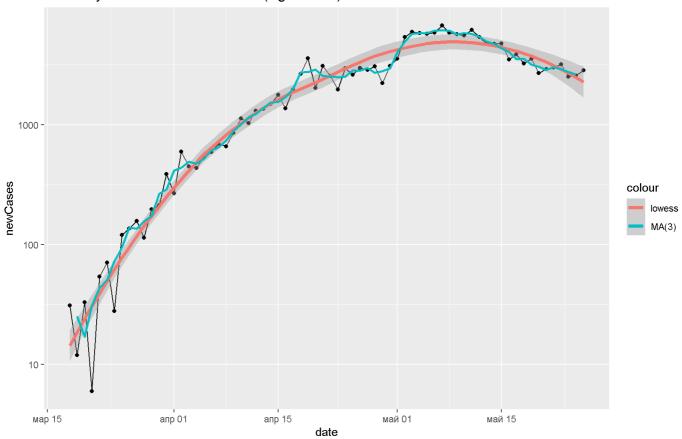
COVID in Moscow

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

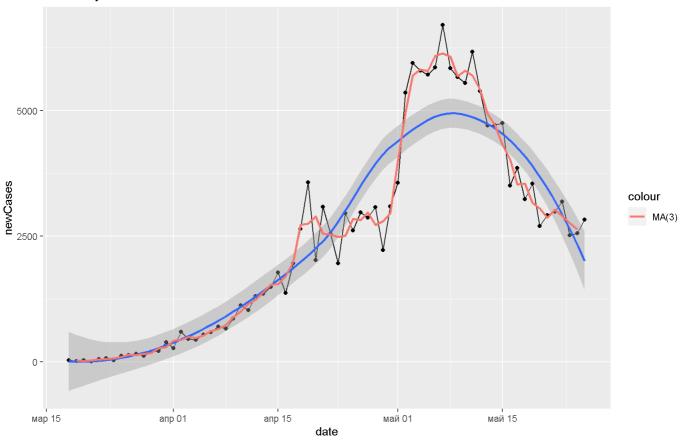
[1] "CovidMoscowDB was NOT updated"

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 10 Median 30
## -3.0490 -0.5902 0.3801 0.6521 1.3083
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.630685 0.211397 21.91 <2e-16 ***
## days from beg 0.070042 0.005287 13.25 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.8938 on 68 degrees of freedom
## Multiple R-squared: 0.7207, Adjusted R-squared: 0.7166
## F-statistic: 175.5 on 1 and 68 DF, p-value: < 2.2e-16
```

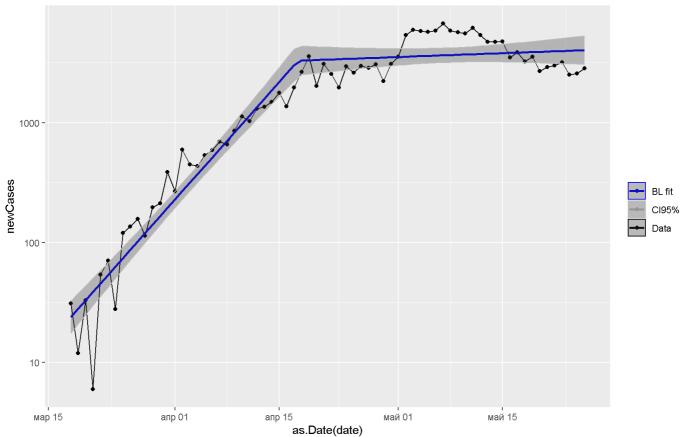
Segmented

```
0.152715 20.80
## (Intercept)
                    3.175955
                                                  <2e-16 ***
## days_from_beg
                                         18.44
                    0.161241
                               0.008744
                                                  <2e-16 ***
                              0.010717 -14.56
## U1.days from beg -0.156035
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4355 on 66 degrees of freedom
## Multiple R-Squared: 0.9357, Adjusted R-squared: 0.9327
##
## Convergence attained in 4 iter. (rel. change 0)
##
## ----- 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 1.237
## psi2.days from beg 49.567 1.489
##
## Meaningful coefficients of the linear terms:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.65796 0.15898 16.719 <2e-16 ***
## days_from_beg 0.23609 0.01806 13.073 <2e-16 ***
## U1.days from beg -0.16261
                               0.01897 -8.570
## U2.days from beg -0.12586
                               0.01416 -8.886
