

# VECM

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In [1]: #=====libraries=====#  
  
import numpy as np  
import pandas as pd  
import statsmodels.api as sm  
import matplotlib.pyplot as plt
```

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In [2]: #=====loading the data from R package=====#  
  
mydata = sm.datasets.get_rdataset("economics", "ggplot2").data  
  
#view the names of the variables  
print(mydata.columns.values)  
  
['date' 'pce' 'pop' 'psavert' 'uempmed' 'unemploy']
```

```
In [3]: #=====Resampling the data to monthly prices=====#  
  
mydata['date'] = pd.to_datetime(mydata['date'])  
mydata = mydata.set_index('date')  
annualdata = mydata.resample('Y').mean()  
print(annualdata)
```

	pce	pop	psavert	uempmed	unemploy
date					
1967-12-31	514.466667	199200.333333	12.433333	4.700000	3012.333333
1968-12-31	556.841667	200663.750000	11.333333	4.500000	2797.416667
1969-12-31	603.650000	202648.666667	10.900000	4.441667	2830.166667
1970-12-31	646.725000	204982.333333	12.800000	4.983333	4127.333333
1971-12-31	699.925000	207589.333333	13.475000	6.275000	5021.666667
1972-12-31	768.150000	209837.583333	12.350000	6.108333	4875.833333
1973-12-31	849.583333	211857.166667	13.450000	5.175000	4359.333333
1974-12-31	930.150000	213814.750000	13.300000	5.183333	5173.333333
1975-12-31	1030.558333	215890.666667	13.408333	8.466667	7939.583333
1976-12-31	1147.666667	217999.250000	11.625000	8.141667	7398.166667
1977-12-31	1273.975000	220193.083333	10.658333	7.066667	6966.916667
1978-12-31	1422.250000	222525.416667	10.725000	5.950000	6187.083333
1979-12-31	1585.425000	225002.833333	10.316667	5.583333	6135.333333
1980-12-31	1750.666667	227621.916667	11.058333	6.658333	7670.666667
1981-12-31	1933.941667	229915.666667	11.716667	7.016667	8276.333333
1982-12-31	2071.241667	232127.833333	12.041667	8.750000	10714.916667
1983-12-31	2281.608333	234246.500000	10.050000	10.158333	10693.750000
1984-12-31	2492.333333	236307.250000	11.325000	7.883333	8529.083333
1985-12-31	2712.841667	238415.500000	9.166667	6.933333	8313.416667
1986-12-31	2886.275000	240592.666667	8.825000	6.958333	8245.000000
1987-12-31	3076.275000	242750.750000	7.908333	6.475000	7413.500000
1988-12-31	3330.000000	244967.500000	8.475000	5.950000	6696.583333
1989-12-31	3576.766667	247285.916667	8.375000	5.216667	6523.666667
1990-12-31	3809.000000	250047.416667	8.366667	5.375000	7061.000000
1991-12-31	3943.450000	253391.833333	8.800000	6.866667	8639.833333
1992-12-31	4197.566667	256777.166667	9.450000	8.658333	9611.166667
1993-12-31	4451.983333	260146.416667	7.925000	8.308333	8926.666667
1994-12-31	4720.958333	263324.583333	6.908333	9.125000	7975.500000
1995-12-31	4962.600000	266458.166667	6.991667	8.241667	7406.916667
1996-12-31	5244.600000	269580.666667	6.558333	8.241667	7231.083333
1997-12-31	5536.783333	272822.083333	6.333333	7.933333	6728.666667
1998-12-31	5877.250000	276022.333333	6.800000	6.683333	6203.833333
1999-12-31	6279.083333	279194.916667	5.075000	6.325000	5878.833333
2000-12-31	6762.150000	282295.916667	4.808333	5.933333	5685.083333
2001-12-31	7065.633333	285215.916667	5.033333	6.733333	6829.666667
2002-12-31	7342.683333	288019.000000	5.833333	9.200000	8375.333333
2003-12-31	7723.108333	290733.333333	5.558333	10.175000	8770.333333
2004-12-31	8212.666667	293388.583333	5.158333	9.816667	8139.666667
2005-12-31	8747.133333	296114.916667	3.175000	8.933333	7579.166667
2006-12-31	9260.350000	298929.916667	3.850000	8.225000	6991.250000
2007-12-31	9706.425000	301903.166667	3.741667	8.500000	7073.083333
2008-12-31	9976.333333	304718.000000	4.975000	9.416667	8948.166667
2009-12-31	9842.191667	307373.750000	6.108333	15.675000	14294.500000
2010-12-31	10185.850000	309736.511000	6.550000	21.466667	14807.750000
2011-12-31	10641.125000	311940.749083	7.158333	21.441667	13738.500000
2012-12-31	11006.825000	314163.103750	8.833333	19.166667	12499.166667
2013-12-31	11317.200000	316329.299167	6.408333	16.808333	11457.166667
2014-12-31	11824.025000	318619.452833	7.350000	14.175000	9601.500000
2015-12-31	12120.125000	320159.059500	7.650000	12.400000	8635.750000

```
In [4]: #=====subsetting the data=====#  
  
mydata = annualdata[["pce" , "psavert"]]
```

