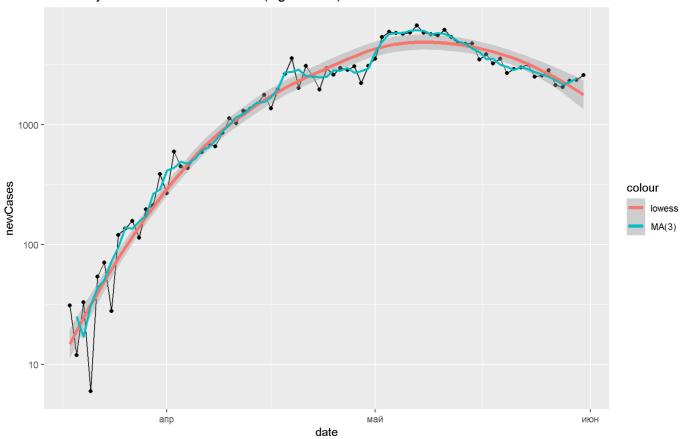
# **COVID** in Moscow

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

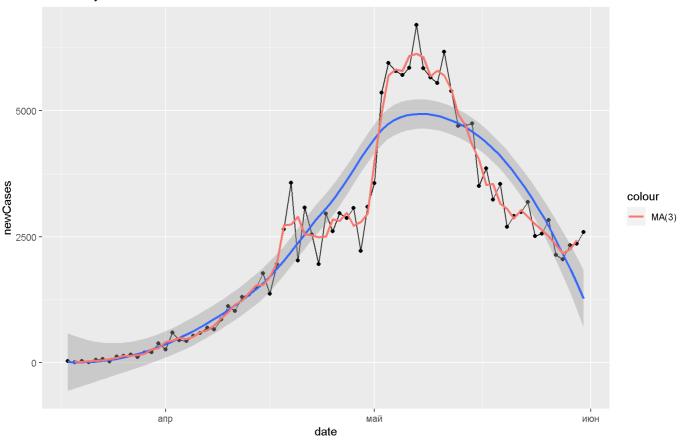
## [1] "CovidMoscowDB was NOT updated"

## 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:
            1Q Median
## Min
                         3Q
## -3.2480 -0.7365 0.3385 0.7486 1.3892
##
## Coefficients:
##
             Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.85860 0.22162 21.92 <2e-16 ***
## days_from_beg 0.06039 0.00517 11.68 <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9692 on 73 degrees of freedom
## Multiple R-squared: 0.6515, Adjusted R-squared: 0.6467
## F-statistic: 136.5 on 1 and 73 DF, p-value: < 2.2e-16
```

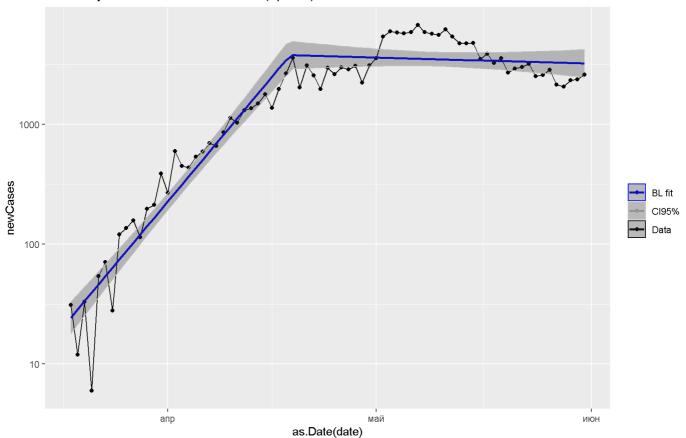
### Segmented

```
0.150909 21.16
## (Intercept)
                    3.192574
                                                  <2e-16 ***
## days_from_beg
                               0.008365 19.08
                    0.159579
                                                  <2e-16 ***
## U1.days from beg -0.163356 0.009939 -16.44
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4369 on 71 degrees of freedom
## Multiple R-Squared: 0.9311, Adjusted R-squared: 0.9282
##
## Convergence attained in 3 iter. (rel. change 1.3012e-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 1.203
## psi2.days from beg 49.239 1.416
##
## Meaningful coefficients of the linear terms:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 2.65796 0.15451 17.202 <2e-16 ***
## days_from_beg 0.23609 0.01755 13.451 <2e-16 ***
## U1.days from beg -0.16261
                               0.01844 -8.818
## U2.days from beg -0.12096
                               0.01061 -11.401
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3236 on 69 degrees of freedom
## Multiple R-Squared: 0.9633, Adjusted R-squared: 0.9606
## Convergence attained in 3 iter. (rel. change 0)
##
## ----- 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 1.879
## psi2.days from beg 27.978 3.880
## psi3.days_from_beg 50.631 1.787
##
## Meaningful coefficients of the linear terms:
##
              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 2.64523 0.15951 16.583 <2e-16 ***
## days_from_beg 0.23928 0.02086 11.473 <2e-16 ***
## U1.days from beg -0.12706
                               0.02949 -4.308
                               0.02308 -2.436
## U2.days from beg -0.05621
                                                     NΑ
## U3.days from beg -0.10285
                               0.01356 -7.586
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3146 on 67 degrees of freedom
## Multiple R-Squared: 0.9663, Adjusted R-squared: 0.9628
##
## Convergence attained in 6 iter. (rel. change 9.7397e-07)
##
## ----- 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):
```

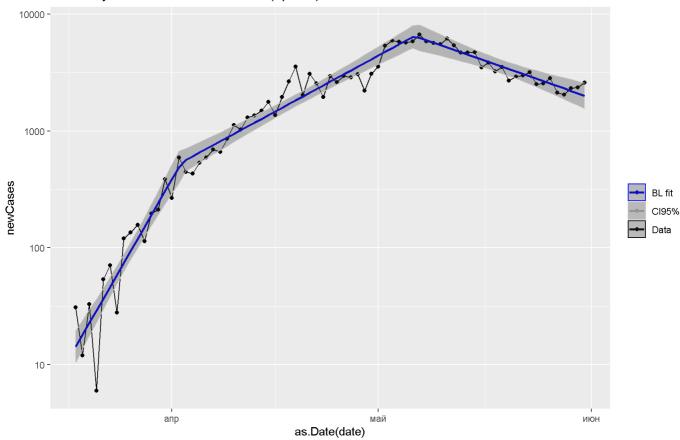
```
##
## psil.days from beg 13.000 1.759
## psi2.days from beg 30.000 3.091
## psi3.days from beg 45.001 3.220
## psi4.days_from_beg 48.000 1.301
##
## Meaningful coefficients of the linear terms:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 2.64425 0.15948 16.580 <2e-16 ***
## days_from_beg 0.23952 0.02085 11.487 <2e-16 ***
## U1.days from beg -0.12822
                            0.02602
                                     -4.927
                           0.02441 -2.980
## U2.days_from_beg -0.07272
                                                 NA
## U3.days_from_beg 0.15671 0.22318 0.702
## U4.days_from_beg -0.24137
                           0.22254 -1.085
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 0.3145 on 65 degrees of freedom
## