

# 1 TSFA Guide

This guide illustrates the steps for estimating a factor model using as an example the data and process which led to results reported in Gilbert and Meijer (2006). The background theory is reported in Gilbert and Meijer (2005).

Plots and some output from the examples below are omitted to save paper. The graphics parameter setting for margins have been adjusted for new versions of `tfplot()` and are not those used in the original paper. (Original settings no longer give the same result.)

The functions in the *tsfa* package are made available with

```
> library("tsfa")
```

The code from the vignette that generates this guide can be loaded into an editor with `edit(vignette("Guide", package="tsfa"))`. This uses the default editor, which can be changed using `options()`. In some examples the code may run into the margins and is truncated in the pdf. If in doubt about the code, please edit the vignette as above or consult its source, which is distributed in the package.

The data is converted to real per capita data as follows

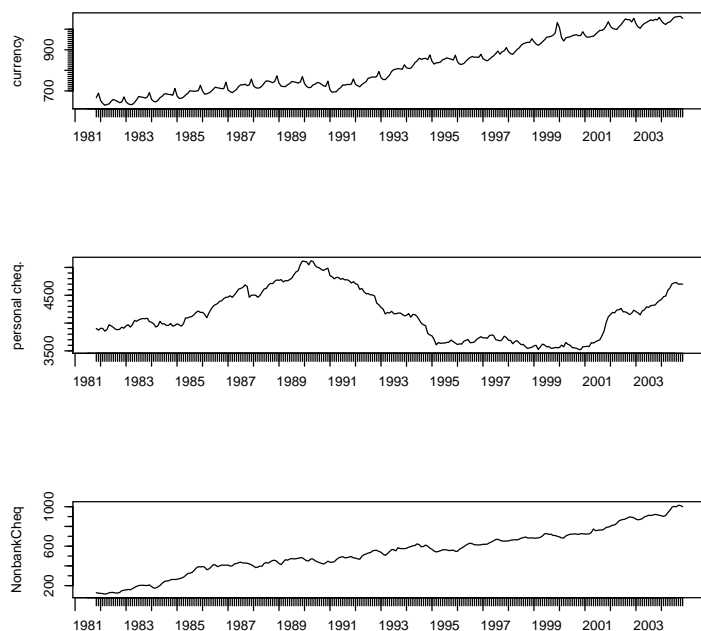
```
> data("CanadianMoneyData.asof.28Jan2005", package="CDNmoney")
> data("CanadianCreditData.asof.28Jan2005", package="CDNmoney")
> cpi <- 100 * M1total / M1real
> seriesNames(cpi) <- "CPI"
> popm <- M1total / M1PerCapita
> seriesNames(popm) <- "Population of Canada"
> z <- tframed(tbind(
  MB2001,
  MB486 + MB452 + MB453 ,
  NonbankCheq,
  MB472 + MB473 + MB487p,
  MB475,
  NonbankNonCheq + MB454 + NonbankTerm + MB2046 + MB2047 + MB2048 +
  MB2057 + MB2058 + MB482),
  names=c("currency", "personal cheq.", "NonbankCheq",
    "N-P demand & notice", "N-P term", "Investment" )
)
> TotalMoney <- tframed(rowSums(z), tframe(z))
> z <- tbind (z, ConsumerCredit, ResidentialMortgage,
  ShortTermBusinessCredit, OtherBusinessCredit)
```

Investment series goes back only to November 1981 and some of the data ends in November 2004, so the data is truncated to that time window.

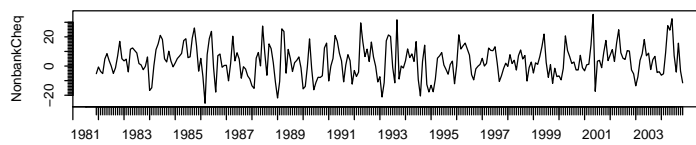
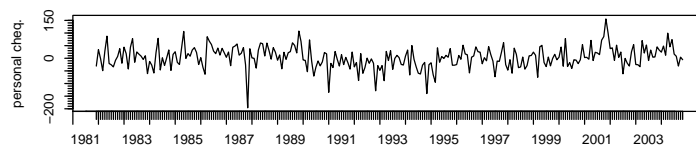
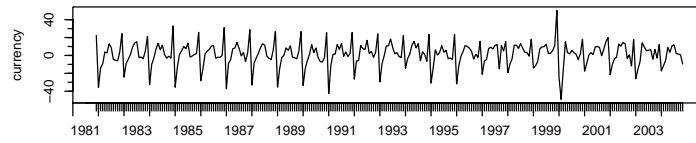
```
> z <- tfwindow(z, start=c(1981,11), end=c(2004,11))
> scale <- tfwindow(1e8 /(popm * cpi), tf=tframe(z))
> MBandCredit <- sweep(z, 1, scale, "*")
```

Multiplying by  $1e8$  gives real dollars per person. (Credit aggregates, B and MB numbers are in millions, CPI is in percentage points, popm is in units.) Plots to check the data can be generated with

```
> tfplot(MBandCredit, graphs.per.page=3)
```



```
> tfplot(diff(MBandCredit), graphs.per.page=3)
```



Various sample statistics are checked with

```
> start(MBandCredit)

[1] 1981    11

> end(MBandCredit)

[1] 2004    11

> Tobs(MBandCredit)

[1] 277

> DX <- diff(MBandCredit, lag=1)
```

The number of observations is

```
> Tobs(MBandCredit)

[1] 277
```

The number of series is

```
> nseries(MBandCredit)
```

```
[1] 10
```

The means of differenced series are

```
> colMeans(DX)
```

currency	personal cheq.
1.398385	2.880048
NonbankCheq	N-P demand & notice
3.157783	9.156883
N-P term	Investment
4.926683	48.744141
Consumer Credit	Residential Mortgage
12.703423	29.878367
Short Term Business Credit	Other Business Credit
-8.066348	37.283246

