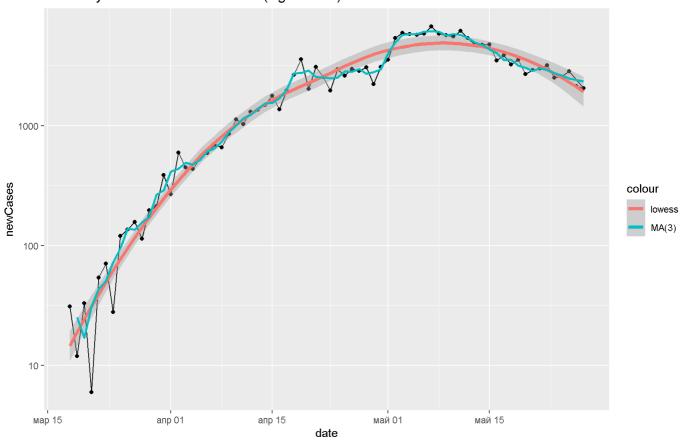
COVID in Moscow

knitr::opts_chunk\$set(echo = FALSE, message=FALSE, warning=FALSE, fig.width=9, fig.height=6, fig.path = "
figures/")

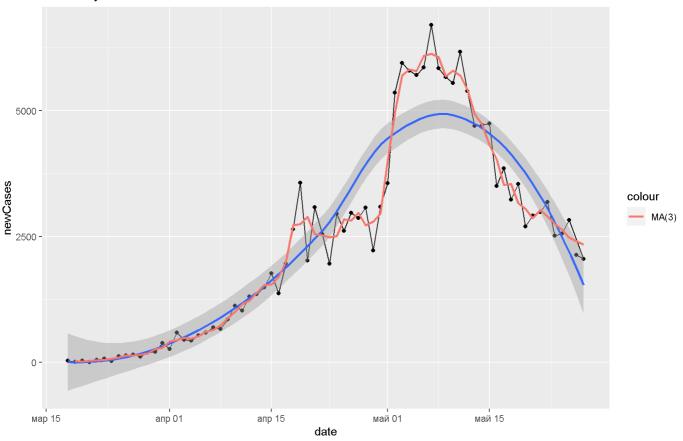
[1] "CovidMoscowDB was NOT updated"

Общий тренд

New daily COVID cases in Moscow (log10 scale)



New daily COVID cases in Moscow



One line

```
## Call:
## lm(formula = log(newCases) ~ days from beg, data = CovidMoscow)
##
## Residuals:
## Min 1Q Median
                         3Q
## -3.1363 -0.6370 0.3773 0.7057 1.3465
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.73087 0.21762 21.74 <2e-16 ***
## days_from_beg 0.06572 0.00529 12.42 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.9329 on 70 degrees of freedom
## Multiple R-squared: 0.6879, Adjusted R-squared: 0.6835
## F-statistic: 154.3 on 1 and 70 DF, p-value: < 2.2e-16
```