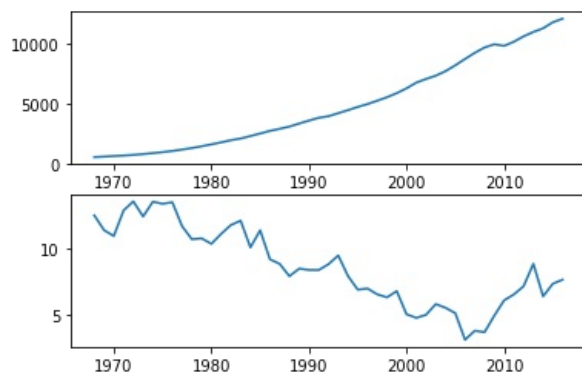
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In [5]: #=====plots=====#

plt.figure()

ax = plt.subplot(211)
ax.plot(mydata["pce"])

ax = plt.subplot(212)
ax.plot(mydata["psavert"])
```

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Out[5]: <matplotlib.lines.Line2D at 0x17ec9c6f5e0>
```



```
In [6]: #=====lag selection=====#

from statsmodels.tsa.vector_ar.vecm import select_order

lags = select_order(data = mydata, maxlags = 20, deterministic="ci")
print(lags.summary())
print(lags.aic, lags.bic, lags.fpe, lags.hqic)

# we select 12 lags
```

```
VECM Order Selection (* highlights the minimums)
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	AIC	BIC	FPE	HQIC
0	9.914	10.29	2.030e+04	10.03
1	9.795	10.37	1.820e+04	9.970
2	9.858	10.62	1.975e+04	10.09
3	9.869	10.82	2.065e+04	10.16
4	9.826	10.97	2.083e+04	10.17
5	9.728	11.06	2.045e+04	10.14
6	9.936	11.46	2.826e+04	10.40
7	9.953	11.67	3.398e+04	10.48
8	9.640	11.54	3.178e+04	10.22
9	9.255	11.35	3.135e+04	9.895
10	7.493	9.776	9838.	8.191
11	4.120	6.594	1093.	4.876
12	-inf*	-inf*	inf	-inf*
13	-96.01	-93.16	2.318e-41	-95.14
14	-89.94	-86.89	2.031e-39	-89.01
15	-102.8	-99.53	1.943e-45	-101.8
16	-100.7	-97.25	7.093e-45	-99.63
17	-100.2	-96.62	5.595e-45	-99.13
18	-97.66	-93.85	4.094e-44	-96.50
19	-106.9	-102.9	2.363e-48*	-105.7
20	-104.2	-100.0	2.032e-47	-103.0

```
-----
12 12 19 12
```

```
In [7]: #=====cointegration test=====#

from statsmodels.tsa.vector_ar.vecm import select_coint_rank

cointest = select_coint_rank(mydata, 0, 12, method="trace", signif= 0.05)
print(cointest.summary())

# we select 1 cointegration relations
```

```
Johansen cointegration test using trace test statistic with 5% significance level
=====
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r_0	r_1	test statistic	critical value
0	2	61.71	15.49
1	2	1.673	3.841

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```
In [8]: #=====VECM model=====#

from statsmodels.tsa.vector_ar.vecm import VECM

vecm_model = VECM(mydata, k_ar_diff = 12, coint_rank = 1, deterministic = 'ci')
```

