# Temporal Aggregation of (Growth) Timeseries

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### 1 Introduction

This vignette discusses temporal aggregation of timeseries, i.e. the conversion of a timeseries to a lower frequency. The vignette is an extension of the vignette "Introduction to regts", but can be read independently.

Section 2 describes the standard aggregation function aggregate. Section 3 discusses aggregation for growth timeseries with several methods, which are described in detail in section 4.

## 2 Function aggregate

The standard function aggregate can be used to convert a timeseries to a lower frequency, for example a quarterly to an annual timeseries, or a monthly to a quarterly timeseries.

An example with a monthly timeseries:

```
> regm <- regts(1:24, start = "2016M1")</pre>
> regm
     Jan Feb Mar Apr May Jun Jul Aug Sep Oct Nov Dec
2016
                               7
                       5
                           6
                                   8
                                        9
                                          10
                                               11
     13 14
             15 16 17
2017
                         18
                             19
                                  20 21
> aggregate(regm, FUN = mean, nfrequency = 4)
     Qtr1 Qtr2 Qtr3 Qtr4
2016
        2
                  8
             5
                      11
```

The observations in the timeseries are split into blocks for each quarter and FUN is applied to the three observations in that quarter. The result is a quarterly timeseries holding the aggregated values.

Another example with a quarterly timeseries:

20

17

2017

```
> regq <- regts(1:10, start = "2016q1")
> regq
```

```
> aggregate(regq, FUN = sum)
```

+1 2016 10 26

The quarters in 2016 add up to 10, the quarters in 2017 to 26. Note that the two quarters in 2018 are ignored. The aggregate function skips the incomplete years at the end.

Incomplete years at the beginning are also skipped. For example, the next timeseries starts in the second quarter:

#### 2017 26

The quarters in 2017 add up to 26 as before. But now the information for 2016 is also not complete, and therefore ignored.

Thus for regts objects the aggregate function skips all incomplete years at the beginning and end of the timeseries. However for standard timeseries (ts) the implementation of aggregate only skips incomplete years at the end. The previous example but now with a ts timeseries:

```
> tq2 <- ts(2:10, start = c(2016,2), frequency = 4)
> tq2
     Qtr1 Qtr2 Qtr3 Qtr4
2016
             2
                   3
                        4
2017
        5
             6
                   7
                        8
2018
        9
            10
> aggregate(tq2, FUN = sum)
```

```
Time Series:
Start = 2016.25
End = 2017.25
Frequency = 1
[1] 14 30
```

The result is an unconventional timeseries with frequency year but a shifted period (the start is 2016.25). The result for 2016.25 is the sum of the observations in period 2016Q2/2017Q1.

In all other cases the aggregate function works exactly the same for regts and ts objects.

### 3 Aggregation of growth timeseries

As is shown in the next section the function aggregate does not yield correct results for timeseries with absolute, relative or percentage changes. For growth timeseries the function aggregate\_gr can be used. It accepts both regts and ts timeseries. The function implements several methods (dif1, dif1s, rel and pct) for different types of input timeseries.

Methods dif1 and dif1s assume that the input timeseries is a first difference:

$$x_t = z_t - z_{t-1}$$

where t is the period (like year or quarter). These methods calculate the low frequency first difference timeseries. For method dif1s the result is also scaled. An example for both methods, the results are annual timeseries:

2020 16

```
> aggregate_gr(xdif, method = "dif1s") # result is also scaled
```

2020 4

The rel and pct methods assume that the input timeseries is a one-period relative or percentage change and calculate the exact low frequency relative or percentage change. The methods are described in detail in the next section.

# 4 Aggregation methods for growth timeseries

This section describes the methods implemented in function aggregate\_gr, including the derivation of the corresponding mathematical formulas.

Suppose we are temporally aggregating a timeseries x with n subperiods to a timeseries X with a lower frequency. In other words we want to convert a timeseries x with a high frequency (monthly, quarterly) to a new timeseries with a lower frequency (quarterly, annual).

Let  $x_{t,i}$  stand for the value of x in subperiod i of main period t. The index i takes on values in the range 1..n. We always interpret  $x_{t,0}$  to mean  $x_{t-1,n}$ . The time index t refers to the periods in the time domain of X. Many conversion methods only need the observations subperiod i of main period t. For example the mean method calculates  $X_t$  as follows

$$X_t = \sum_{i=1}^n x_{t,i}/n$$

However there are transformations which are not so straightforward. A case in point is when x is a high frequency first difference and must be converted to a low frequency first difference. A simple averaging of n high frequency observations will not give the correct answer. For example, for a quarterly growth series the yearly aggregates also depend on the quarterly observations of the previous year. Therefore we cannot use function aggregate, function aggregategr must be applied. In the next subsections a more detailed description of the available frequency conversion methods for this function is given.