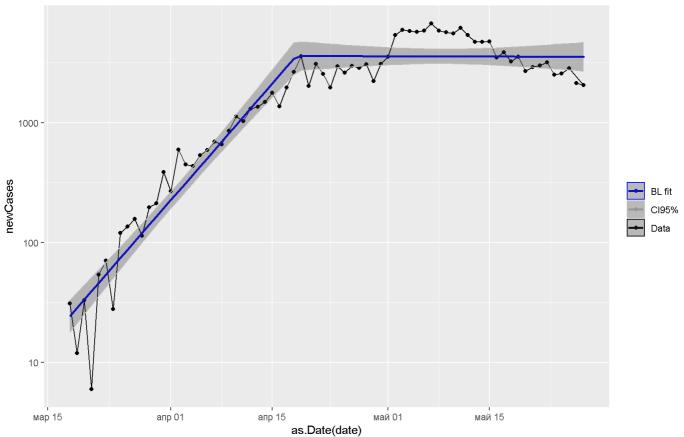
Segmented

```
0.152308 20.96
## (Intercept)
                    3.192574
                                                  <2e-16 ***
## days_from_beg
                    0.159579
                               0.008442
                                          18.90
                                                  <2e-16 ***
## U1.days from beg -0.159935 0.010380 -15.41
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4409 on 68 degrees of freedom
## Multiple R-Squared: 0.9323, Adjusted R-squared: 0.9293
##
## Convergence attained in 2 iter. (rel. change 2.6671e-16)
##
## ----- N. Breaking points = 2 =-----
##
##
   ***Regression Model with Segmented Relationship(s)***
##
## Call:
## segmented.lm(obj = lm msk0, psi = start psi[1:npt])
##
## Estimated Break-Point(s):
##
                        Est. St.Err
## psil.days from beg 15.408 1.219
## psi2.days from beg 49.640 1.407
##
## Meaningful coefficients of the linear terms:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.65796 0.15661 16.97 <2e-16 ***
## days_from_beg 0.23609 0.01779 13.27 <2e-16 ***
## U1.days from beg -0.16261
                               0.01869
                                         -8.70
## U2.days from beg -0.12722
                               0.01243 -10.24
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.328 on 66 degrees of freedom
## Multiple R-Squared: 0.9636, Adjusted R-squared: 0.9609
## Convergence attained in 2 iter. (rel. change 1.7055e-06)
##
## ----- N. Breaking points = 3 =----
##
##
   ***Regression Model with Segmented Relationship(s)***
##
## Call:
## segmented.lm(obj = lm msk0, psi = start psi[1:npt])
##
## Estimated Break-Point(s):
                        Est. St.Err
##
## psil.days from beg 13.000 1.904
## psi2.days from beg 28.003 3.616
## psi3.days_from_beg 51.253 1.778
##
## Meaningful coefficients of the linear terms:
             Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 2.64540 0.16713 15.829 < 2e-16 ***
## days_from_beg 0.23923 0.02364 10.122 6.53e-15 ***
## U1.days from beg -0.12692
                               0.02929 -4.334
## U2.days from beg -0.05672
                               0.01999 -2.838
                                                      NA
                               0.01592 -6.912
## U3.days from beg -0.11002
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3189 on 64 degrees of freedom
## Multiple R-Squared: 0.9667, Adjusted R-squared: 0.963
##
\#\# Convergence attained in 10 iter. (rel. change 7.7821e-08)
##
## ----- N. Breaking points = 4 =-----
##
##
   ***Regression Model with Segmented Relationship(s)***
##
## Call:
## segmented.lm(obj = lm_msk0, psi = start_psi[1:npt])
##
## Estimated Break-Point(s):
```