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.333 on 64 degrees of freedom
## Multiple R-Squared: 0.9635, Adjusted R-squared: 0.9607
## Convergence attained in 3 iter. (rel. change 6.1708e-16)
##
## ----- 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 12.999 1.896
## psi2.days from beg 28.011 3.735
## psi3.days_from_beg 51.160 1.894
##
## Meaningful coefficients of the linear terms:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.64535 0.16422 16.108 <2e-16 ***
## days_from_beg 0.23925 0.02147 11.143 <2e-16 ***
## U1.days from beg -0.12695
                               0.02891 -4.392
                               0.02187 -2.593
## U2.days from beg -0.05671
                                                     NΑ
                               0.01789 -6.062
## U3.days from beg -0.10846
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3239 on 62 degrees of freedom
## Multiple R-Squared: 0.9666, Adjusted R-squared: 0.9628
##
## Convergence attained in 9 iter. (rel. change 4.9091e-06)
##
## ----- 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):
```

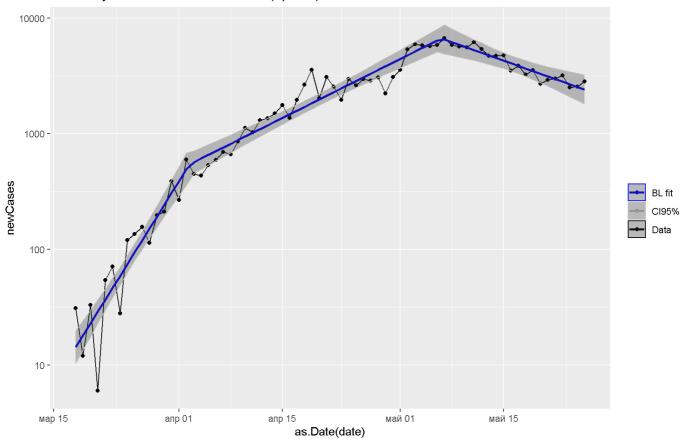
```
##
## psil.days from beg 13.000 1.815
## psi2.days from beg 30.000 3.179
## psi3.days from beg 45.001 3.135
## psi4.days_from_beg 48.000 1.304
##
## Meaningful coefficients of the linear terms:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.64432 0.16448 16.077 < 2e-16 ***
## days_from_beg 0.23950 0.02150 11.137 3.1e-16 ***
## U1.days from beg -0.12816
                           0.02684 -4.775
                          0.02517 -2.898
## U2.days_from_beg -0.07294
## U3.days_from_beg 0.16598 0.23017 0.721
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3244 on 60 degrees of freedom
