

Seasonality and pricing – The case of Danish holiday cottages

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Overview

1. Introduction
2. Literature / causes / model of seasonality
3. Seasonality and holiday cottages rentals in Denmark
4. Seasonality and pricing of holiday cottage weeks
5. Summary / Discussion

Purpose

The purpose of the paper is to model seasonality in the demand for Danish cottage rentals, and to apply the model for pricing.

Causes of seasonality (~Bar-On, 1975)

Natural:

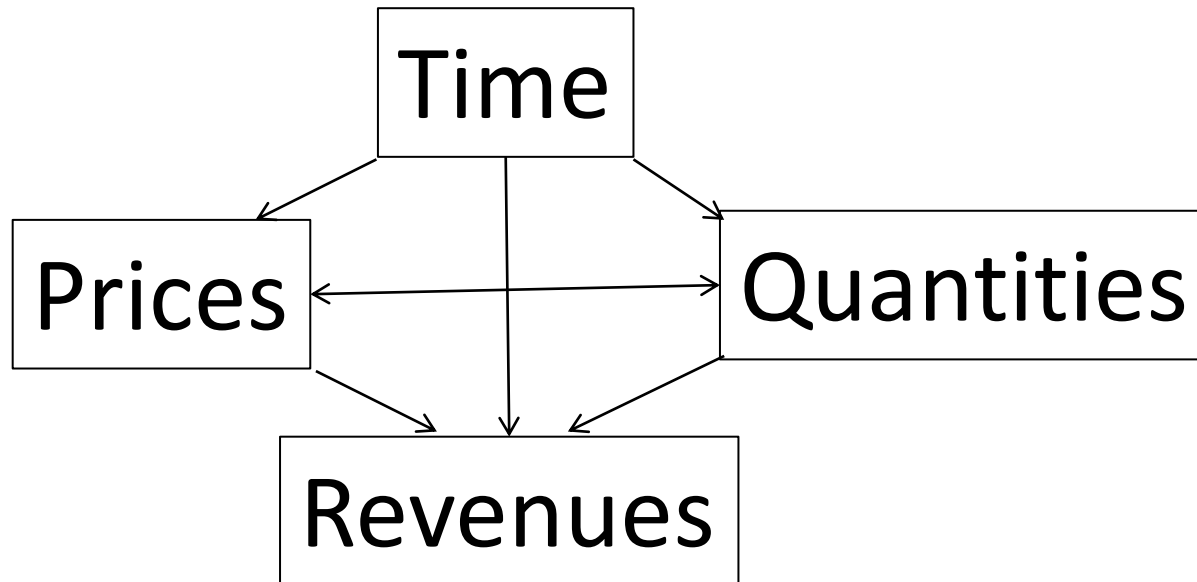
- Temperature (normal rather than actual)
- Hours of sunshine (rather than hours of daylight)
- Latitude and altitude
- Climate, rain/snow fall
- Snow depth

Institutional:

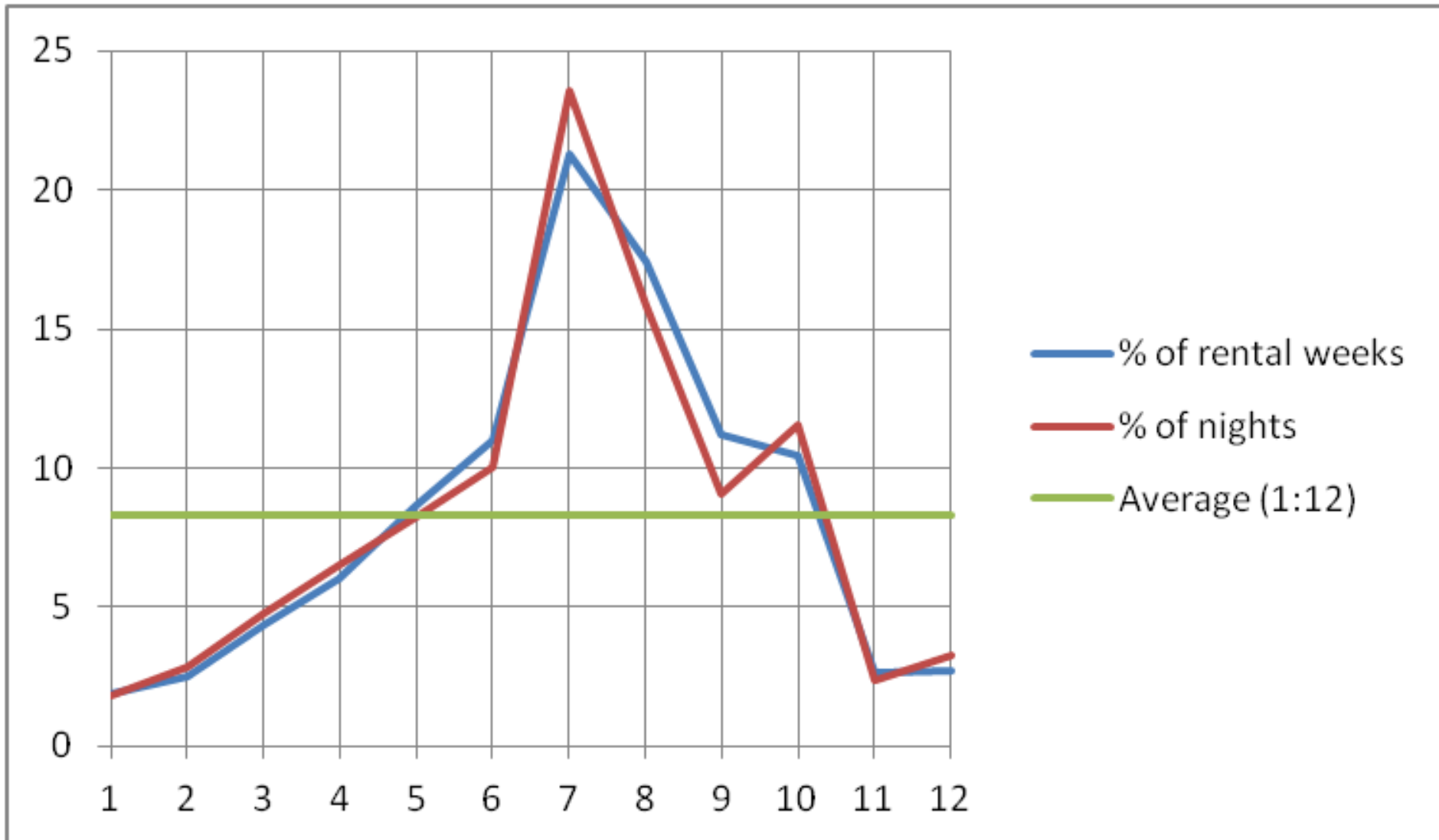
- School holidays (moving, in some key markets)
- Religious holidays (moving, in the spring)
- Calendar effects (number of days per month)
- Business seasons