The standard deviations of differenced series are

```
> sqrt(diag(cov(DX)))
```

currency	personal cheq.
13.46788	42.94902
NonbankCheq	N-P demand & notice
10.74835	46.04252
N-P term	Investment
71.93040	116.98518
Consumer Credit	Residential Mortgage
30.00178	50.32996
Short Term Business Credit	Other Business Credit
71.83893	56.30949

## 2 Checking for the number of factors

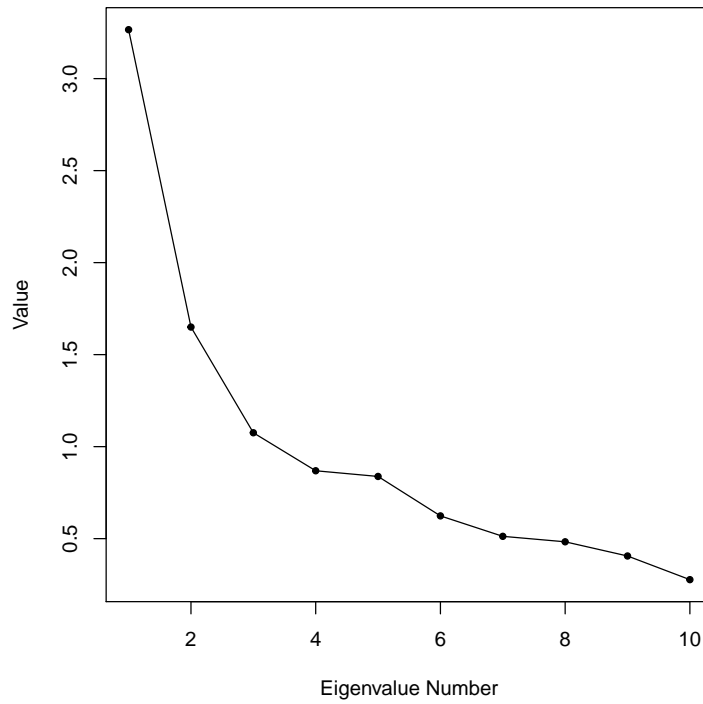
Eigenvalues for the scree plot are

```
> zz <- eigen(cor(diff(MBandCredit, lag=1)), symmetric=TRUE)[["values"]]
> print(zz)
```

```
[1] 3.2659246 1.6498244 1.0754761 0.8689273 0.8380089 0.6236201 0.5125156
[8] 0.4828785 0.4056941 0.2771303
```

The scree plot (Figure 1) is generated by

```
> par(omi=c(0.1,0.1,0.1,0.1),mar=c(4.1,4.1,0.6,0.1))
> plot(zz, ylab="Value", xlab="Eigenvalue Number", pch=20:20,cex=1,type="o")
```



*FAfitStats* is used to calculate the fit statistics for different numbers of factors (Table 1).

```
> z <- FAfitStats(MBandCredit)
> print(z, digits=3)
```

```
$fitStats
```

	0	1	2	3	4	5	6	saturated
chisq	7.22e+02	2.52e+02	1.26e+02	3.79e+01	17.8435	8.5095	1.217	0.000
df	4.50e+01	3.50e+01	2.60e+01	1.80e+01	11.0000	5.0000	0.000	0.000
pval	8.88e-123	1.38e-34	3.93e-15	4.01e-03	0.0853	0.1303	0.000	NA
delta	6.77e+02	2.17e+02	1.00e+02	1.99e+01	6.8435	3.5095	1.217	0.000
RMSEA	2.33e-01	1.50e-01	1.18e-01	6.33e-02	0.0475	0.0504	Inf	NA
RNI	0.00e+00	6.80e-01	8.52e-01	9.71e-01	0.9899	0.9948	0.998	1.000
CFI	0.00e+00	6.80e-01	8.52e-01	9.71e-01	0.9899	0.9948	0.998	1.000
MCI	2.94e-01	6.75e-01	8.34e-01	9.65e-01	0.9877	0.9937	0.998	1.000
GFI	5.84e-01	8.33e-01	9.13e-01	9.75e-01	0.9881	0.9941	0.999	1.000
AGFI	4.91e-01	7.37e-01	8.16e-01	9.22e-01	0.9406	0.9347	-Inf	1.000
AIC	6.32e+02	1.82e+02	7.43e+01	1.89e+00	-4.1565	-1.4905	1.217	0.000
CAIC	7.88e+02	3.84e+02	3.18e+02	2.83e+02	309.1411	339.5296	365.339	364.122
SIC	7.78e+02	3.64e+02	2.89e+02	2.46e+02	265.1411	289.5296	310.339	309.122
CAK	2.69e+00	1.06e+00	6.68e-01	4.05e-01	0.3835	0.3932	0.403	0.399
CK	2.69e+00	1.06e+00	6.76e-01	4.17e-01	0.3967	0.4082	0.420	0.415

```
$seqfitStats
      0 vs 1    1 vs 2    2 vs 3    3 vs 4 4 vs 5 5 vs 6 6 vs saturated
chisq 4.70e+02 1.25e+02 8.84e+01 20.04248 9.334 7.29      1.22
df    1.00e+01 9.00e+00 8.00e+00 7.00000 6.000 5.00      0.00
pval  1.15e-94 1.08e-22 9.91e-16 0.00548 0.156 0.20      0.00
```

```
> c2withML <- estTSF.ML(MBandCredit, 2)
```

