

# Assignment 8 - Time series

June 29, 2019

## 0.1 Assignment 8

### 0.1.1 Problem

Using a suitable Python library, extrapolate the time series of the Paladini example over the next 6 months.

### 0.1.2 Resolution

We can extrapolate the decomposition through the python library *statsmodels*. In particular we can use the function *seasonal\_decompose()* in order to extract and print the 4 variables.

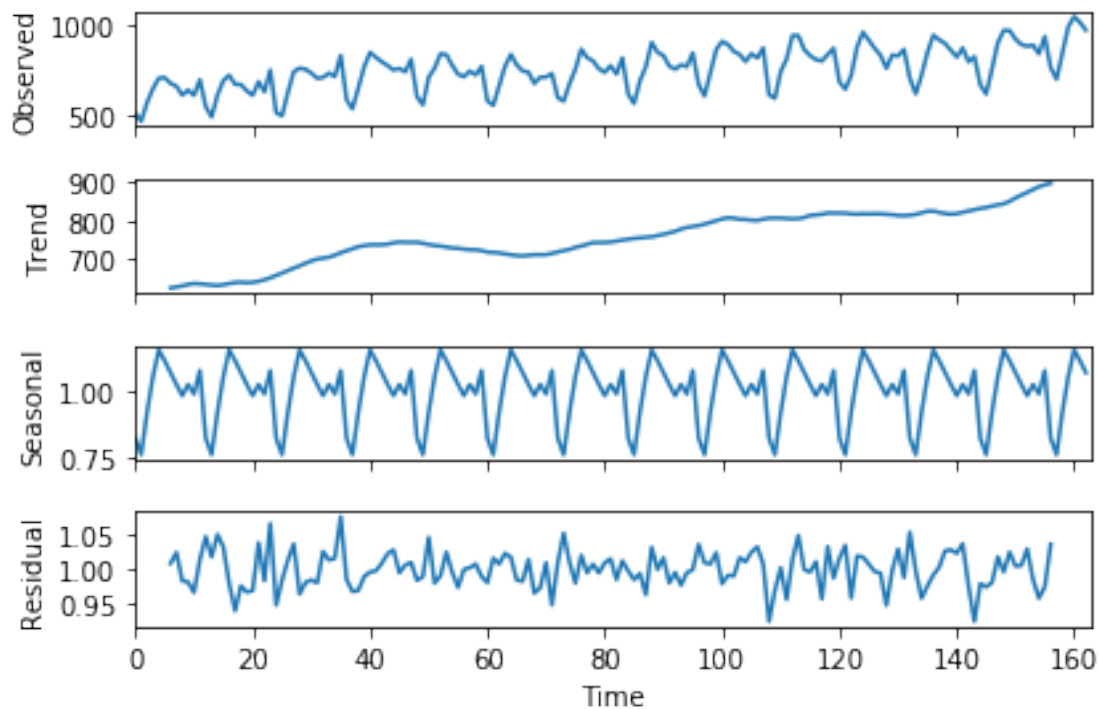
```
In [1]: from random import randrange
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from statsmodels.tsa.seasonal import seasonal_decompose
%matplotlib inline

In [2]: file = "Paladini Associates example.csv"

data = open(file, "r")
data = data.read()
data = data.split(",")
data = [float(d) for d in data]
data = np.array(data)

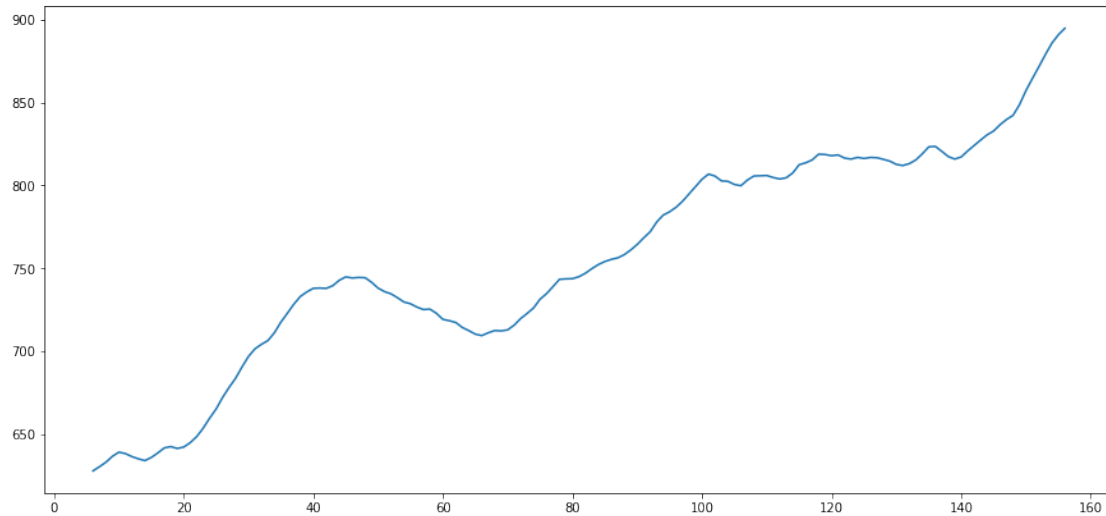
In [3]: m = data.shape[0]

In [4]: pal = seasonal_decompose(data, model='multiplicative', freq=12)
pal.plot()
plt.show()
```



```
In [5]: M = 12
        T = data.shape[0]
        qvt = np.zeros(data.shape[0])
        for i in range(int(M/2), T-int(M/2)):
            tmp = 0
            tmp = data[i-6]/2
            tmp += data[i+6]/2
            for j in range(i-5, i+6):
                tmp += data[j]
            qvt[i] = tmp/M

In [6]: x = np.arange(qvt.shape[0])
        plt.figure(figsize=(15,7))
        plt.plot(x[int(M/2):T-int(M/2)], qvt[int(M/2):T-int(M/2)])
        plt.show()
```



```

In [7]: z = np.polyfit(x[int(M/2):T-int(M/2)], qvt[int(M/2):T-int(M/2)], 1)

In [8]: q = np.zeros(data.shape[0])
        q[int(M/2):T-int(M/2)] = z[0] * x[int(M/2):T-int(M/2)] + z[1]

In [9]: v = np.zeros(data.shape[0])
        v[int(M/2):T-int(M/2)] = qvt[int(M/2):T-int(M/2)] / q[int(M/2):T-int(M/2)]

In [10]: srt = np.zeros(data.shape[0])
         srt[int(M/2):T-int(M/2)] = data[int(M/2):T-int(M/2)]/qvt[int(M/2):T-int(M/2)]

In [11]: s = np.zeros(12)
         for i in range(12):
             count = 0
             for j in range(i+12,T-int(M/2)-1,12):
                 s[i] += srt[j]
                 count += 1
             s[i] = s[i]/count

In [12]: xx = v[149:157]
         yy=np.arange(xx.shape[0])
         zz = np.polyfit(yy, xx, 2)

In [13]: t = np.arange(xx.shape[0],xx.shape[0]+7)
         a = zz[0] * (t**2) + zz[1]*t + zz[2]

In [14]: t = np.arange(157,163)
         qt = z[0]*t+z[1]

In [15]: p = a[:6] * qt * s[1:7]