Generic and diamond model of revenues:

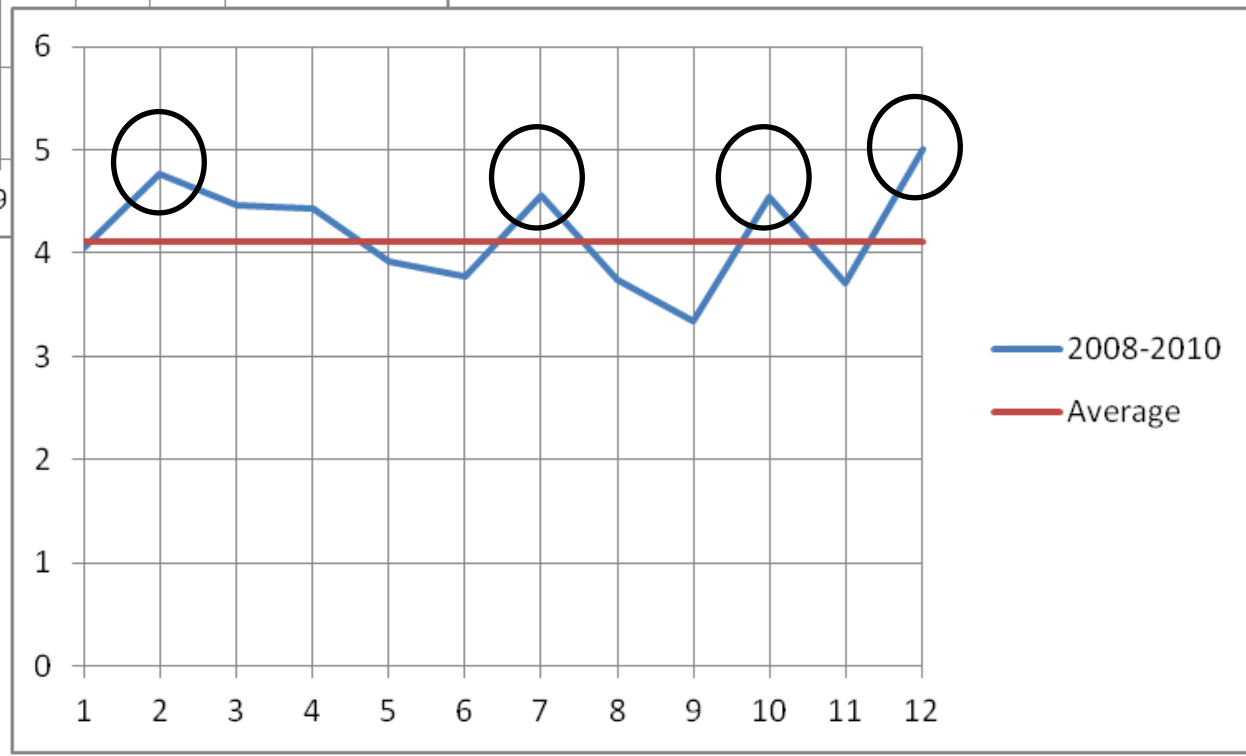
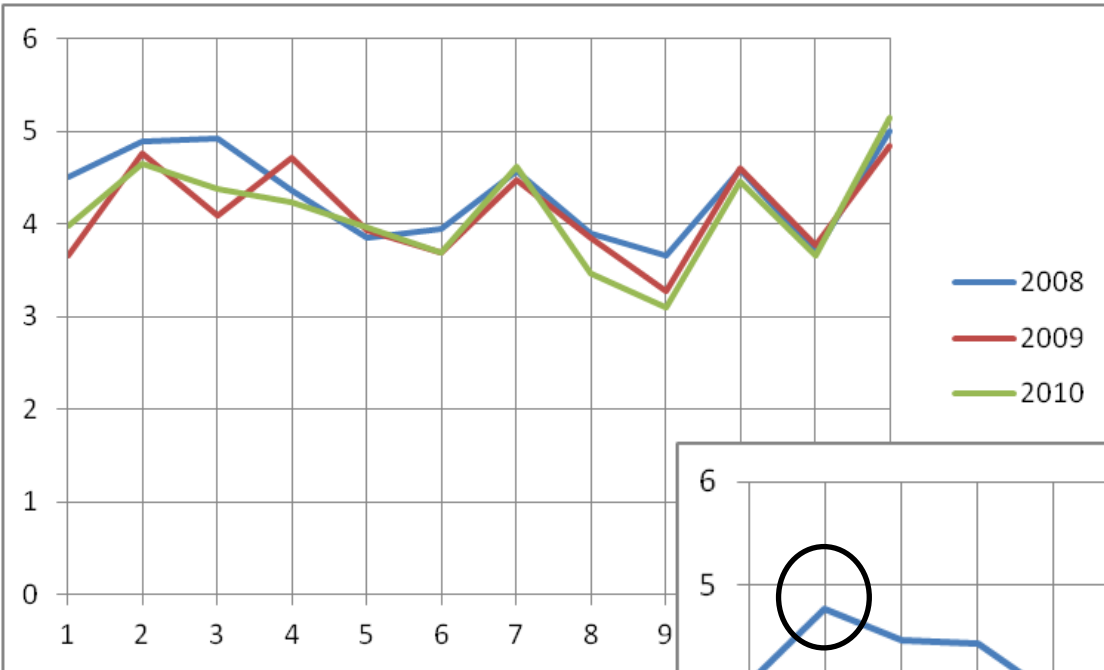
Sum of revenues per time unit = sum of (price * quantity)