```

modelfit = vecm_model.fit()
print(modelfit.summary())

```

Det. terms outside the coint. relation & lagged endog. parameters for equation pce

	coef	std err	z	P> z	[0.025	0.975]
-----						
L1.pce	0.1655	0.169	0.978	0.328	-0.166	0.497
L1.psavert	66.9593	24.956	2.683	0.007	18.047	115.872
L2.pce	-0.8143	0.219	-3.723	0.000	-1.243	-0.386
L2.psavert	23.6269	17.984	1.314	0.189	-11.621	58.874
L3.pce	-0.7112	0.221	-3.224	0.001	-1.143	-0.279
L3.psavert	28.0656	14.592	1.923	0.054	-0.534	56.665
L4.pce	-0.4089	0.141	-2.891	0.004	-0.686	-0.132
L4.psavert	19.0218	16.229	1.172	0.241	-12.786	50.830
L5.pce	-0.5055	0.213	-2.368	0.018	-0.924	-0.087
L5.psavert	-32.5523	14.753	-2.206	0.027	-61.468	-3.637
L6.pce	-0.6719	0.188	-3.569	0.000	-1.041	-0.303
L6.psavert	-51.5589	15.639	-3.297	0.001	-82.211	-20.907
L7.pce	0.4663	0.314	1.484	0.138	-0.149	1.082
L7.psavert	-68.6615	16.779	-4.092	0.000	-101.547	-35.776
L8.pce	0.2916	0.287	1.015	0.310	-0.271	0.855
L8.psavert	-12.4937	16.137	-0.774	0.439	-44.121	19.134
L9.pce	-1.4340	0.288	-4.976	0.000	-1.999	-0.869
L9.psavert	-36.5960	14.600	-2.507	0.012	-65.212	-7.980
L10.pce	1.0142	0.303	3.344	0.001	0.420	1.609
L10.psavert	-33.0850	16.398	-2.018	0.044	-65.224	-0.946
L11.pce	-1.0510	0.471	-2.230	0.026	-1.975	-0.127
L11.psavert	-58.8692	13.579	-4.335	0.000	-85.483	-32.255
L12.pce	-1.9491	0.457	-4.261	0.000	-2.846	-1.053
L12.psavert	-29.8472	21.687	-1.376	0.169	-72.352	12.658

Det. terms outside the coint. relation & lagged endog. parameters for equation psavert

	coef	std err	z	P> z	[0.025	0.975]
-----						
L1.pce	-0.0076	0.001	-5.243	0.000	-0.011	-0.005
L1.psavert	-0.5534	0.215	-2.574	0.010	-0.975	-0.132
L2.pce	0.0015	0.002	0.816	0.415	-0.002	0.005
L2.psavert	-0.2576	0.155	-1.662	0.096	-0.561	0.046
L3.pce	-0.0047	0.002	-2.450	0.014	-0.008	-0.001
L3.psavert	-0.1465	0.126	-1.165	0.244	-0.393	0.100
L4.pce	0.0071	0.001	5.851	0.000	0.005	0.010
L4.psavert	0.3019	0.140	2.159	0.031	0.028	0.576
L5.pce	-0.0002	0.002	-0.130	0.897	-0.004	0.003
L5.psavert	0.1076	0.127	0.846	0.397	-0.142	0.357
L6.pce	0.0024	0.002	1.460	0.144	-0.001	0.006
L6.psavert	0.4373	0.135	3.245	0.001	0.173	0.701
L7.pce	-0.0059	0.003	-2.168	0.030	-0.011	-0.001
L7.psavert	0.3583	0.145	2.479	0.013	0.075	0.642
L8.pce	0.0094	0.002	3.817	0.000	0.005	0.014
L8.psavert	0.2546	0.139	1.831	0.067	-0.018	0.527
L9.pce	0.0023	0.002	0.941	0.347	-0.003	0.007
L9.psavert	0.7439	0.126	5.913	0.000	0.497	0.990
L10.pce	-0.0100	0.003	-3.810	0.000	-0.015	-0.005
L10.psavert	0.4386	0.141	3.104	0.002	0.162	0.716
L11.pce	0.0059	0.004	1.459	0.144	-0.002	0.014
L11.psavert	1.0276	0.117	8.783	0.000	0.798	1.257
L12.pce	-0.0052	0.004	-1.327	0.185	-0.013	0.002
L12.psavert	0.1464	0.187	0.784	0.433	-0.220	0.513

Loading coefficients (alpha) for equation pce

	coef	std err	z	P> z	[0.025	0.975]
-----						
ec1	0.1137	0.018	6.434	0.000	0.079	0.148

Loading coefficients (alpha) for equation psavert

	coef	std err	z	P> z	[0.025	0.975]
-----						
ec1	0.0001	0.000	0.905	0.365	-0.000	0.000

Cointegration relations for loading-coefficients-column 1

	coef	std err	z	P> z	[0.025	0.975]
-----						
beta.1	1.0000	0	0	0.000	1.000	1.000
beta.2	-939.5824	128.460	-7.314	0.000	-1191.360	-687.805
const	1.463e+04	1893.644	7.725	0.000	1.09e+04	1.83e+04