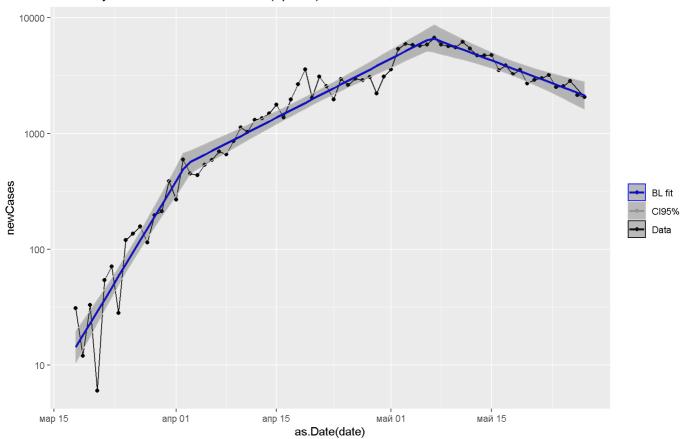
```
##
## psil.days from beg 13.000 1.788
## psi2.days from beg 30.000 3.126
## psi3.days from beg 45.001 3.014
## psi4.days_from_beg 48.000 1.245
##
## Meaningful coefficients of the linear terms:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 2.64435 0.16194 16.329 <2e-16 ***
## days_from_beg 0.23950 0.02117 11.311 <2e-16 ***
                              0.02117 11.311
## U1.days from beg -0.12813
                              0.02642
                                       -4.849
                             0.02478 -2.947
## U2.days_from_beg -0.07303
## U3.days_from_beg 0.16998 0.22662 0.750
## U4.days_from_beg -0.25904
                            0.22604 -1.146
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3194 on 62 degrees of freedom
## Multiple R-Squared: 0.9676, Adjusted R-squared: 0.9629
##
## Convergence attained in 1 iter. (rel. change -2.94e-06)
##
## ----- N. Breaking points = 5 =----
##
## ***Regression Model with Segmented Relationship(s)***
##
## Call:
## segmented.lm(obj = lm msk0, psi = start psi[1:npt])
##
## Estimated Break-Point(s):
##
                      Est. St.Err
## psil.days from beg 13.000 1.720
## psi2.days from beg 31.998 2.152
## psi3.days_from_beg 42.622 1.502
## psi4.days_from_beg 45.464 1.742
## psi5.days from beg 51.370 4.402
## Meaningful coefficients of the linear terms:
##
              Estimate Std. Error t value Pr(>|t|)
                  ## (Intercept)
## days_from_beg 0.23941 0.02071 11.560 <2e-16 ***
## U1.days from beg -0.12831 0.02511 -5.111
## U2.days from beg -0.10861 0.03299 -3.292
## U3.days from beg 0.26804 0.22288 1.203
## U4.days_from_beg -0.25980 0.23316 -1.114
## U5.days from beg -0.06510 0.07565 -0.861
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3124 on 60 degrees of freedom
## Multiple R-Squared: 0.97, Adjusted R-squared: 0.9645
##
## Convergence attained in 13 iter. (rel. change 5.0844e-07)
## breakpoint estimate(s): 0.4723582 11.56513 17.68183 33.42125 40.15426 49.20468
##
## ----- N. Breaking points = 6 =----
##
## ***Regression Model with Segmented Relationship(s)***
##
## Call:
## segmented.lm(obj = lm msk0, psi = start psi[1:npt])
## Estimated Break-Point(s):
##
                       Est. St.Err
## psil.days from beg 13.004 1.719
## psi2.days_from_beg 31.997 2.382
## psi3.days from beg 42.678 1.481
## psi4.days from beg 45.543 1.623
## psi5.days from beg 53.945 4.279
## psi6.days from beg 60.894 8.857
##
## Meaningful coefficients of the linear terms:
```

```
Estimate Std. Error t value Pr(>|t|)
                  2.64474 0.16059 16.468 <2e-16 ***
## (Intercept)
                  0.23940
## days from beg
                              0.02100 11.402
                                                <2e-16 ***
                              0.02484 -5.168
## U1.days_from_beg -0.12836
                                                    NA
## U2.days_from_beg -0.10815
                              0.03731 -2.899
                                                    NA
                                       1.206
## U3.days_from_beg 0.27330
                               0.22664
                                                    NA
                               0.22921 -1.211
## U4.days from beg -0.27755
                                                    NA
## U5.days from beg -0.07661
                               0.07727 -0.992
                                                    NA
## U6.days_from_beg 0.03598
                                       0.537
                               0.06704
                                                    NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3167 on 58 degrees of freedom
## Multiple R-Squared: 0.9702, Adjusted R-squared: 0.9635
##
## Convergence attained in 5 iter. (rel. change 4.441e-06)
```

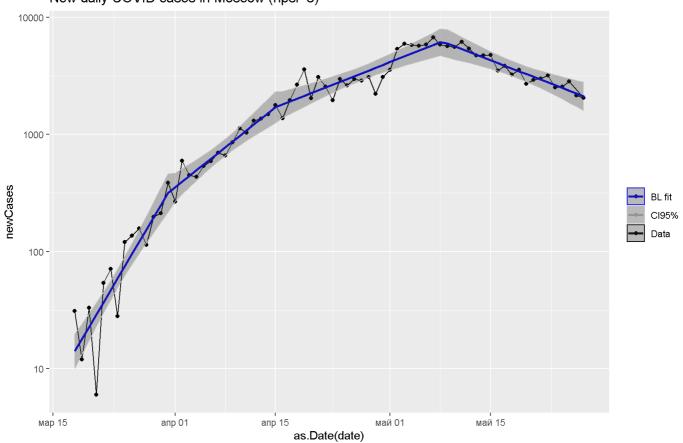
New daily COVID cases in Moscow (npsi=1)