The sign and order of factors is arbitrary. For simulation and estimation comparisons it is useful to put them in the same order when different results are compared. This is done by specifying the *BpermuteTarget* argument in *estTSF.ML*. Other than the sign, this does not affect the estimated values, it only rearranges their order. Here they are arranged so 1=transactions, 2=long term, 3=potential spending, 4=consumer credit , 5=N-P term (which is the order they appear as factors are added). The *BpermuteTarget*, *z* below was determined by an initial run.

```
> z <- matrix(0,10,3)
> z[matrix(c( 1,6,2,1:3),3,2)] <- c(10, 56, 41)
> c3withML <- estTSF.ML(MBandCredit, 3, BpermuteTarget=z)
> z <- matrix(0,10,4)
> z[matrix(c( 1,6,2,7,1:4),4,2)] <- c(13, 54, 37, 24)
> c4withML <- estTSF.ML(MBandCredit, 4, BpermuteTarget=z)
> z <- matrix(0,10,5)
> z[matrix(c( 1,6,2,7,5,1:5),5,2)] <- c(13, 67, 34, 30, 72)
> c5withML <- estTSF.ML(MBandCredit, 5, BpermuteTarget=z)
```

The standardized loadings for the four factor model (Table 2) are

```
> print(DstandardizedLoadings(c4withML) )
```

	Factor 1	Factor 2	Factor 3	Factor 4
currency	1.01983211	0.09441862	-0.09386876	-0.08001695
personal cheq.	-0.02898840	0.20319639	0.87632955	-0.07894600
NonbankCheq	0.20214387	-0.13884613	0.36792951	0.11966332
N-P demand & notice	0.55851463	-0.00326414	0.20169780	0.24090075
N-P term	0.02120284	0.10107575	-0.23481508	0.33307726
Investment	0.15013102	0.45888993	-0.23944032	0.06862474
Consumer Credit	0.01841197	0.10508972	0.10142580	0.79551987
Residential Mortgage	0.21597515	0.49350633	0.22622020	0.14854196
Short Term Business Credit	-0.09041731	0.23970572	-0.01376311	0.09703765
Other Business Credit	0.09632589	0.66331904	0.23994908	0.04224045

The estimated  $\Phi$  matrix is

```
> print(c4withML$Phi, digits=3)
```

	Factor1	Factor2	Factor3	Factor4
Factor1	1.000	0.1783	0.3003	0.3400
Factor2	0.178	1.0000	-0.0245	0.5685
Factor3	0.300	-0.0245	1.0000	0.0124
Factor4	0.340	0.5685	0.0124	1.0000

Communalities for the 4 factor model are

```
> print(1 - c4withML$stats$uniquenesses)
```

currency	0.97758543	personal cheq.	0.77183960
NonbankCheq	0.24563691	N-P demand & notice	0.56948987
N-P term	0.21682570	Investment	0.34592814
Consumer Credit	0.76283210	Residential Mortgage	0.53132528
Short Term Business Credit	0.08886927	Other Business Credit	0.57238659

Communalities for other models are given by the following, but output is omitted here.

```
> print(1 - c2withML$stats$uniquenesses)
> print(1 - c3withML$stats$uniquenesses)
> print(1 - c5withML$stats$uniquenesses)
```

Loadings for 4 factor model (Table 3) are

```
> print(loadings(c4withML) )
```

	Factor 1	Factor 2	Factor 3	Factor 4
currency	13.7349771	1.2716188	-1.2642133	-1.077659
personal cheq.	-1.2450233	8.7270851	37.6374926	-3.390653
NonbankCheq	2.1727138	-1.4923673	3.9546366	1.286184
N-P demand & notice	25.7154238	-0.1502893	9.2866761	11.091679
N-P term	1.5251290	7.2704190	-16.8903434	23.958381
Investment	17.5631039	53.6833203	-28.0109680	8.028077
Consumer Credit	0.5523918	3.1528783	3.0429543	23.867010
Residential Mortgage	10.8700198	24.8381514	11.3856526	7.476110
Short Term Business Credit	-6.4954824	17.2202022	-0.9887271	6.971081
Other Business Credit	5.4240616	37.3511568	13.5114103	2.378538

Figure 2 is generated by

```
> tfplot(ytoypc(factors(c4withML)),
         Title= "Factors from 4 factor model (year-to-year growth rate)",
```

```

lty=c("solid"),
col=c("black"),
xlab=c(""),ylab=c("factor 1","factor 2","factor 3","factor 4"),
par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
reset.screen=TRUE)

