOW, MID and HIGH appears to have somewhat positive corrletaion in between the variables *internet user rate* and *life expectancy* (altough none of them statistically significant at 5% level), but the income group VERY HIGH almost have no correlation (very weak negative correlation). It implies that the variable *income per person* moderates the relation in between *internet user rate* and *life expectancy* significantly.

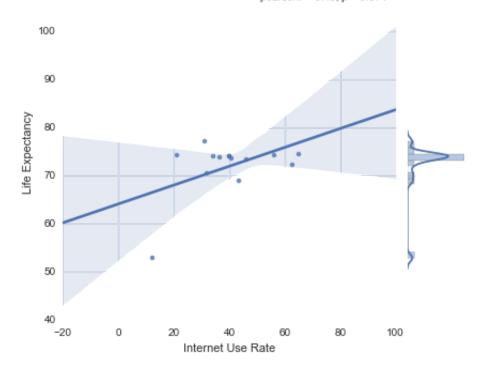
<seaborn.axisgrid.JointGrid object at 0x068F5E30>



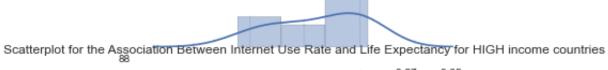


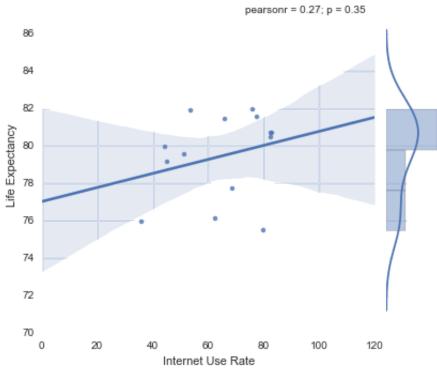
<seaborn.axisgrid.JointGrid object at 0x065FA530>





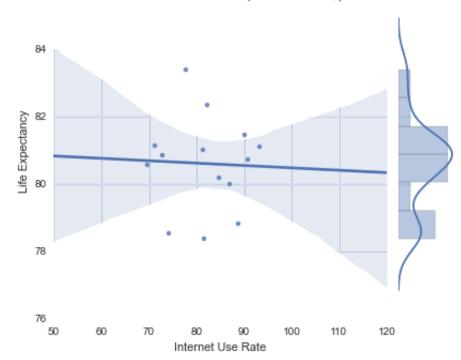
<seaborn.axisgrid.JointGrid object at 0x07F38C90>





<seaborn.axisgrid.JointGrid object at 0x0817A7F0>





Analysis 2 (with one-way ANOVA)

First, let's convert the quantitative variable *internet use rate* to a categoriecal variable by dividing it into **2 groups**: 1=LOW (below 50%) and 2=HIGH (otherwise). First we shall use one-way ANOVA to find whether the association between the categorical explanatory variable *internet use rate* and the quantitative response variable *life expectancy* is significant and also compute mean *life expectancy* for the *internet use rate* groups. As can be seen, the relationship in between these two variables was found to be statistically significant at 5% level (with low *p-value* and *F-statistic* 52.11).

```
56.000000
count
         52.464245
mean
         26.218205
std
min
          3.700003
25%
         33.049632
50%
         48.980090
75%
         77.533598
max
         93.277508
```

Name: internetuserate, dtype: float64

C:\Python27\lib\site-packages\IPython\kernel__main__.py:6: SettingWithCopyWarnin
g:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

OLS Regression Results

Dep. Variable:	lifeexpectancy	eexpectancy R-squared:		0.491				
Model:	OLS	Adj. R	-squared:	0.482				
Method:	Least Squares			52.11				
Date:	Mon, 29 Feb 2016	Prob (I	-statistic):	1.82e-09				
Time:	17:41:03	Log-Lil	Log-Likelihood:		-158.36			
No. Observations:	56	AIC:	•		320.7			
Df Residuals:	54	BIC:	BIC:		324.8			
Df Model:	1							
Covariance Type:	nonrobust							
=======================================								
=======	C	-4-4		D. [4]	FOF 09/ C-			
nf. Int.]	coef	std err	t	P> t	[95.0% Co			
Intercept	71.4784	0.787	90.777	0.000	69.900			
73.057								
C(internetuserate)[T.	.2] 8.0382	1.114	7.218	0.000	5.806			
10.271								
Omnibus:	========== 34.166	Durbin	======== -Watson:	======	1.801			
Prob(Omnibus):	0.000	Jarque-Bera (JB):			108.341			
Skew:	-1.649	-	• •	2.98e-24				
Kurtosis:	8.962	•	•		2.62			
=======================================	3.302 ============	========	 	=======				

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

means for lifeexpectancy by internetuserate LOW vs. HIGH

lifeexpectancy

internetuserate

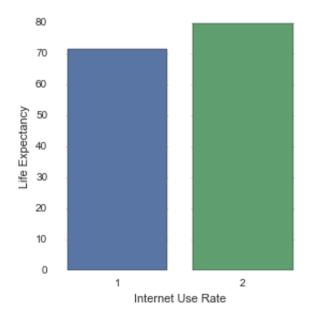
1 71.478357 2 79.516536

standard deviation for mean lifeexpectancy by internetuserate LOW vs. HIGH lifeexpectancy

internetuserate

1 5.217157 2 2.738937

Out[12]:
<matplotlib.text.Text at 0x871c790>



Now, we want to answer the following question: does the variable *income per person* **moderate** the relationship between the (converted) categorical explanatory variable *internet use rate* and the quantitative response variable *life expectancy*? To answer this, the income per person variable was partitioned into **2 groups** (LOW=1, HIGH=2) using the median, after removing the NA values. Then one-way ANOVA test (and Tukey post-hoc test) was conducted on each of the groups. As can be seen, the association between the categorical explanatory variable *internet use rate* and the quantitative response variable *life expectancy* is not significant inside both the income groups, which implies that the variable *income per person* **moderates** the relationship.

OLS Regression Results

=======================================		=======		=======	=======	
Dep. Variable:	lifeexpectancy	R-squared:		0.040		
Model:	OLS	OLS Adj. R-squared:			0.003	
Method:	Least Squares	F-stati	istic:		1.070	
Date:	Mon, 29 Feb 2016	Prob (F	-statistic):	0.311		
Time:	17:41:04	Log-Lik	Log-Likelihood:		-81.965	
No. Observations:	28	AIC:			167.9	
Df Residuals:	26	BIC:	BIC:		170.6	
Df Model:	1					
Covariance Type:	nonrobust					
	:========:	=======		======	=========	
	coef	std err	t	P> t	[95.0% Co	
nf. Int.]						
Intercept 72.586	70.6583	0.938	75.330	0.000	68.730	
C(internetuserate)[T. 8.854	2] 2.9637	2.866	1.034	0.311	-2.927	
Omnibus:	28.841	Durbin-Watson:			2.106	
Prob(Omnibus): 0.000		Jarque-	Bera (JB):		64.278	
Skew:	-2.095	Prob(JE	3):		1.10e-14	
Kurtosis: 9.127		Cond. N	No.	3.27		

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

means for lifeexpectancy by incomegrp

lifeexpectancy

internetuserate

1 70.65832 2 73.62200

standard deviations for lifeexpectancy by incomegrp

lifeexpectancy

internetuserate

1 4.868961 2 1.208500

C:\Python27\lib\site-packages\IPython\kernel__main__.py:7: SettingWithCopyWarnin
g:

A value is trying to be set on a copy of a slice from a DataFrame.

Try using .loc[row_indexer,col_indexer] = value instead

See the the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

OLS Regression Results

		•					
=======================================		=====	=======		:=====	=======	=======
Dep. Variable: lif		lifeexpectancy		R-squared:			0.096
Model: OL			Adj. R-squared:				0.062
Method:	Least Squares						2.769
Date:	Mon, 29 F	eb 2016					0.108
Time:	1	7:41:38		Log-Likelihood:			-56.372
No. Observations:		28	U	•			116.7
Df Residuals:		26					119.4
Df Model:		1					
Covariance Type:	no	nrobust					
=======================================					.=====		
		coef	std err		t	P> t	[95.0% Co
nf. Int.]		COET	stu en		·	7/10	[93.6% CO
m. inc.j							
Intoncont	70	3120	1.086	72	140	0.000	76.081
Intercept	70.	3120	1.000	72.	140	0.000	76.081
80.543	0.7	0440	4 440	4		0.100	0.450
C(internetuserate)[T	.2] 1.	9119	1.149	1.	664	0.108	-0.450
4.273							
=======================================		=====	======		=====	=======	=======
Omnibus:		6.132	Durbir	n-Watson	1:		1.869
Prob(Omnibus): 0.047		Jarque	Jarque-Bera (JB):				
Skew: -0.957		-0.957	Prob(JB):				0.102
Kurtosis: 3.495		Cond. No.				5.95	
=======================================		=====:	======		======	=======	=======

Warnings:

[1] Standard Errors assume that the covariance matrix of the errors is correctly specified.

means for lifeexpectancy by incomegrp

lifeexpectancy

internetuserate

 1
 78.31200

 2
 80.22388

standard deviations for lifeexpectancy by incomegrp

lifeexpectancy

internetuserate

12.12548721.858336

Multiple Comparison of Means - Tukey HSD, FWER=0.05

group1 group2 meandiff lower upper reject

1 2 1.9119 -0.4496 4.2734 False