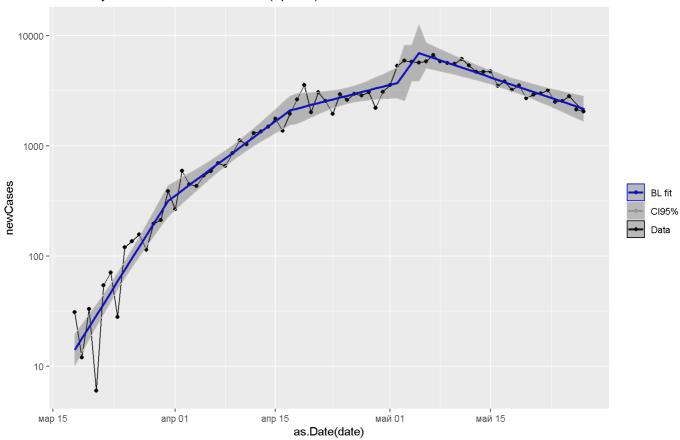
New daily COVID cases in Moscow (npsi=2)



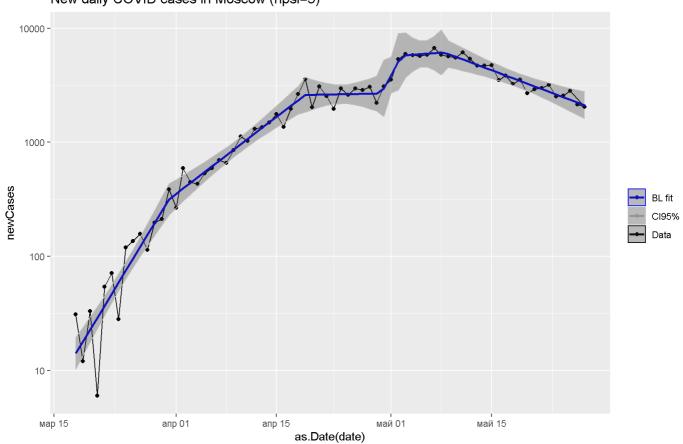
New daily COVID cases in Moscow (npsi=3)



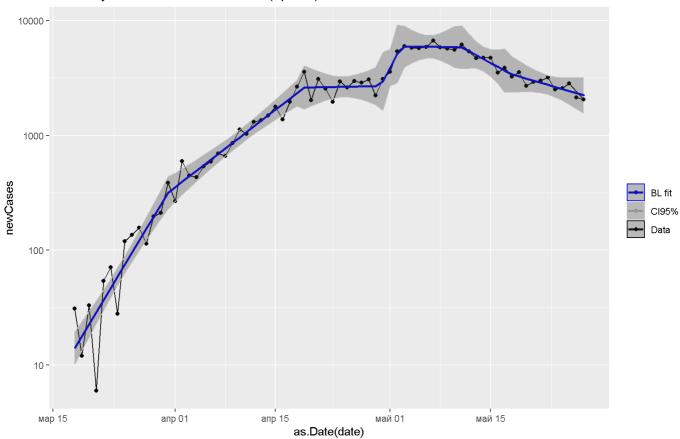
New daily COVID cases in Moscow (npsi=4)



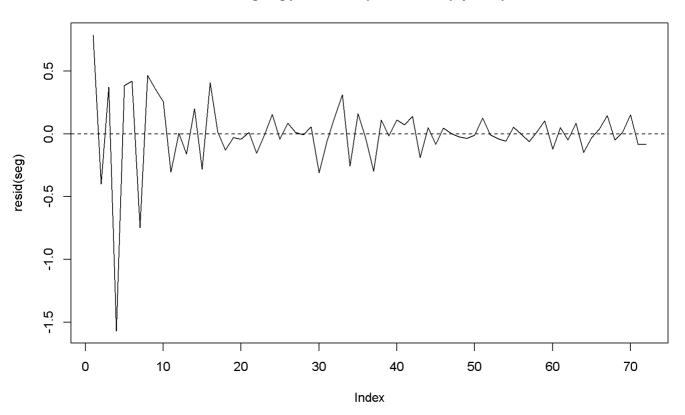
New daily COVID cases in Moscow (npsi=5)



New daily COVID cases in Moscow (npsi=6)



Seg log(newCases) residuals (npsi=6)



```
##
## -----
N. Breaking points = 1 =-----

## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u

## slope1 0.15958000 0.0084420 18.903000 0.142730 0.176430
## slope2 -0.00035581 0.0060396 -0.058913 -0.012408 0.011696
```

```
## Doubling of cases:
## slope1 slope2
     4.343572 -1948.082349
##
##
## ----- N. Breaking points = 2 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.236090 0.0177900 13.2710 0.200570 0.271600
## slope2 0.073471 0.0057342 12.8130 0.062022 0.084919
## slope3 -0.053751 0.0110230 -4.8761 -0.075760 -0.031742
##
## Doubling of cases:
## slope1 slope2 slope3
   2.935945 9.434296 -12.895522
##
##
## ----- N. Breaking points = 3 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.239230 0.023635 10.1220 0.192020 0.286450
## slope2 0.112320 0.017292 6.4950 0.077770 0.146860
## slope3 0.055593 0.010023 5.5465 0.035570 0.075617
## slope4 -0.054423 0.012365 -4.4015 -0.079124 -0.029721
##
## Doubling of cases:
   slope1 slope2 slope3 slope4
2.897409 6.171182 12.468246 -12.736291
##
##
## ----- N. Breaking points = 4 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.239500 0.021173 11.31100 0.19717000 0.281820
## slope2 0.111360 0.015810 7.04360 0.07975600 0.142970
## slope3 0.038332 0.019085 2.00850 0.00018156 0.076482
## slope4 0.208310 0.225820 0.92248 -0.24309000 0.659710
## slope5 -0.050727 0.010039 -5.05320 -0.07079500 -0.030660
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.894143 6.224382 18.082729 3.327479 -13.664265
##
##
## ----- N. Breaking points = 5 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.2394100 0.020710 11.560000 0.197980 0.280830
## slope2 0.1111000 0.014191 7.828500 0.082710 0.139480
## slope3 0.0024823 0.029783 0.083347 -0.057093 0.062058
## slope4 0.2705200 0.220880 1.224700 -0.171300 0.712350
## slope5 0.0107170 0.074671 0.143530 -0.138650 0.160080
## slope6 -0.0543810 0.012113 -4.489400 -0.078611 -0.030150
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.895231 6.238949 279.235862 2.562277 64.677352 -12.746128
##
##
## ----- N. Breaking points = 6 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.2394000 0.020997 11.402000 0.197370 0.281430
## slope2 0.1110400 0.013266 8.370300 0.084482 0.137590
## slope3 0.0028818 0.034867 0.082652 -0.066913 0.072676
## slope4 0.2761800 0.223940 1.233300 -0.172080 0.724450
## slope5 -0.0013717 0.048867 -0.028069 -0.099191 0.096447
## slope6 -0.0779840 0.059850 -1.303000 -0.197790 0.041820
## slope7 -0.0420070 0.030196 -1.391200 -0.102450 0.018436
##
## Doubling of cases:
   slope1 slope2 slope3 slope4
                                                slope5
```

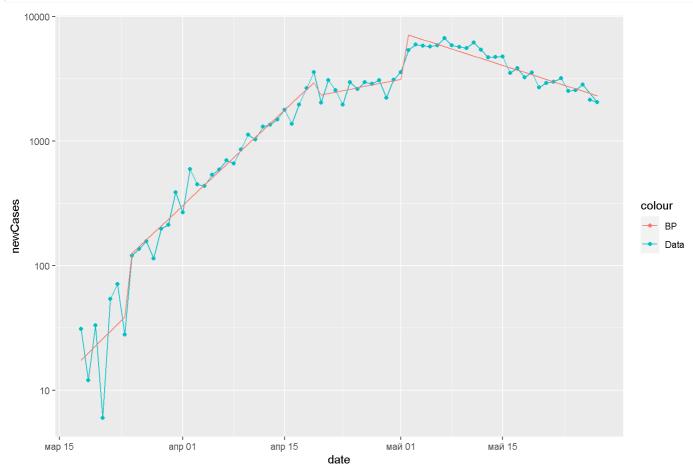
```
## 2.895352 6.242320 240.525776 2.509766 -505.319808 -8.888326

## slope7

## -16.500754
```