### 4.1 dif1 method

Define x to be the first difference of z. Thus

$$x_{t,i} = z_{t,i} - z_{t,i-1}$$

We can also define  $z_{t,i}$  in terms of x as follows

$$z_{t,i} = z_{t,i-1} + x_{t,i}$$
  
=  $z_{t-1,n} + \sum_{j=1}^{i} x_{t,j}$ 

Now define  $Z_t$  as the level timeseries in time domain t corresponding to z

$$Z_{t} = \sum_{i=1}^{n} z_{t,i}$$

$$= \sum_{i=1}^{n} z_{t-1,n} + \sum_{i=1}^{n} \sum_{j=1}^{i} x_{t,j}$$

$$= \sum_{i=1}^{n} z_{t-2,n} + \sum_{i=1}^{n} \sum_{j=1}^{n} x_{t-1,j} + \sum_{i=1}^{n} \sum_{j=1}^{i} x_{t,j}$$

Then X in any period t can be calculated from

$$X_t = Z_t - Z_{t-1}$$

Since  $Z_{t-1}$  can be written as

$$Z_{t-1} = \sum_{i=1}^{n} z_{t-2,n} + \sum_{i=1}^{n} \sum_{j=1}^{i} x_{t-1,j}$$

it is easy to show that  $X_t$  does not depend on  $z_{t-2,n}$ . In any practical algorithm  $z_{t-2,n}$  can be set to 0.

A compact equation for  $X_t$  can be derived by substituting the equations for  $Z_t$  and  $Z_{t-1}$  into the definition of  $X_t$ 

$$X_{t} = \sum_{i=1}^{n} \left( \sum_{j=i+1}^{n} x_{t-1,j} + \sum_{j=1}^{i} x_{t,j} \right)$$

This expression can be futher simplified by changing the order of summation. For the first summation, we can write

$$\sum_{i=1}^{n} \sum_{j=i+1}^{n} x_{t-1,j} = \sum_{j=2}^{n} \sum_{i=1}^{j-1} x_{t-1,j} = \sum_{j=2}^{n} (j-1)x_{t-1,j}$$

Similarly,

$$\sum_{i=1}^{n} \sum_{j=1}^{i} x_{t,j} = \sum_{i=1}^{n} \sum_{j=1}^{n} x_{t,j} = \sum_{i=1}^{n} (n-j+1)x_{t,j}$$

The final equation is given by

$$X_{t} = \left(\sum_{j=2}^{n} (j-1)x_{t-1,j} + \sum_{j=1}^{n} (n-j+1)x_{t,j}\right)$$

#### 4.2 dif1s method

The dif1s method is similar to the dif1 method described in the previous subsection. The only difference is that the aggregate  $Z_t$  is calculated with an average instead of a sum:

$$Z_t = \sum_{i=1}^n z_{t,i}/n$$

Further derivations are analogous. The result for dif1s is the result for dif1 divided by n:

$$X_{t} = \left(\sum_{j=2}^{n} (j-1)x_{t-1,j} + \sum_{j=1}^{n} (n-j+1)x_{t,j}\right)/n$$

### 4.3 rel method

Define x to be the relative change in z. Thus

$$x_{t,i} = (z_{t,i} - z_{t,i-1})/z_{t,i-1}$$

We can also define  $z_{t,i}$  in terms of x as follows

$$z_{t,i} = z_{t,i-1}(1+x_{t,i})$$

$$= z_{t-1,n} \prod_{j=1}^{i} (1+x_{t,j})$$

$$= z_{t-2,n} \prod_{j=1}^{n} (1+x_{t-1,j}) \prod_{j=1}^{i} (1+x_{t,j})$$

Now define  $Z_t$  as the level timeseries in time domain t corresponding to z

$$Z_{t} = \sum_{i=1}^{n} z_{t,i}$$

$$= z_{t-1,n} \sum_{i=1}^{n} \prod_{j=1}^{i} (1 + x_{t,j})$$

$$= z_{t-2,n} \prod_{j=1}^{n} (1 + x_{t-1,j}) \sum_{i=1}^{n} \prod_{j=1}^{i} (1 + x_{t,j})$$

Then X in any period t can be calculated from

$$X_t = Z_t/Z_{t-1} - 1$$

Using the expression for  $Z_{t-1}$ 

$$Z_{t-1} = z_{t-2,n} \sum_{i=1}^{n} \prod_{i=1}^{i} (1 + x_{t-1,j})$$

we obtain

$$X_{t} = \frac{\prod_{j=2}^{n} (1 + x_{t-1,j}) \sum_{i=1}^{n} \prod_{j=1}^{i} (1 + x_{t,j})}{1 + \sum_{i=2}^{n} \prod_{j=2}^{i} (1 + x_{t-1,j})} - 1$$

### 4.4 pct method

Define x to be the percentage change in z. Thus

$$x_{t,i} = 100(z_{t,i} - z_{t,i-1})/z_{t,i-1}$$

The derivations are analogous to the rel method described in the previous subsection.

The result is

$$X_{t} = 100 \frac{\prod_{j=2}^{n} (1 + x_{t-1,j}/100) \sum_{i=1}^{n} \prod_{j=1}^{i} (1 + x_{t,j}/100)}{1 + \sum_{i=2}^{n} \prod_{j=2}^{i} (1 + x_{t-1,j}/100)} - 1$$