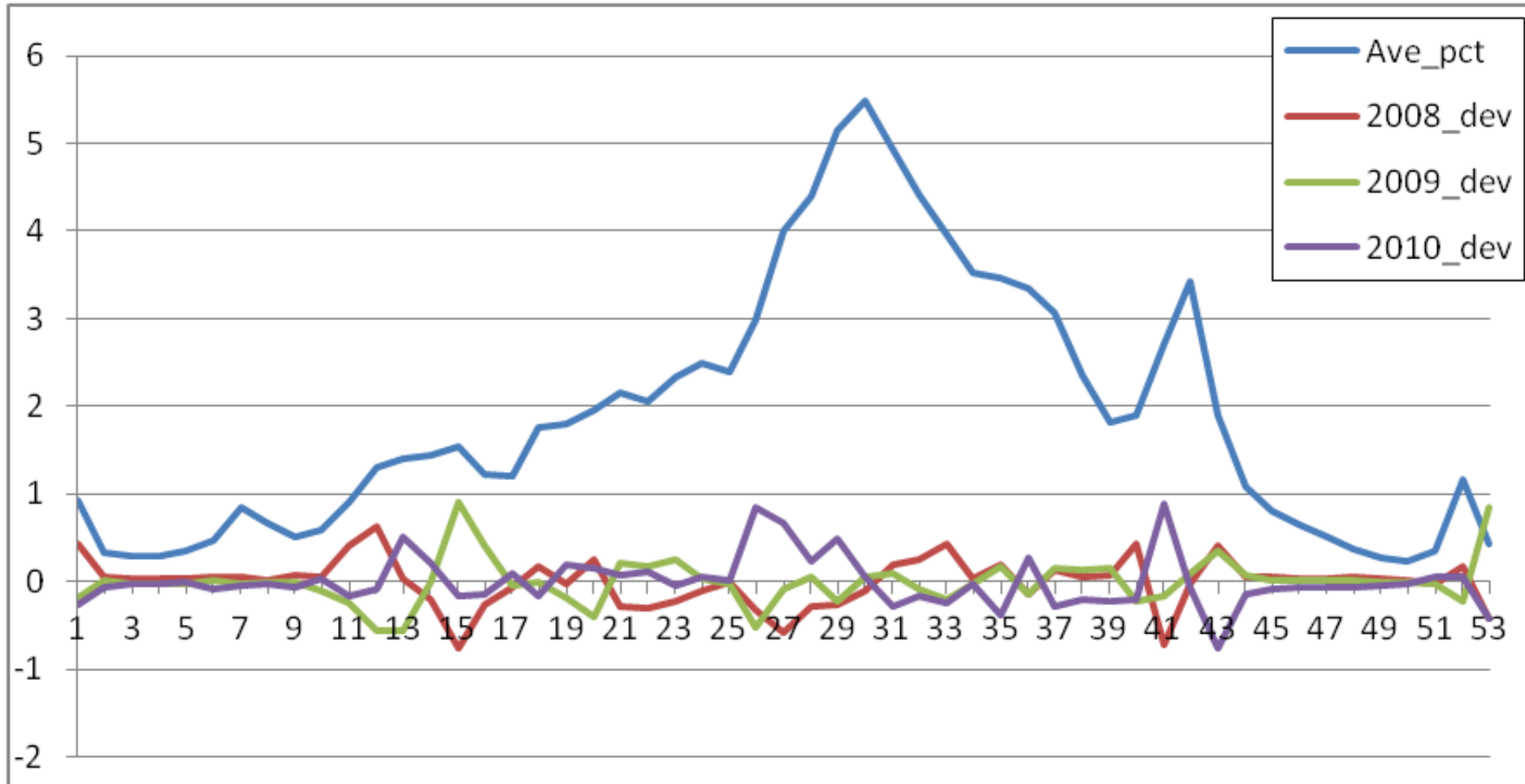
Percent of bednights and percent of rental weeks per month 2008-2010



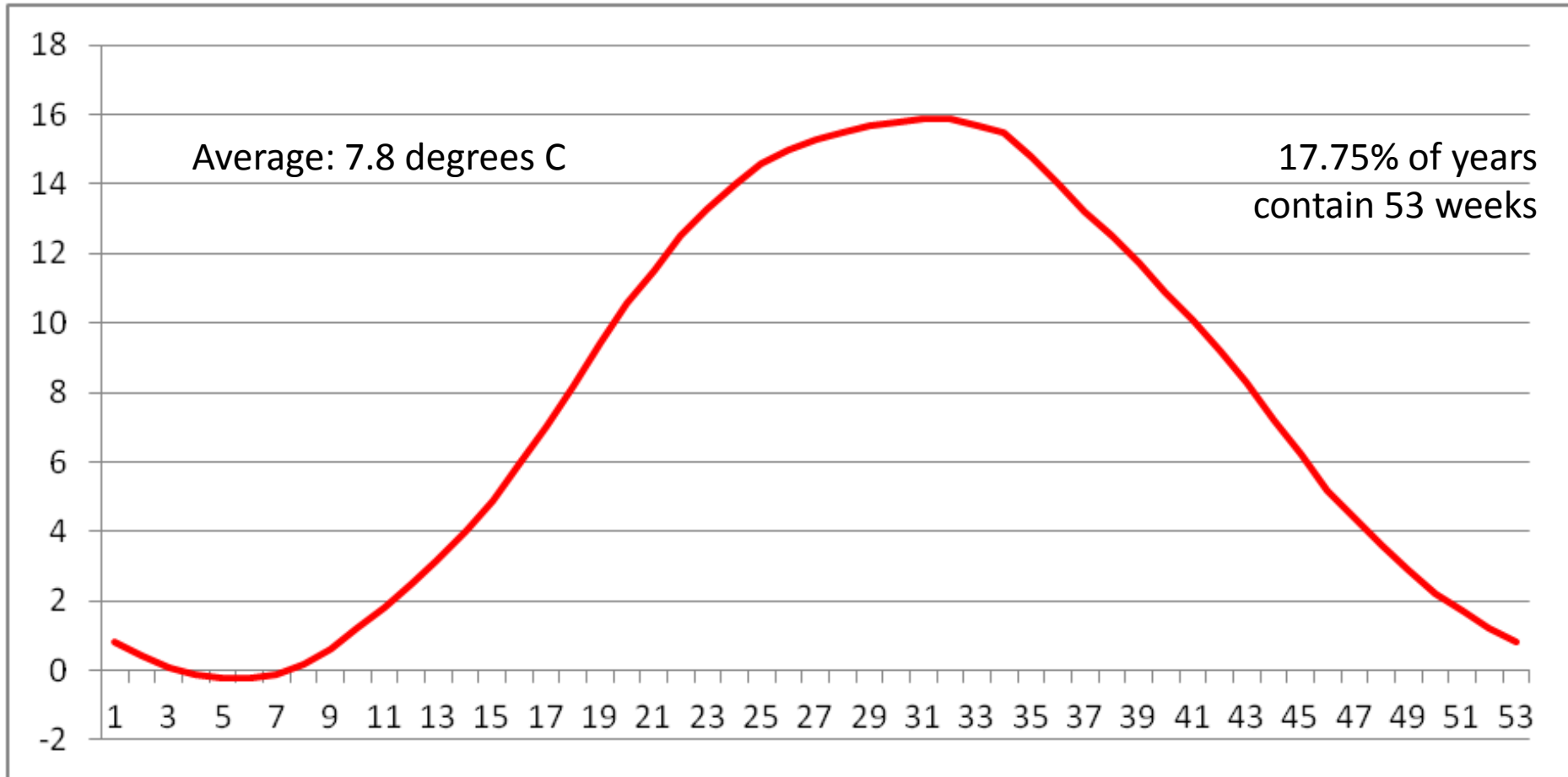
Persons per holiday cottage per month



Rentals of holiday cottages per week in Denmark, 2008-2010



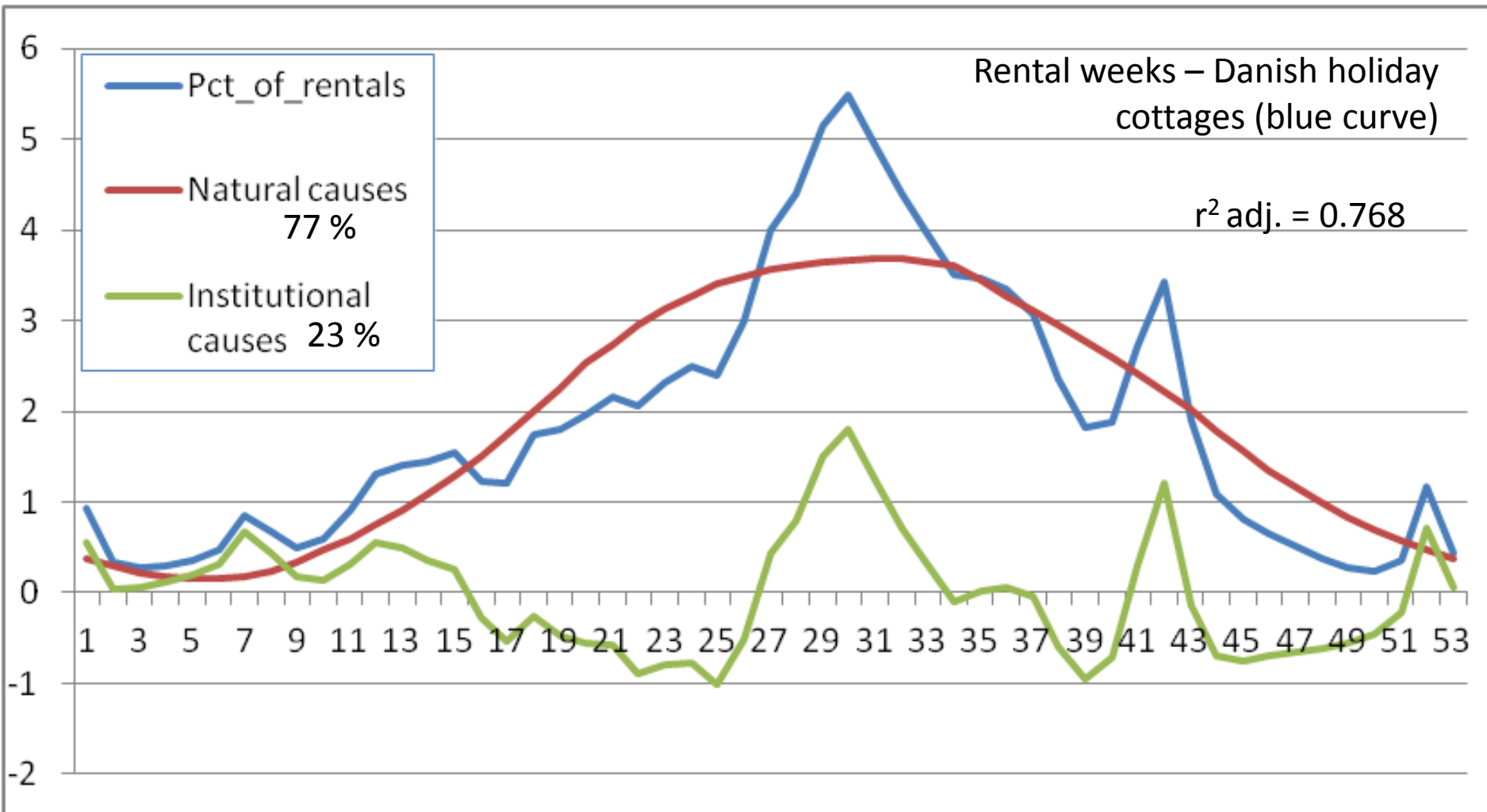
Normal* temperatures per week in Denmark



* Average historic highest daily temperatures.