```

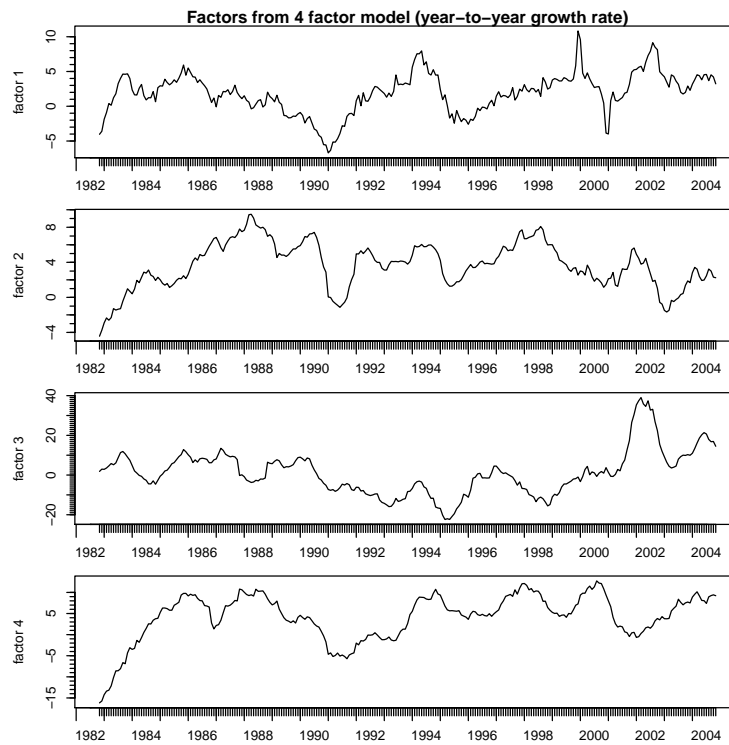


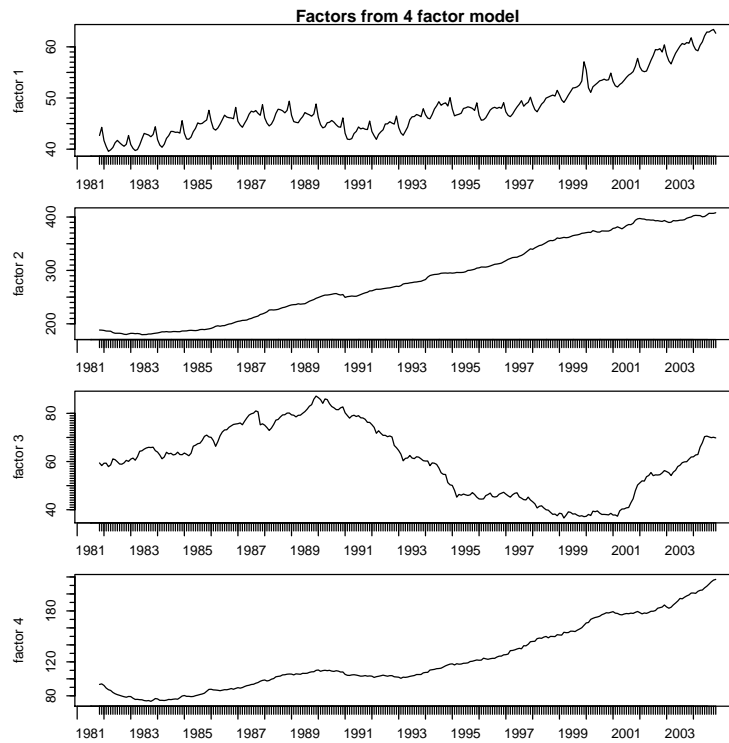
Figure 3 is generated by

```

> tfplot(factors(c4withML),
        Title="Factors from 4 factor model",
        lty=c("solid"),
        col=c("black"),
        xlab=c(""),ylab=c("factor 1","factor 2","factor 3","factor 4"),
        par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
        reset.screen=TRUE)

```





Some points are close to zero and cause problems in growth rate graphics. One solution is to set them to NA, but truncating the graphic works better. Figure 4 is generated by

```
> z <- explained(c4withML)
> tfplot(ytoypc(MBandCredit), ytoypc(z), series=1:5, graphs.per.page=5,
  lty=c("solid", "dashed"),
  col=c("black", "red"),
  ylab=c("currency", "personal cheq.", "NonbankCheq",
    "N-P demand & notice", "N-P term"),
  ylim=list(NULL,NULL,c(-100,100),NULL,NULL),
  Title=
    "Explained money indicator 1-5 (year-to-year growth rate) using 4 factors",
  par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
  reset.screen=TRUE)
```

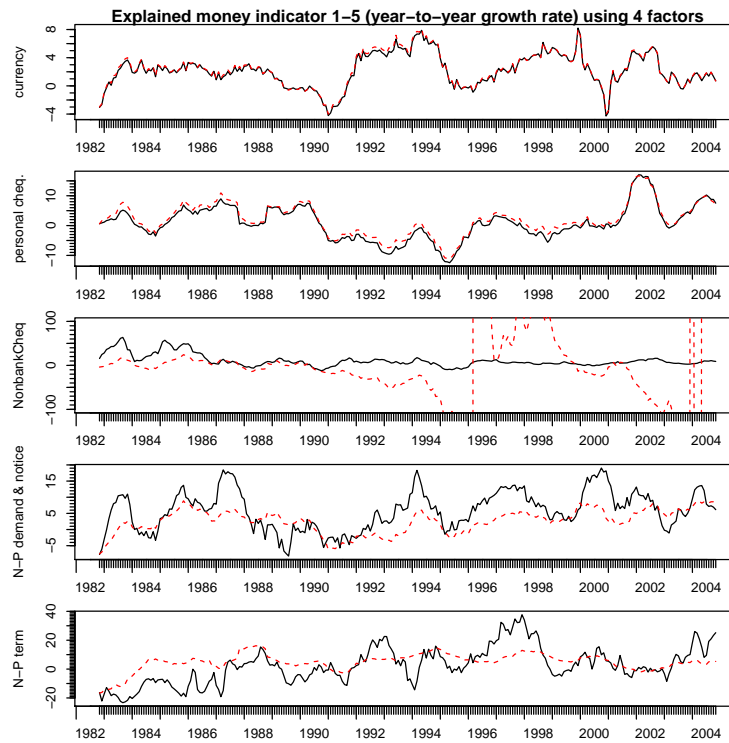


Figure 5 is generated by

```
> tfplot(ytoypc(MBandCredit), ytoypc(explained(c4withML)), series=6:10,
graphs.per.page=5,
lty=c("solid", "dashed"),
col=c("black", "red"),
ylab=c("", "", "", "", "",
"Investment", "Consumer Credit", "Residential Mortgage",
"Short Term Business Credit", "Other Business Credit"),
Title=
"Explained money indicator 6-10 (year-to-year growth rate)using 4 factors",
par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
reset.screen=TRUE)
```