```

```
In [16]: print(pal.trend)
```

```
[      nan      nan      nan      nan      nan
      nan 627.69583333 630.23333333 632.94583333 636.50416667
639.0125 638.1375 636.32916667 634.9875 633.90833333
635.80416667 638.51666667 641.55 642.36666667 641.22916667
642.075 644.7 648.375 653.45 659.51666667
665.02916667 672. 678.06666667 683.55 690.43333333
696.79166667 701.37083333 704.1125 706.35833333 711.1125
717.5 722.8375 728.35 732.95833333 735.72916667
737.8875 738.0625 737.85833333 739.4625 742.72916667
744.8 744.15833333 744.45 744.30416667 741.53333333
738.03333333 735.90416667 734.5625 732.2 729.63333333
728.58333333 726.57083333 725.08333333 725.34583333 722.8375
719.1625 718.34583333 717.26666667 714.2625 712.36666667
710.26666667 709.3625 711.05416667 712.425 712.22083333
712.80416667 715.6625 719.62916667 722.72083333 726.075
731.325 734.67916667 738.9375 743.34166667 743.60416667
743.72083333 744.91666667 747.01666667 749.7875 752.2375
754.075 755.3875 756.29166667 758.21666667 761.075
764.4 768.36666667 772.04166667 777.90416667 782.075
784.11666667 786.85833333 790.475 794.90833333 799.225
803.74583333 806.86666667 805.64166667 802.725 802.4625
800.59583333 799.8375 803.25 805.7 805.81666667
805.99166667 804.76666667 803.89166667 804.65 807.5375
812.55416667 813.6625 815.4125 818.88333333 818.65
817.97916667 818.41666667 816.52083333 815.90833333 816.92916667
816.2875 816.9 816.725 815.7625 814.625
812.7 812.02916667 813.16666667 815.4125 819.11666667
823.2875 823.57916667 820.60416667 817.45416667 815.90833333
817.22083333 820.8375 824.1625 827.4 830.55
832.85416667 836.675 839.825 842.30416667 848.77916667
857.325 864.64583333 871.73333333 879.05416667 885.90833333
890.95416667 894.8625      nan      nan      nan
      nan      nan      nan]
```

```
In [17]: print(pal.seasonal)
```

```
[0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445]
```

```

1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
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0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315 1.02505042 0.98307259 1.02505984 0.98940811 1.07755577
0.82382161 0.76156155 0.92124624 1.05494351 1.15456277 1.11418445
1.06953315]