## Multiple R-Squared: 0.9675, Adjusted R-squared: 0.9627
##
## Convergence attained in 1 iter. (rel. change -3.5944e-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 1.724
## psi2.days from beg 32.000 2.382
## psi3.days_from_beg 42.650 1.509
## psi4.days_from_beg 45.449 1.735
## psi5.days from beg 51.263 4.593
## Meaningful coefficients of the linear terms:
##
             Estimate Std. Error t value Pr(>|t|)
                 ## (Intercept)
## days_from_beg 0.23942 0.02106 11.371 <2e-16 ***
## U1.days from beg -0.12835 0.02491 -5.153
## U3.days from beg 0.27173 0.22728 1.196
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3176 on 58 degrees of freedom
## Multiple R-Squared: 0.9699, Adjusted R-squared: 0.9642
##
## Convergence attained in 12 iter. (rel. change 8.5346e-08)
## breakpoint estimate(s): 0.7288397 12.95754 32.71265 33.94685 42.02732 49.09063
##
## ----- 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 1.768
## psi2.days_from_beg 32.000 2.222
## psi3.days from beg 42.673 1.490
## psi4.days from beg 45.546 1.654
## psi5.days from beg 53.811 4.271
## psi6.days from beg 62.940 5.378
##
## Meaningful coefficients of the linear terms:
```

```
Estimate Std. Error t value Pr(>|t|)
                 ## (Intercept)
## days from beg
                 0.23940
                            0.02130 11.239 5.6e-16 ***
                             0.02582 -4.972
## U1.days_from_beg -0.12837
                                                 NA
## U2.days_from_beg -0.10820
                             0.03393 -3.189
                                                 NA
## U3.days_from_beg 0.27254
                             0.22923
                                     1.189
                                                 NA