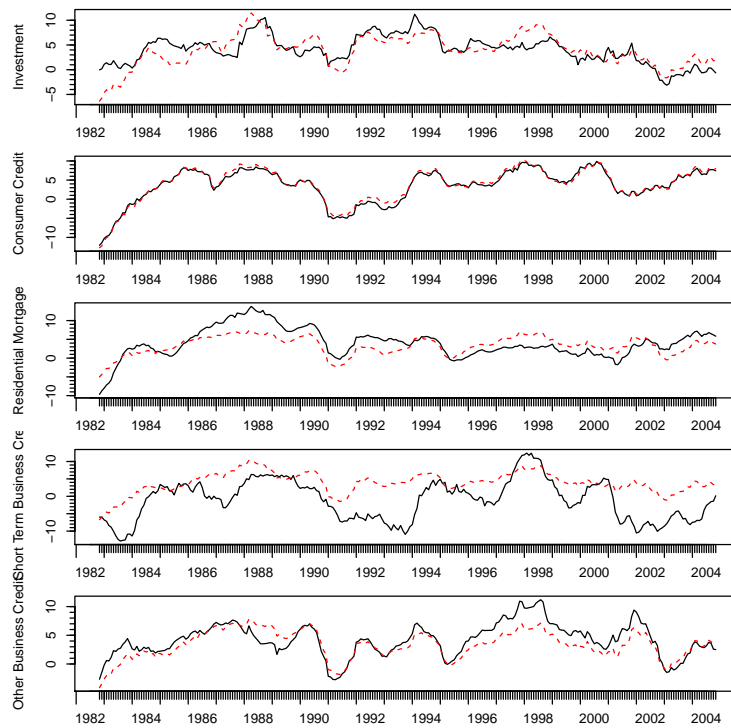


Figure 6 is generated by

```
> tfplot( MBandCredit, explained(c4withML), series=1:5, graphs.per.page=5,
  lty=c("solid", "dashed"),
  col=c("black", "red"),
  Title= "Explained money indicators 1-5 using 4 factors",
  par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
  reset.screen=TRUE)
```

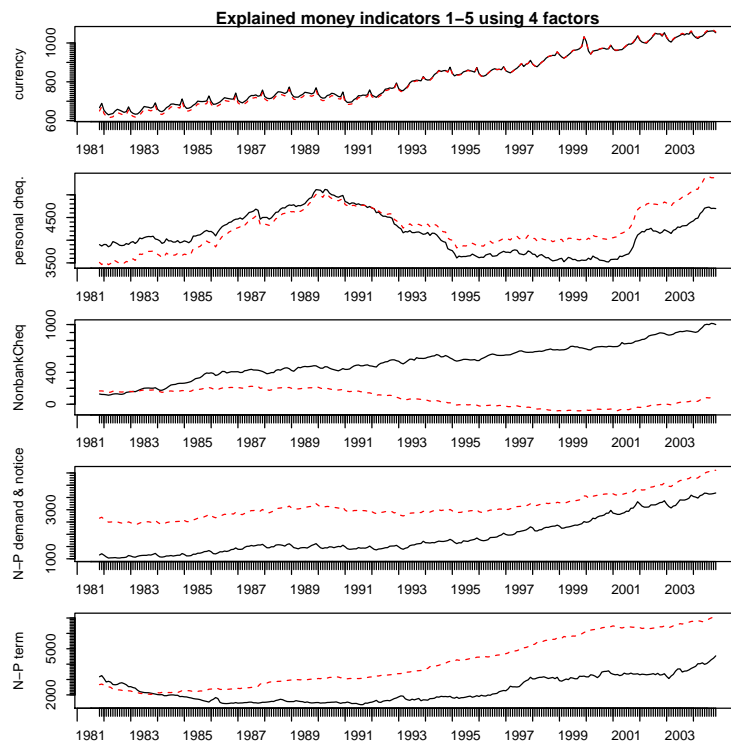
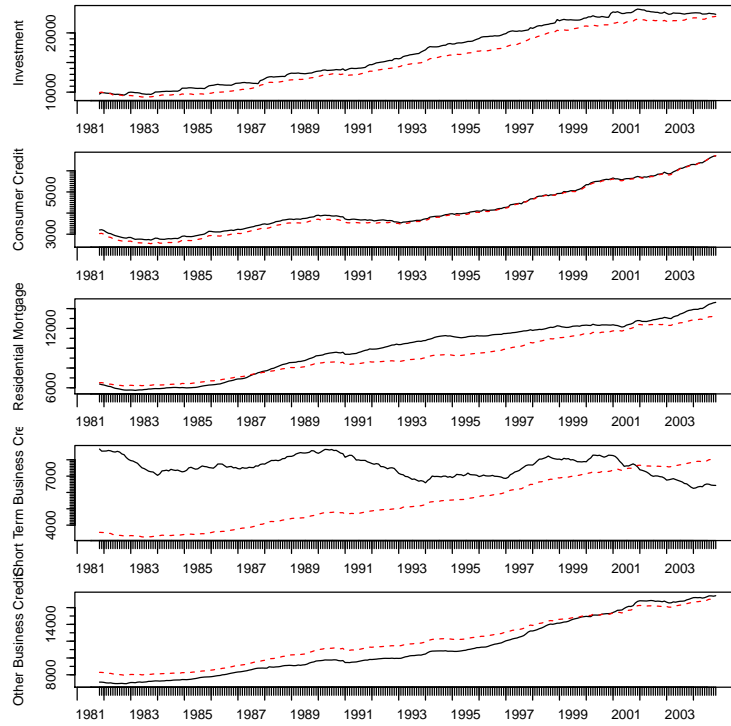