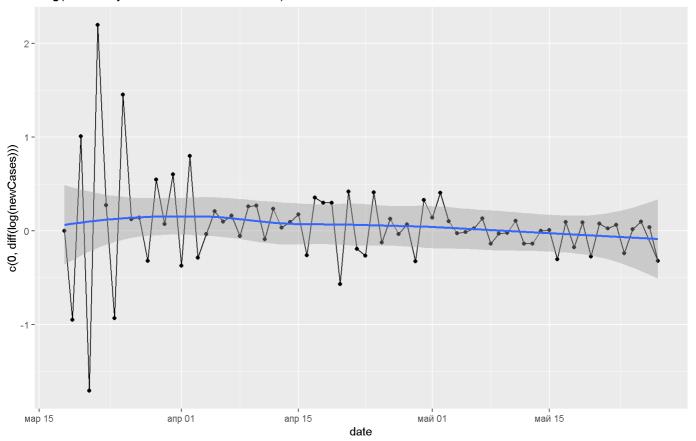
Struct change

```
##
## Optimal 4-segment partition:
##
## Call:
## breakpoints.formula(formula = log(newCases) ~ days_from_beg,
## h = 7/CovidMoscow[, .N], data = CovidMoscow)
##
## Breakpoints at observation number:
## 7 33 45
##
## Corresponding to breakdates:
## 0.09722222 0.4583333 0.625
```

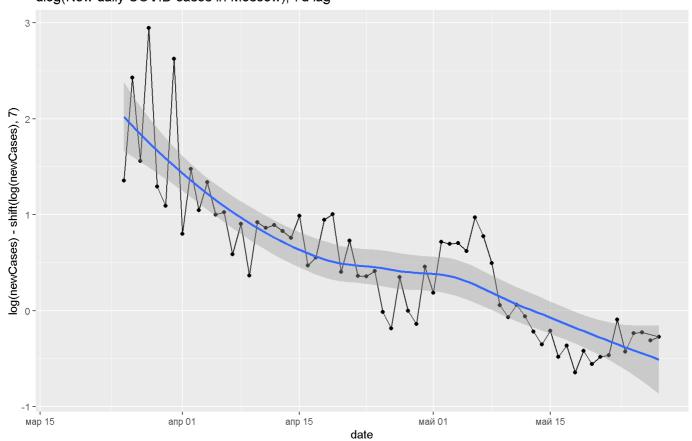


Дифференциалы

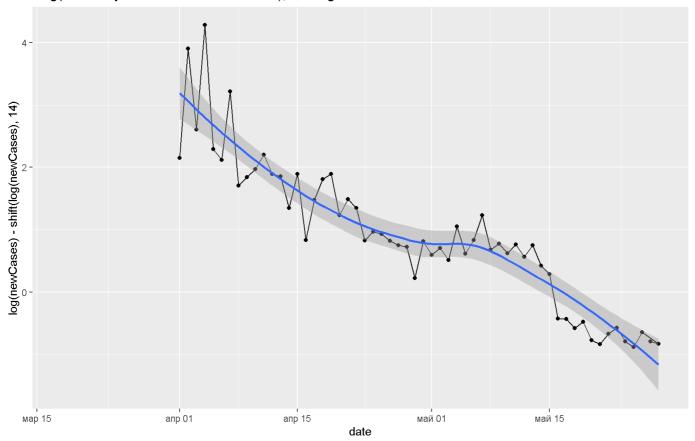
dlog(New daily COVID cases in Moscow)