Natural causes of seasonality: 77%

Institutional causes: 23%



Moving holidays, spring. DK

Easter:

1. (Wednesday before) **Maundy Thursday**
2. (Maundy Thursday before) **Good Friday**
3. (Easter Sunday before) **Easter Monday**

4. (Thursday before) **Prayer Day** (Friday) *
5. (Wednesday before) **Ascension Day** (Thursday)
6. (Whit - Sunday - before) **second Whit** (Monday)

* Prayer Day is an official Danish public holiday that falls on the fourth Friday after Easter, and thus three weeks before Pentecost (a.k.a. Whit)

Moving summer holidays, Germany: e.g., Niedersachsen

| Week\year | 2008 | 2009 | 2010 | 2011 | 2012 | Total |
|-----------|------|------|------|------|------|-------|
| 25 | 0 | 0 | 4 | 0 | 0 | 4 |
| 26 | 0 | 4 | 7 | 0 | 0 | 11 |
| 27 | 0 | 7 | 7 | 4 | 0 | 18 |
| 28 | 4 | 7 | 7 | 7 | 0 | 25 |
| 29 | 7 | 7 | 7 | 7 | 0 | 28 |
| 30 | 7 | 7 | 7 | 7 | 7 | 35 |
| 31 | 7 | 7 | 3 | 7 | 7 | 31 |
| 32 | 7 | 3 | 0 | 7 | 7 | 24 |
| 33 | 7 | 0 | 0 | 3 | 7 | 17 |
| 34 | 3 | 0 | 0 | 0 | 7 | 10 |
| 35 | 0 | 0 | 0 | 0 | 5 | 5 |
| Total | 42 | 42 | 42 | 42 | 40 | 208 |

Explanatory variables for percentage of rental per week

| | | |
|-------------------|---------|-------------|
| (Constant) | Week_30 | Week_37 |
| | Week_41 | Week_28 |
| 1. Temp_norm | Week_36 | Week_27 |
| 2. Sum_Niedersach | Week_35 | Week_34 |
| 3. Week_42 | Week_32 | Easter_2nd |
| 4. Easter_3_days | Week_31 | Palmesondag |
| 5. New_Years_eve | Week_29 | Week_1 |
| | Week_33 | Week_43 |
| | | Week_52 |

R2 adjusted=0.956

The model – R2 adjusted=0.956

Coefficients^a

| Model | | Unstandardized Coefficients | | Standardized Coefficients | t | Sig. | Collinearity Statistics | |
|-------|----------------|-----------------------------|------------|---------------------------|--------|------|-------------------------|-------|
| | | B | Std. Error | Beta | | | Tolerance | VIF |
| 1 | (Constant) | .270 | .047 | | 5.804 | .000 | | |
| | Temp_norm | .147 | .006 | .587 | 24.209 | .000 | .478 | 2.091 |
| | Sum_Niedersach | .104 | .023 | .151 | 4.601 | .000 | .260 | 3.848 |
| | Week_27 | .993 | .209 | .093 | 4.744 | .000 | .731 | 1.369 |
| | Week_28 | 1.219 | .224 | .114 | 5.433 | .000 | .636 | 1.571 |
| | Week_29 | 1.845 | .237 | .173 | 7.771 | .000 | .568 | 1.761 |
| | Week_30 | 2.164 | .238 | .203 | 9.110 | .000 | .568 | 1.762 |
| | Week_31 | 1.741 | .221 | .163 | 7.888 | .000 | .657 | 1.522 |
| | Week_32 | 1.454 | .199 | .136 | 7.308 | .000 | .809 | 1.236 |
| | Week_33 | 1.144 | .193 | .107 | 5.928 | .000 | .860 | 1.162 |
| | Week_34 | .865 | .189 | .081 | 4.579 | .000 | .897 | 1.115 |
| | Week_35 | 1.029 | .188 | .096 | 5.470 | .000 | .906 | 1.104 |
| | Week_36 | 1.016 | .187 | .095 | 5.443 | .000 | .919 | 1.088 |
| | Week_37 | .854 | .185 | .080 | 4.604 | .000 | .931 | 1.074 |
| | Week_41 | .966 | .182 | .091 | 5.313 | .000 | .969 | 1.032 |
| | Week_42 | 1.802 | .181 | .169 | 9.949 | .000 | .977 | 1.024 |
| | Week_43 | .404 | .181 | .038 | 2.236 | .027 | .982 | 1.018 |
| | Week_52 | .542 | .186 | .051 | 2.915 | .004 | .928 | 1.078 |
| | New_Years_eve | .556 | .147 | .067 | 3.790 | .000 | .903 | 1.107 |
| | Palmesondag | .657 | .181 | .062 | 3.633 | .000 | .980 | 1.021 |
| | Easter_3_days | 1.314 | .180 | .123 | 7.285 | .000 | .984 | 1.016 |
| | Easter_2nd | .665 | .180 | .062 | 3.693 | .000 | .987 | 1.013 |