```

In [9]: #=====forecasting=====#
print(modelfit.predict(steps=10))

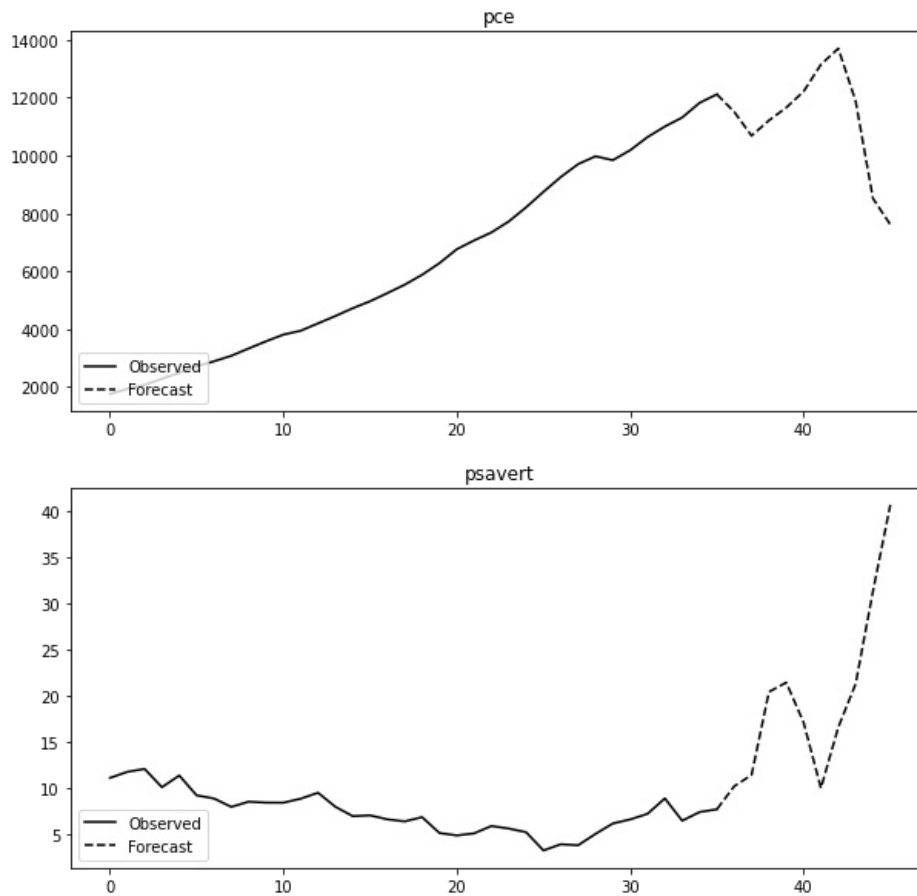
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```
[[1.15118139e+04 1.01575219e+01]
[1.06878488e+04 1.13787053e+01]
[1.12159436e+04 2.04128371e+01]
[1.16591077e+04 2.14294977e+01]
[1.22190876e+04 1.70809221e+01]
[1.31504450e+04 9.94287748e+00]
[1.37058097e+04 1.65653263e+01]
[1.18861348e+04 2.12821936e+01]
[8.51848150e+03 3.13500634e+01]
[7.61383834e+03 4.07475730e+01]]
```

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In [10]: #=====forecasted plots=====#

modelfit.plot_forecast(steps=10, plot_conf_int=False)

#=====END=====#
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In [ ]:

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