## U4.days from beg -0.27642
                             0.23252 -1.189
                                                 NA
## U5.days from beg -0.07413
                             0.06464 -1.147
                                                 NA
## U6.days_from_beg 0.05959
                             0.07353
                                     0.810
                                                 NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3213 on 56 degrees of freedom
## Multiple R-Squared: 0.9703, Adjusted R-squared: 0.9634
##
## Convergence attained in 1 iter. (rel. change 3.5549e-07)
```

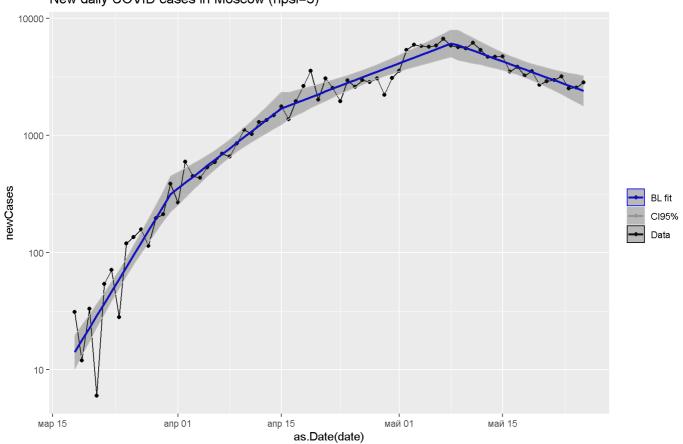
New daily COVID cases in Moscow (npsi=1)



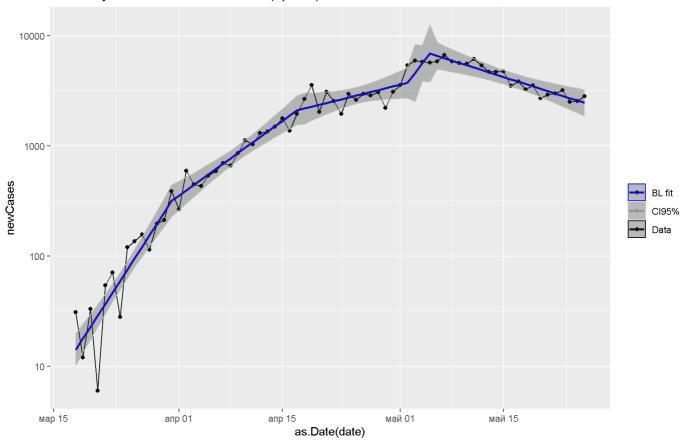
New daily COVID cases in Moscow (npsi=2)



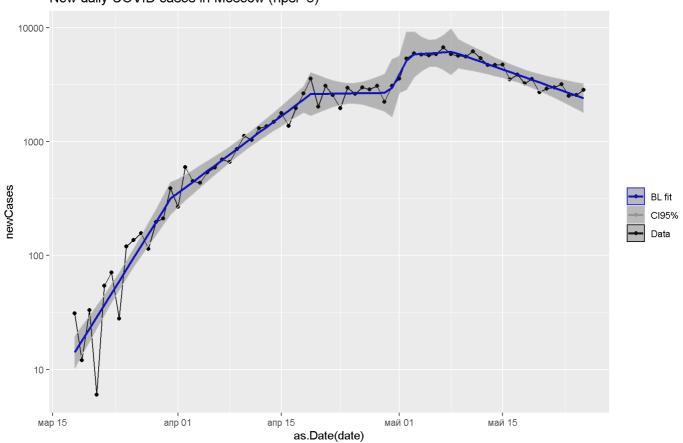
New daily COVID cases in Moscow (npsi=3)



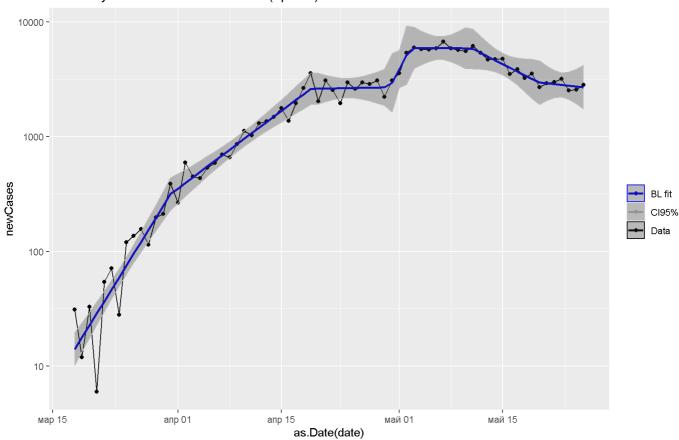
New daily COVID cases in Moscow (npsi=4)



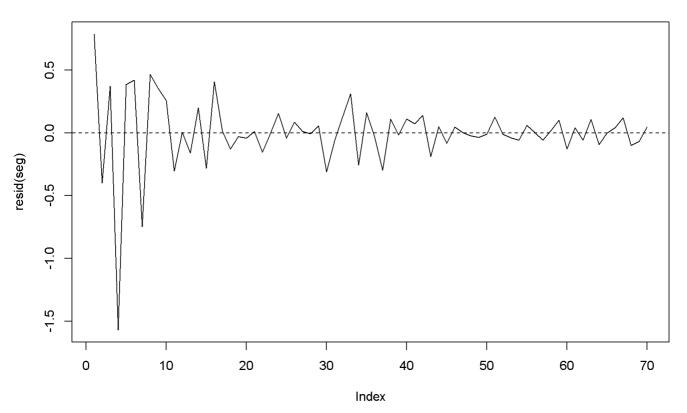
New daily COVID cases in Moscow (npsi=5)



New daily COVID cases in Moscow (npsi=6)



Seg log(newCases) residuals (npsi=6)

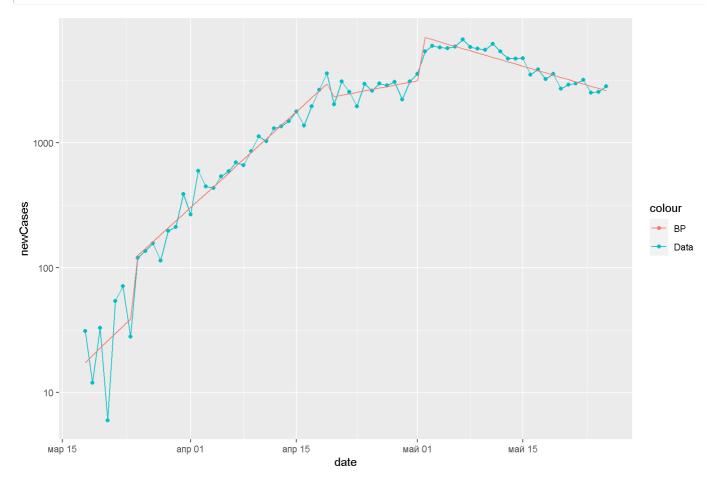


```
##
## ------
N. Breaking points = 1 =-----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.1612400 0.0087444 18.43900 0.1437800 0.178700
## slope2 0.0052063 0.0061957 0.84031 -0.0071639 0.017577
```

```
## Doubling of cases:
## slope1 slope2
   4.298854 133.136235
##
##
## ----- N. Breaking points = 2 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.236090 0.018059 13.0730 0.200010 0.272160
## slope2 0.073471 0.005821 12.6220 0.061842 0.085100
## slope3 -0.052392 0.012913 -4.0573 -0.078189 -0.026595
##
## Doubling of cases:
## slope1 slope2 slope3
   2.935945 9.434296 -13.230019