a. Dependent Variable: Pct_of_rentals

Y=Pct of rentals

Application of the model for future price decision support. R2 adj.=0.929

Coefficients^a

| Model | | Unstandardized Coefficients | | Standardized Coefficients | t | Sig. | Collinearity Statistics | |
|-------|---------------|-----------------------------|------------|---------------------------|--------|------|-------------------------|-------|
| | | B | Std. Error | Beta | | | Tolerance | VIF |
| 1 | (Constant) | 71.951 | .848 | | 84.873 | .000 | | |
| | Temp_norm | 2.445 | .099 | .516 | 24.687 | .000 | .625 | 1.599 |
| | New_Years_eve | 65.242 | 3.394 | .327 | 19.225 | .000 | .942 | 1.061 |
| | Week_27 | 31.079 | 3.409 | .156 | 9.117 | .000 | .934 | 1.071 |
| | Week_28 | 41.181 | 3.414 | .207 | 12.062 | .000 | .931 | 1.074 |
| | Week_29 | 52.470 | 3.419 | .263 | 15.346 | .000 | .928 | 1.077 |
| | Week_30 | 52.176 | 3.422 | .262 | 15.246 | .000 | .927 | 1.079 |
| | Week_31 | 51.932 | 3.425 | .260 | 15.163 | .000 | .925 | 1.081 |
| | Week_32 | 49.163 | 3.425 | .247 | 14.352 | .000 | .925 | 1.081 |
| | Week_33 | 35.210 | 3.421 | .177 | 10.292 | .000 | .927 | 1.079 |
| | Week_34 | 22.425 | 3.416 | .112 | 6.564 | .000 | .930 | 1.075 |
| | Week_52 | 18.674 | 3.388 | .094 | 5.512 | .000 | .945 | 1.058 |
| | Palmesondag | 12.430 | 3.320 | .062 | 3.744 | .000 | .985 | 1.016 |
| | Easter_3_days | 15.074 | 3.313 | .076 | 4.550 | .000 | .989 | 1.011 |
| | Easter_2nd | 12.824 | 3.307 | .064 | 3.878 | .000 | .992 | 1.008 |

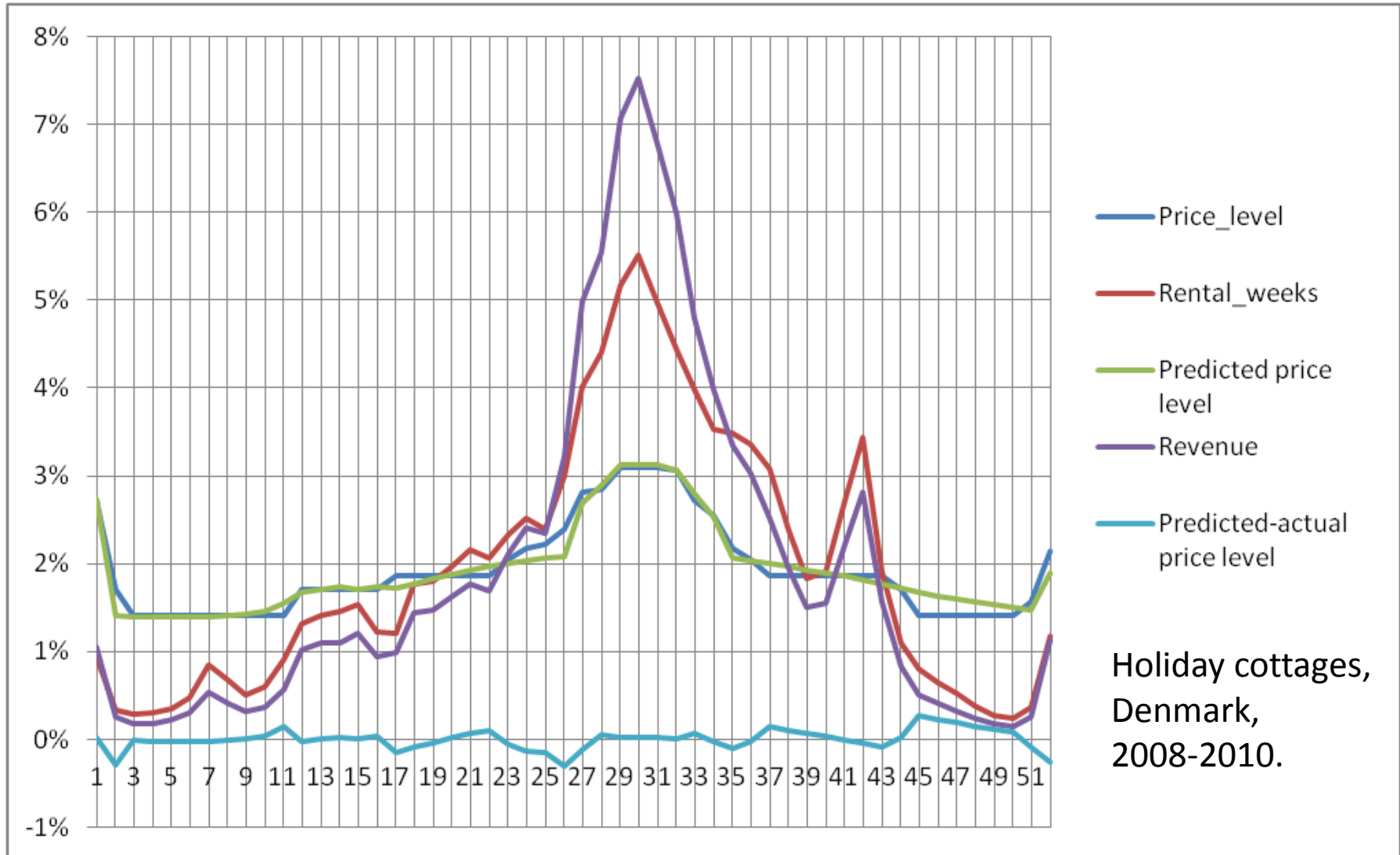
a. Dependent Variable: Price_level
Y=Price_level

Note: Excludes week 53.