Figure 7 is generated by

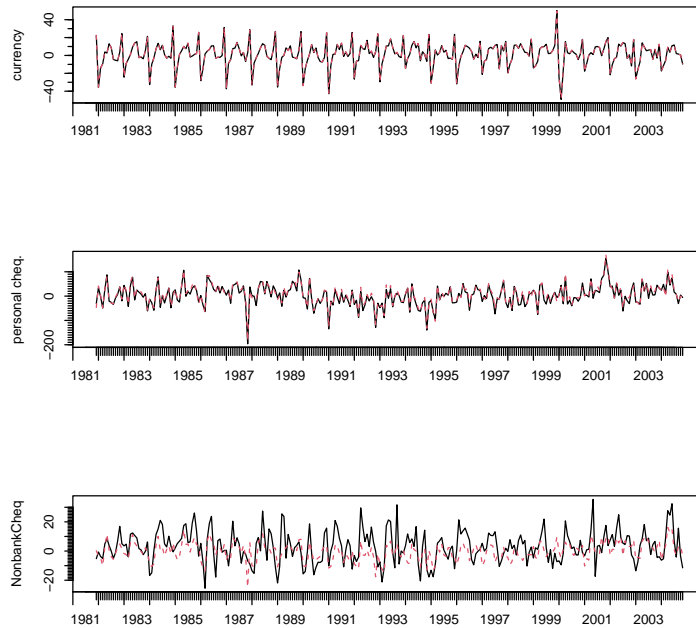
```
> tfplot( MBandCredit, explained(c4withML), series=6:10, graphs.per.page=5,
          lty=c("solid", "dashed"),
          col=c("black", "red"),
          Title= "Explained money indicator 6-10 using 4 factors",
          par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
          reset.screen=TRUE)
```



The differenced

version of the above is not in the paper, but is given by

```
> tfplot( diff(MBandCredit), diff(explained(c4withML)), graphs.per.page=3)
```



Summary infor-

mation about the model is calculated with

```
> summary(MBandCredit)
```

### 3 Two, three and five factors models

Tables of standardized loadings for differenced models, and the estimated  $\Phi$  are given by the following (output is omitted).

```
> DstandardizedLoadings(c2withML)
> print(c2withML$Phi, digits=3)
> DstandardizedLoadings(c3withML)
> print(c3withML$Phi, digits=3)
> print(DstandardizedLoadings(c5withML), digits=3)
> print(c5withML$Phi, digits=3)
> DstandardizedLoadings(c4withML)
> print(c2withML$Phi, digits=3)
```

Figure 8 part 1 is generated by

```
> tfplot(ytoypc(factors(c4withML)), ytoypc(factors(c2withML)),
        ytoypc(factors(c3withML)),
        ytoypc(factors(c5withML)), series=1:2,
```

```

xlab=c(""),ylab=c("factor 1","factor 2"),
lty=c("solid", "dotdash", "dashed", "dotted"),
col=c("black","green","red","blue"),
Title= paste("Factors transaction and long ",
"(year-to-year growth rate) using 2, 3, 4 and 5 factor models", sep=""),
par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
reset.screen=TRUE)

```

Figure 8 part 2 is generated by

```

> tfplot(ytoypc(factors(c4withML)),
ytoypc(factors(c3withML)),
ytoypc(factors(c5withML)), series=3,
lty=c("solid", "dashed", "dotted"),
xlab=c(""),ylab=c("", "", "factor 3"),
col=c("black", "red", "blue"),
Title= paste("Factor near ",
"(year-to-year growth rate) using 3, 4 and 5 factor models", sep=""),
par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
reset.screen=TRUE)

```

## 4 Rotation method sensitivity

*BpermuteTarget* just helps put the factors in the same order, and with the same signs as *c4withML*. It does not otherwise affect the estimation or rotation. The oblimin rotation with  $\gamma = 0.5$  is given by

```

> c4withMLg0.5 <- estTSF.ML(MBandCredit, 4, BpermuteTarget=loadings(c4withML),
rotation="oblimin", rotationArgs=list(gam=0.5))
> loadings(c4withMLg0.5)

```

	Factor 1	Factor 2	Factor 3	Factor 4
currency	17.204376	-0.2993325	-2.9221503	-4.161355
personal cheq.	-5.888054	11.8489892	40.8218129	-9.350992
NonbankCheq	2.306075	-3.1917702	3.5808888	2.094587
N-P demand & notice	28.858827	-9.6930351	4.4360441	12.301306
N-P term	-2.519688	1.6858628	-21.7335744	32.446348
Investment	14.121979	70.4973501	-30.5725683	-12.535677
Consumer Credit	-4.945051	-4.9008176	-0.7956332	33.697057
Residential Mortgage	6.854980	29.4305518	10.4874854	-1.255816
Short Term Business Credit	-12.201107	22.7939544	-0.5132386	4.095379
Other Business Credit	-1.130633	49.8454385	14.9460409	-13.213999

```

> DstandardizedLoadings(c4withMLg0.5)

```

	Factor 1	Factor 2	Factor 3	Factor 4
currency	1.27743748	-0.02222566	-0.216971802	-0.30898366

personal cheq.	-0.13709404	0.27588499	0.950471411	-0.21772307
NonbankCheq	0.21455145	-0.29695432	0.333156946	0.19487511
N-P demand & notice	0.62678638	-0.21052354	0.096346674	0.26717270
N-P term	-0.03502953	0.02343742	-0.302147267	0.45107975
Investment	0.12071597	0.60261780	-0.261337111	-0.10715611
Consumer Credit	-0.16482527	-0.16335091	-0.026519536	1.12316872
Residential Mortgage	0.13620079	0.58475219	0.208374622	-0.02495165
Short Term Business Credit	-0.16983977	0.31729252	-0.007144296	0.05700780
Other Business Credit	-0.02007891	0.88520494	0.265426680	-0.23466735

```
> DstandardizedLoadings(c4withMLg0.5) - DstandardizedLoadings(c4withML)
```