```

```
In [18]: print(pal.resid)
```

```

[      nan      nan      nan      nan      nan      nan
 1.00827969 1.02396061 0.98323503 0.98167711 0.96655699 1.01391922
 1.04688528 1.01761425 1.05002576 1.03214748 0.97896444 0.94011649
 0.976082   0.96699759 0.96925517 1.0380475   0.98315299 1.06571409
 0.94823701 0.98408436 1.01312037 1.03631647 0.96413921 0.98092987
 0.98437851 0.98047061 1.02543552 1.01317632 1.0158042   1.07560501
 0.98507252 0.96793962 0.96929201 0.98936716 0.99584777 0.9976436
 1.00853568 1.02231393 1.02772462 0.99480713 1.00587241 1.00961573
 0.98406022 0.98915703 1.04602013 0.98011519 0.99045097 1.02450846
 0.99568532 0.9738434   0.99961938 1.0020803   1.00757591 0.98947523
 0.98065426 1.01596783 1.00744693 1.02281787 1.01705586 0.98449764
 0.98261903 1.01417767 0.96549589 0.97319572 1.00942324 0.94856329
 1.00953695 1.0517874   1.00987621 0.97990093 1.0200025   0.99476181
 1.00725942 0.99457939 1.00912071 1.01481937 0.98497594 1.01195903
 0.99740281 0.98489481 0.99382396 0.96334503 1.03151815 0.99967225
 1.01632786 0.98030182 0.99516057 0.97705247 0.9951003   0.99913669
 1.03558779 1.00931009 1.00750189 1.02367678 0.97912118 0.99121434
 0.99029726 1.01660895 1.01067479 1.02442107 1.03226611 1.00768685
 0.92489215 0.96956228 1.00213222 0.95643805 1.01439067 1.04860022
 0.99931906 0.99590713 1.01076487 0.95723305 1.03244862 0.98714374
 1.0159246   1.03437543 0.95850033 1.01901022 1.01749795 1.00748236
 0.99667893 0.9941658   0.94793496 0.99839604 1.02898677 0.98799242
 1.05328482 0.99760573 0.95824568 0.97602616 0.99161419 1.00294825
 1.0264278   1.02780283 1.02291871 1.03659635 0.97604579 0.92488568
 0.97906386 0.97450657 0.98081991 1.01764406 0.99620251 1.02443009

```

```
1.00464862 1.00540945 1.02919898 0.98503676 0.95832989 0.97484221
1.0359369      nan      nan      nan      nan      nan
      nan]
```

```
In [19]: print(pal.observations)
```

```
[ 511.7  468.3  571.9  648.2  705.6  709.1  676.9  661.5  611.8  640.5
   611.1  697.2  548.8  492.1  613.2  692.3  721.7  672.  670.6  635.6
   611.8  686.  630.7  750.4  515.2  498.4  627.2  741.3  760.9  754.6
   733.6  704.9  709.8  733.6  714.7  831.6  586.6  536.9  654.5  767.9
   848.4  820.4  795.9  774.9  750.4  759.5  740.6  809.9  603.4  558.6
   711.2  760.9  840.  835.8  777.  727.3  714.  744.8  723.1  770.7
   581.  555.8  665.7  770.7  836.5  779.1  745.5  739.2  676.2  710.5
   711.9  731.5  598.5  578.9  675.5  756.  865.2  819.  800.8  758.1
   737.8  774.9  728.  817.6  618.1  565.6  691.6  768.6  903.  847.7
   830.9  772.1  755.3  779.1  770.  844.2  671.3  607.6  737.8  863.1
   908.6  891.1  853.3  836.5  797.3  840.7  816.9  872.2  613.9  595.
   744.1  812.  941.5  940.1  863.1  829.5  808.5  800.1  836.5  870.8
   684.6  644.7  721.  877.1  959.7  916.3  870.8  832.3  760.2  833.7
   827.4  864.5  705.6  619.5  723.1  847.7  942.9  917.  897.4  859.6
   821.8  872.2  795.9  824.6  669.9  618.1  756.  901.6  968.8  968.8
   921.2  891.1  882.  887.6  840.  935.9  763.7  700.  844.2  989.1
  1045.8 1012.9  970.9]
```