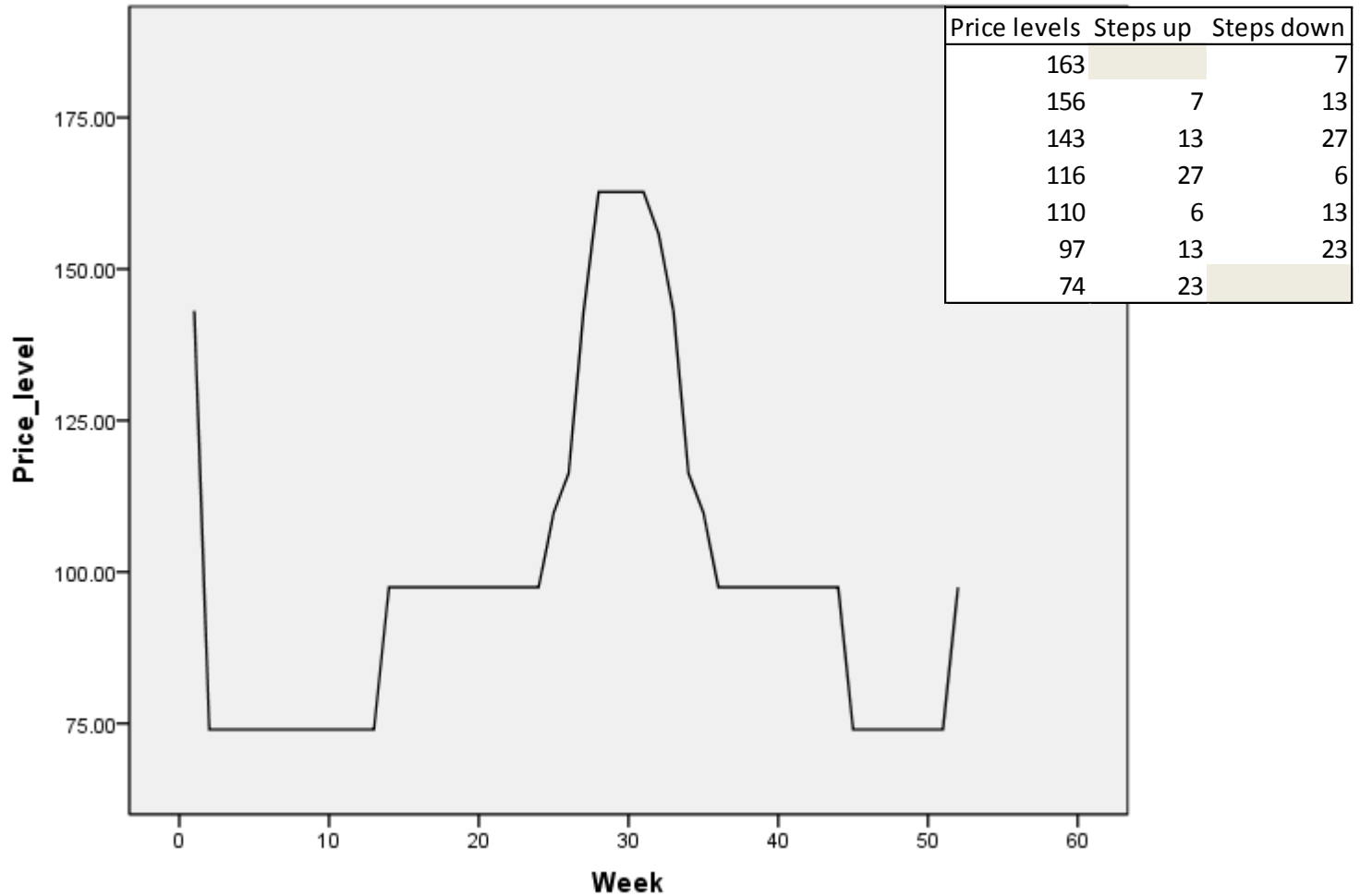
Eliminated in a stepwise multiple regression analysis:

Week_35, Week_36, Week_37, Week_41, Week_42, Week_43

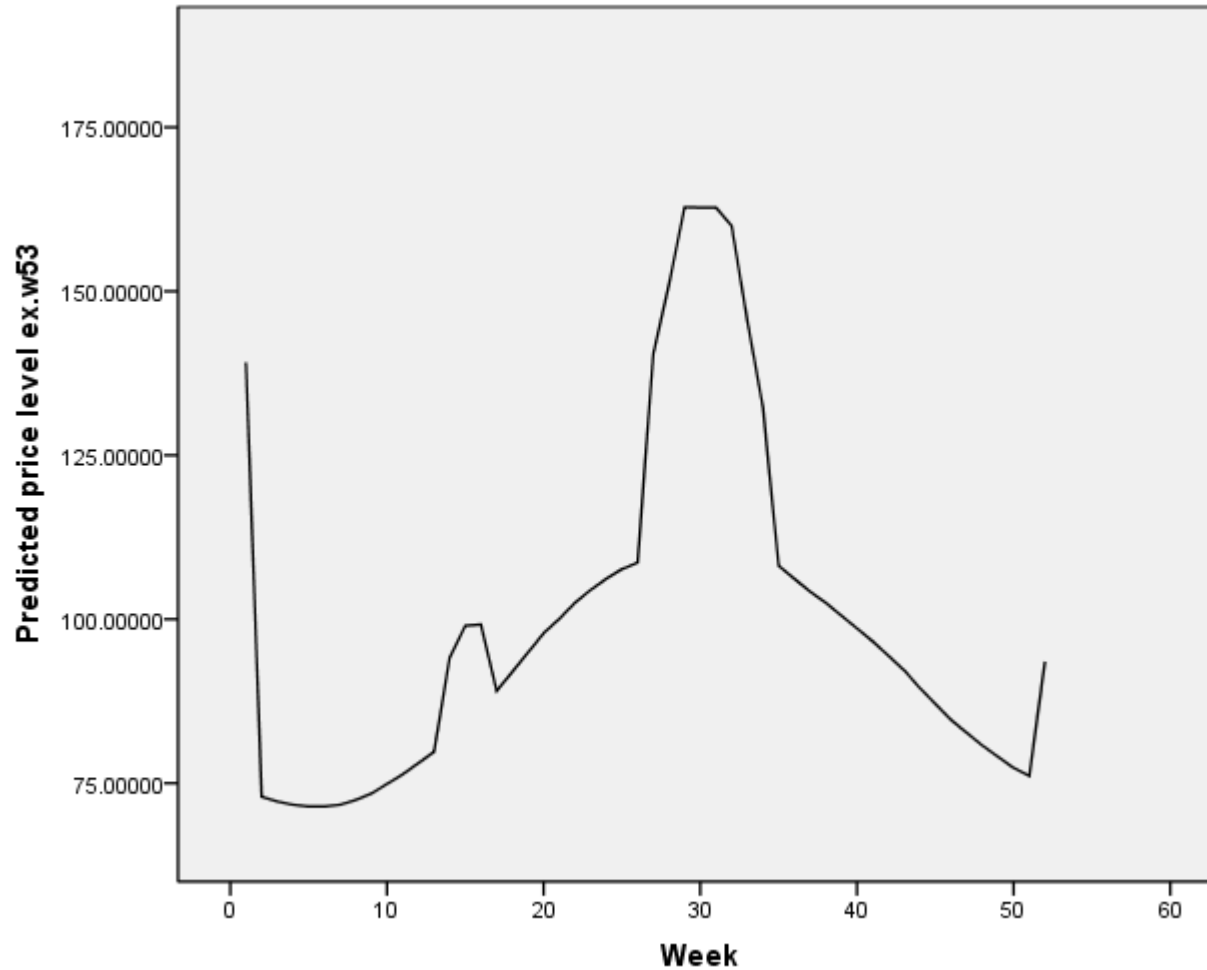
Both seasonality in quantities demanded and in prices contribute to seasonality in revenue. Recommended prices can be predicted. Probably, the same could be done for other tourism services.