	Factor 1	Factor 2	Factor 3	Factor 4
currency	0.25760538	-0.11664428	-0.123103042	-0.22896671
personal cheq.	-0.10810564	0.07268860	0.074141866	-0.13877707
NonbankCheq	0.01240759	-0.15810820	-0.034772560	0.07521179
N-P demand & notice	0.06827175	-0.20725940	-0.105351130	0.02627195
N-P term	-0.05623237	-0.07763833	-0.067332182	0.11800249
Investment	-0.02941505	0.14372787	-0.021896794	-0.17578085
Consumer Credit	-0.18323724	-0.26844063	-0.127945338	0.32764886
Residential Mortgage	-0.07977436	0.09124587	-0.017845579	-0.17349361
Short Term Business Credit	-0.07942246	0.07758680	0.006618814	-0.04002985
Other Business Credit	-0.11640480	0.22188590	0.025477600	-0.27690780

Summary information is produced with (output omitted)

```
> summary(c4withMLg0.5)
```

Other rotation results are produced by the following, but outputs are omitted.

```
> c4withMLgneg0.5 <- estTSF.ML(MBandCredit, 4, BpermuteTarget=loadings(c4withML),
  rotation="oblimin", rotationArgs=list(gam=-0.5))
> loadings(c4withMLgneg0.5)
> DstandardizedLoadings(c4withMLgneg0.5)
> DstandardizedLoadings(c4withMLgneg0.5) - DstandardizedLoadings(c4withML)
> summary(c4withMLgneg0.5)
> c4withMLgneg1.0 <- estTSF.ML(MBandCredit, 4, BpermuteTarget=loadings(c4withML),
  rotation="oblimin", rotationArgs=list(gam=-1.0))
> loadings(c4withMLgneg1.0)
> DstandardizedLoadings(c4withMLgneg1.0)
> DstandardizedLoadings(c4withMLgneg1.0) - DstandardizedLoadings(c4withML)
> summary(c4withMLgneg1.0)
> c4withMLbQ <- estTSF.ML(MBandCredit, 4, rotation="bentlerQ",
  BpermuteTarget=loadings(c4withML))
> loadings(c4withMLbQ)
> DstandardizedLoadings(c4withMLbQ)
> DstandardizedLoadings(c4withMLbQ) - DstandardizedLoadings(c4withML)
> summary(c4withMLbQ)
```



Figure 9 is generated by

```
> tfplot(ytoypc(factors(c4withML)), ytoypc(factors(c4withMLg0.5)),
        ytoypc(factors(c4withMLgneg0.5)), ytoypc(factors(c4withMLgneg1.0)),
        ytoypc(factors(c4withMLbQ)),
        xlab=c(""), ylab=c("factor 1", "factor 2", "factor 3", "factor 4"),
        lty=c("solid", "dashed", "dotted", "dotdash", "longdash"),
        col=c("black", "red", "blue", "green", "pink"),
        Title= paste(
            "Factors from various 4 factor models (year-to-year growth rate)",
            "\n and oblimin with gam=0 (solid)"),
        par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
        reset.screen=TRUE)
```

Geomin factors 2 and 3 each have one modestly different loading. Factor 2 has personal chequing mixed in with investment and credit. Factor 3 explains on currency, personal chequing, and investment, so the separation is not so interesting. Output is omitted from these.

```
> c4withMLgm <- estTSF.ML(MBandCredit, 4, rotation="geominQ",
    BpermuteTarget=loadings(c4withML))
> loadings(c4withMLgm)
> DstandardizedLoadings(c4withMLgm)
```

The difference between the estimates can be checked with (output omitted)

```
> DstandardizedLoadings(c4withMLgm) - DstandardizedLoadings(c4withML)
```

The summary of the 4 factor geomin estimate is given by

```
> summary(c4withMLgm)
```

Figure 10 is generated by

```
> tfplot(ytoypc(factors(c4withML)), ytoypc(factors(c4withMLgm)),
        xlab=c(""), ylab=c("factor 1", "factor 2", "factor 3", "factor 4"),
        lty=c("solid", "dashed"),
        col=c("black", "red"),
        Title= paste(
            "Factors from geomin (dashed) 4 factor model (year-to-year growth rate)",
            "\n and oblimin with gam=0 (solid)"),
        par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
        reset.screen=TRUE)
> c4withMLnotNorm <- estTSF.ML(MBandCredit, 4, normalize=FALSE,
    BpermuteTarget=loadings(c4withML))
>
```

There is only a qualitative statement about the next in the paper (outputs omitted).

```
> DstandardizedLoadings(c4withML)
> DstandardizedLoadings(c4withMLnotNorm)
> DstandardizedLoadings(c4withML) - DstandardizedLoadings(c4withMLnotNorm)
```

## 5 Sensitivity to sample period

*BpermuteTarget=loadings(c4withML)* is not good enough in some cases. There are difficulties interpreting factors 2 and 3 in here.

```
> z <- matrix(0,10,4)
> z[matrix(c( 1,6,2,7,1:4),4,2)] <- c(11, 104, 20, 13)
> c4withMLbefore90 <- estTSF.ML(tfwindow(MBandCredit, end=c(1989,12)), 4,
  BpermuteTarget=z)
> c4withMLafter95 <- estTSF.ML(tfwindow(MBandCredit, start=c(1995,1)), 4,
  BpermuteTarget=loadings(c4withML))
> z <- matrix(0,10,4)
> z[matrix(c( 1,6,2,7,1:4),4,2)] <- c(11, 104, 20, 13)
> c4withMLbefore95 <- estTSF.ML(tfwindow(MBandCredit, end=c(1994,12)), 4,
  BpermuteTarget=z)
> c4withMLafter00 <- estTSF.ML(tfwindow(MBandCredit, start=c(2000,1)), 4,
  BpermuteTarget=loadings(c4withML))
> c4withML90to00 <- estTSF.ML(tfwindow(MBandCredit, start=c(1990,1), end=c(2000,1)), 4,
  BpermuteTarget=loadings(c4withML))
```