##
##
## ----- N. Breaking points = 3 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.239250 0.021472 11.1430 0.196330 0.282170
## slope2 0.112300 0.019354 5.8022 0.073608 0.150980
## slope3 0.055584 0.010180 5.4600 0.035234 0.075934
## slope4 -0.052878 0.014713 -3.5940 -0.082289 -0.023467
##
## Doubling of cases:
   slope1 slope2 slope3 slope4
2.897167 6.172281 12.470264 -13.108423
##
##
## ----- N. Breaking points = 4 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.239500 0.021505 11.13700 0.19649000 0.282520
## slope2 0.111340 0.016058 6.93380 0.07922200 0.143460
## slope3 0.038406 0.019384 1.98130 -0.00036775 0.077180
## slope4 0.204390 0.229360 0.89115 -0.25439000 0.663170
## slope5 -0.049135 0.011689 -4.20350 -0.07251700 -0.025754
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.894143 6.225500 18.047888 3.391297 -14.106995
##
##
## ----- N. Breaking points = 5 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.2394200 0.021056 11.371000 0.197270 0.281570
## slope2 0.1110700 0.013302 8.349800 0.084445 0.137700
## slope3 0.0026226 0.034966 0.075004 -0.067369 0.072614
## slope4 0.2743500 0.224570 1.221700 -0.175180 0.723880
## slope5 0.0102560 0.075919 0.135090 -0.141710 0.162220
## slope6 -0.0529130 0.014429 -3.667300 -0.081795 -0.024031
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.895110 6.240634 264.297712 2.526507 67.584553 -13.099752
##
##
## ----- N. Breaking points = 6 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.2394000 0.021300 11.239000 0.196730 0.2820700
## slope2 0.1110300 0.014596 7.607000 0.081791 0.1402700
## slope3 0.0028346 0.030633 0.092536 -0.058530 0.0642000
## slope4 0.2753800 