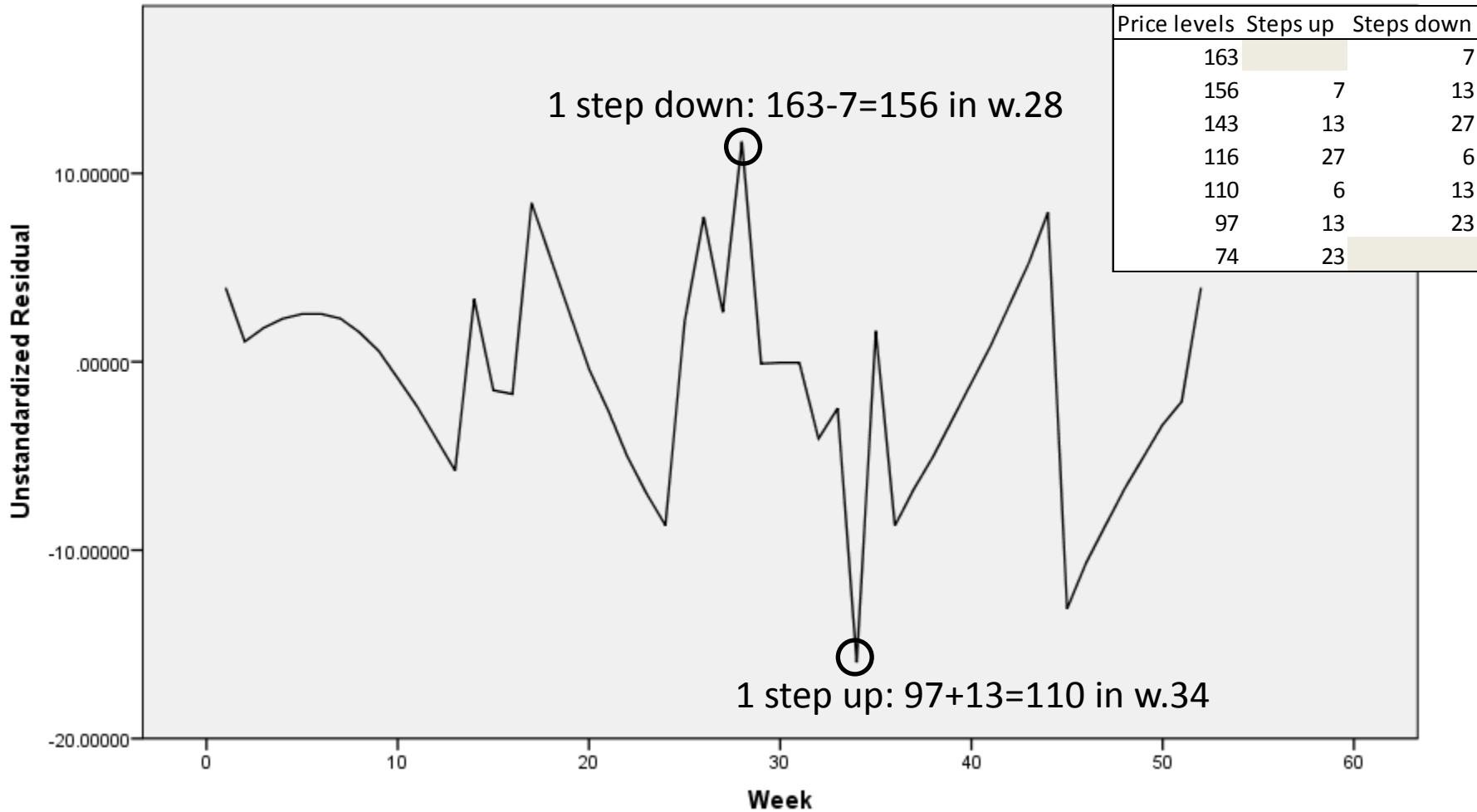
Planned price-level for 2012



Predicted = suggested price level 2012



Deviations (un.std.) between published planned price-level versus predicted price - 2012



Conclusions / Summary

- It is possible to predict the future distribution of cottage renting weeks, and other lodging services, with known normal temperatures, calendar and moving holidays.
- Application of a model as a decision support tool for pricing: 1. Explain seasonal demand. – 2. Explain price. 3. Predict (recommend) price, given calendar etc.. 4. ...
- Temperature is a key driver for demand in leisure tourism.
- Consider taking the holiday periods, and any known changes in these of main markets, into account when setting seasonal prices or predicting seasonal demand.