Figure 11 is generated by

```
> tfplot(ytoypc(factors(c4withML)), ytoypc(factors(c4withMLbefore90)),
  ytoypc(factors(c4withMLbefore95)), ytoypc(factors(c4withMLafter95)),
  ytoypc(factors(c4withMLafter00)), ytoypc(factors(c4withML90to00)),
  xlab=c(""), ylab=c("factor 1", "factor 2", "factor 3", "factor 4"),
  ylim=list(NULL, c(-20,20), c(-25,40), NULL),
  graphs.per.page=4,
  lty=c("dashed", "dotted", "dotdash", "longdash", "dotted",
    "twodash"),
  col=c("red", "blue", "green", "pink", "violet", "brown"),
  Title= paste(
    "Factors (year to year growth) using full sample and sub-samples\n",
    "ML estimation with quartimin rotation objective", sep=""),
  par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
  reset.screen=TRUE)
```

Figure 12 is generated by

```
> tfplot(ytoypc(MBandCredit), ytoypc(explained(c4withML)),
  ytoypc(explained(c4withMLbefore90)), ytoypc(explained(c4withMLbefore95)),
  ytoypc(explained(c4withMLafter95)), ytoypc(explained(c4withMLafter00)),
  ytoypc(explained(c4withML90to00)), series=1:5, graphs.per.page=5,
  ylab=c("currency", "personal cheq.", "NonbankCheq",
    "N-P demand & notice", "N-P term"),
  ylim=list(NULL, NULL, c(-70,70), NULL, c(-70,70)),
  lty=c("solid", "dashed", "dotted", "dotdash", "longdash", "dotted",
```

```

        "twodash"),
col=c("black", "red", "blue", "green", "pink", "violet", "brown"),
Title= paste("Explained money indicators 1-5 (year to year growth)\n",
        "using 4 factors, full sample and sub-samples", sep=""),
par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
reset.screen=TRUE)

```

Figure 13 is generated by

```

> tfplot(ytoypc(MBandCredit), ytoypc(explained(c4withML)),
        ytoypc(explained(c4withMLbefore90)), ytoypc(explained(c4withMLbefore95)),
        ytoypc(explained(c4withMLafter95)), ytoypc(explained(c4withMLafter00)),
        ytoypc(explained(c4withML90to00)), series=6:10, graphs.per.page=5,
        ylab=c("", "", "", "", "",
        "Investment", "Consumer Credit", "Residential Mortgage",
        "Short Term Business Credit", "Other Business Credit"),
        lty=c("solid", "dashed", "dotted", "dotdash", "longdash", "dotted",
        "twodash"),
        col=c("black", "red", "blue", "green", "pink", "violet", "brown"),
        Title= paste("Explained money indicators 6-10 (year to year growth)\n",
        "using 4 factors, full sample and sub-samples", sep=""),
        par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
        reset.screen=TRUE)

```

## 6 Comparison with Aggregates

Compute aggregates M1+ and M2++

```

> M1 <- tfwindow(M1total, start=c(1981,11), end=c(2004,11)) * scale
> seriesNames(M1) <- "Real per Capita M1"
> z <- tframed(MB2001 + MB486 + MB487p + MB452 + MB452adj + MB472 + NonbankCheq)
> M1p <- tfwindow(z, start=c(1981,11), end=c(2004,11)) * scale
> seriesNames(M1p) <- "Real per Capita M1+"
> M2pp <- tfwindow(M1total
        + MB472 + MB473 + MB452 + MB453 + MB454
        + NonbankCheq + NonbankNonCheq + NonbankTerm +
        + MB2046 + MB2047 + MB2048
        + MB2057 + MB2058, start=c(1981,11), end=c(2004,11))* scale
> seriesNames(M2pp) <- "Real per Capita M2++"

```

and put factors on the same scale.

```

> f <- tframed(factors(c4withML)[,1:2], tf=tframe(factors(c4withML)))
> mnF <- colMeans(f)
> mnM <- colMeans(cbind(M1p, M2pp))
> f <- sweep(f, 2, mnM/mnF, "*")

```

Now compare the transaction factor with M1+ and the long term factor with M2++. Figure 14 is generated by

```
> tfplot(ytoypc(f), ytoypc(cbind(M1, M2pp)), graphs.per.page=2,
        lty=c("dashed", "solid"),
        col=c("red", "black"),
        Title=
        paste("year to year growth) M1+ and M2++ (solid) and scaled Bartlett Predictors\n",
              "computed using ordinary ML parameters (dashed)", sep=""),
        ylab=c("M1+ vs. factor 1", "M2++ vs. factor 2" ),
        par=list(mar=c(2.1, 4.1, 1.1, 0.1)),
        reset.screen=TRUE)
```

## 7 References

Paul D. Gilbert and Erik Meijer, (2006) "Money and Credit Factors", Bank of Canada Working Paper 2006-3, available at <<https://www.bankofcanada.ca/2006/03/working-paper-2006-3>>.

Paul D. Gilbert and Erik Meijer, (2005) "Time Series Factor Analysis with an Application to Measuring Money", Research Report 05F10, University of Groningen, SOM Research School, available at  
<<https://hdl.handle.net/11370/d7d4ea3d-af1d-487a-b9b6-c0816994ef5a>>.