0.227180 1.212200 -0.179710 0.7304700
## slope5 -0.0010377 0.049574 -0.020933 -0.100350 0.0982700
## slope6 -0.0751710 0.041476 -1.812400 -0.158260 0.0079165
## slope7 -0.0155830 0.060715 -0.256650 -0.137210 0.1060400
##
## Doubling of cases:
   slope1 slope2 slope3 slope4
                                                slope5
                                                           slope6
                                                                      slope7
```

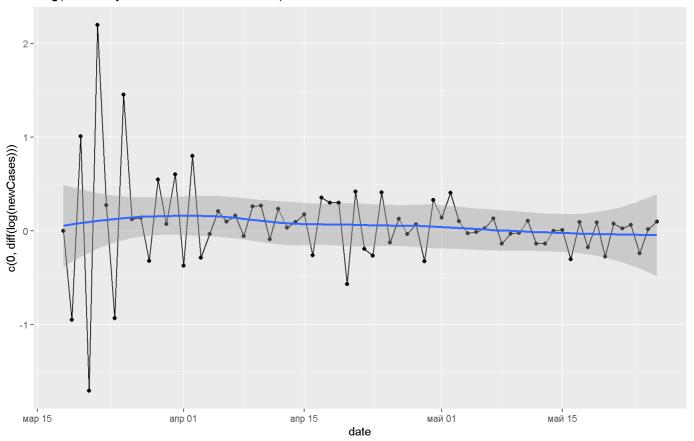
2.895352 6.242882 244.530862 2.517057 -667.964904 -9.220939 -44.480984

Struct change

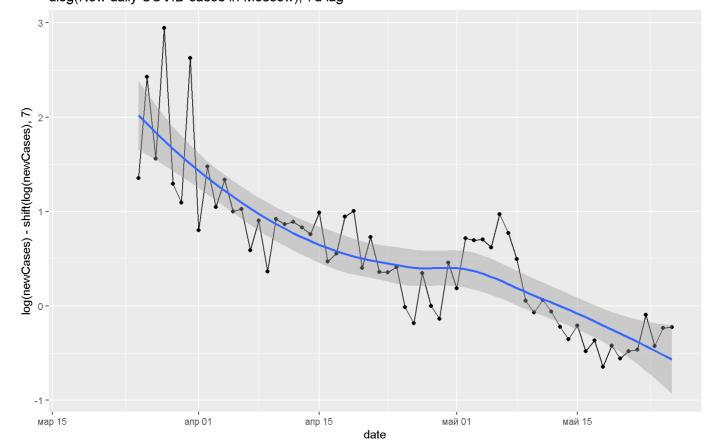


Дифференциалы

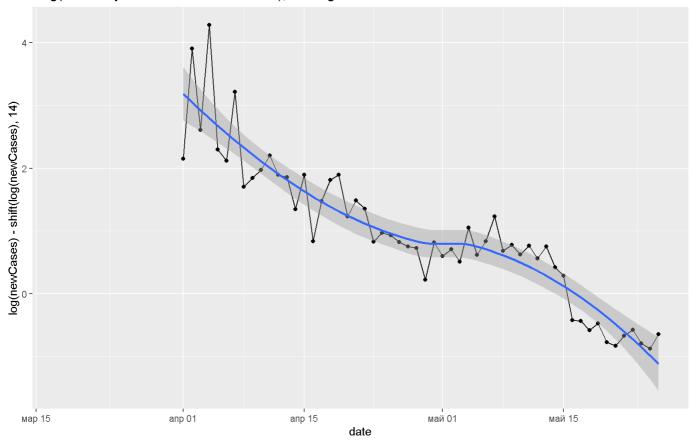
dlog(New daily COVID cases in Moscow)



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



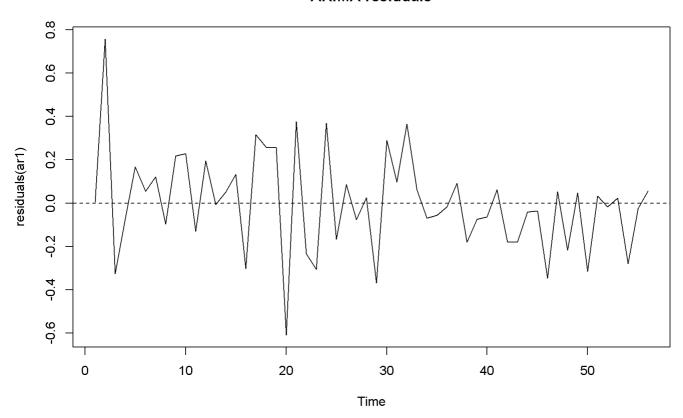
dlog(New daily COVID cases in Moscow), 14d lag



ARMA errors

```
## Series: log(d1$newCases)
## Regression with ARIMA(0,1,0) errors
## Coefficients:
##
        0.0429
##
## s.e. 0.0311
##
\#\# sigma^2 estimated as 0.05428: log likelihood=2.59
## AIC=-1.17 AICc=-0.94 BIC=2.84
##
## Training set error measures:
##
                              RMSE
                                        MAE
                                                    MPE
                  ME
                                                            MAPE
                                                                     MASE
## Training set 8.904142e-05 0.228788 0.1713657 0.03301363 2.268126 0.9640508 -0.3048746
```

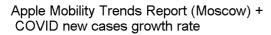
ARIMA residuals

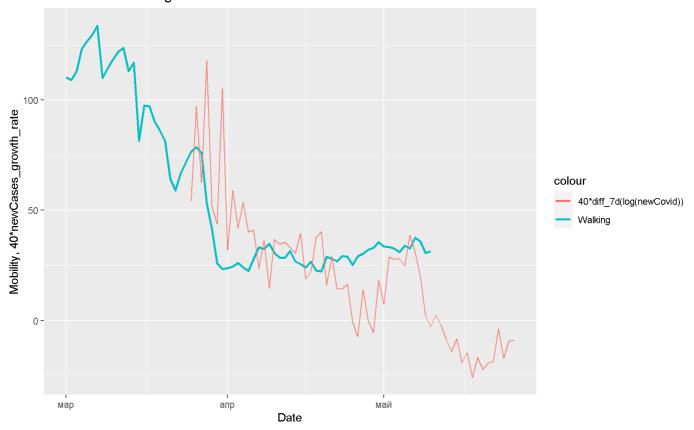


Apple mobility trends









Rolling sum over 2 weeks (aka active cases)