dlog(New daily COVID cases in Moscow), 7d lag



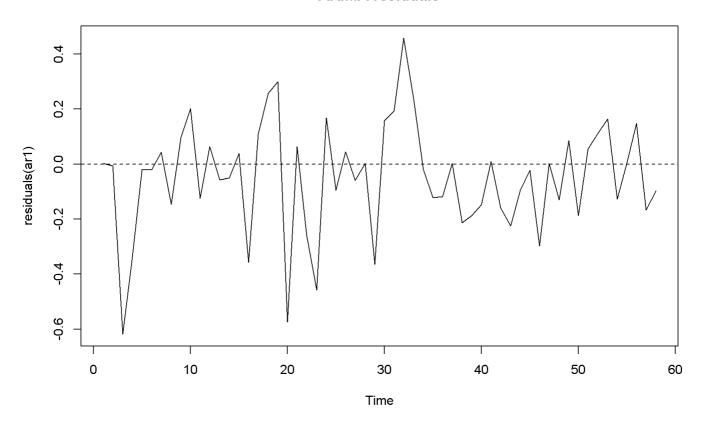
dlog(New daily COVID cases in Moscow), 14d lag



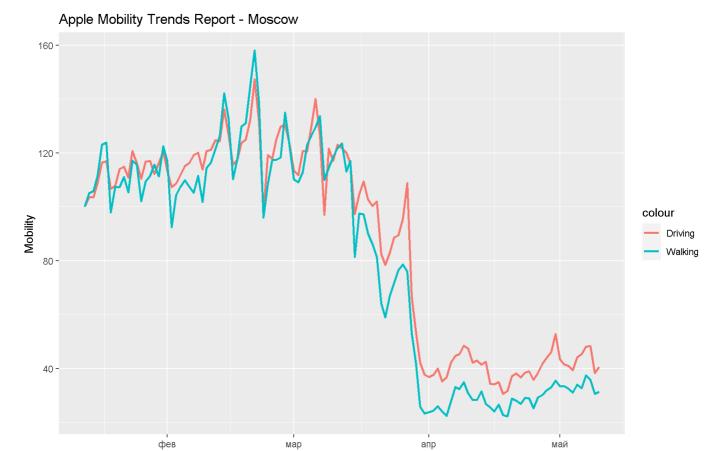
ARMA errors

```
## Series: log(d1$newCases)
\#\# Regression with ARIMA(0,2,2) errors
## Coefficients:
##
                 ma2
        -1.3685 0.4588 -0.0545
##
## s.e. 0.1140 0.1099 0.1103
##
## sigma^2 estimated as 0.04692: log likelihood=6.44
## AIC=-4.89 AICc=-4.11 BIC=3.21
##
## Training set error measures:
               ME
                             RMSE
##
                                       MAE
                                                  MPE
                                                           MAPE
## Training set -0.04907045 0.2070673 0.1530651 -0.6989692 2.003948 0.8605882
##
## Training set 0.06702633
```

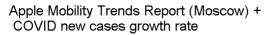
ARIMA residuals

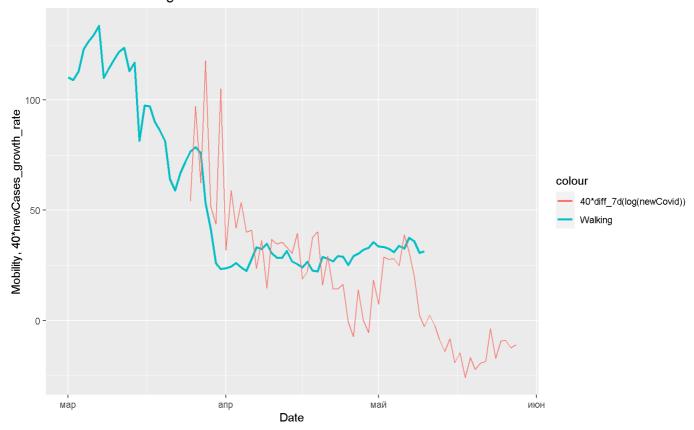


Apple mobility trends



Date





Rolling sum over 2 weeks (aka active cases)



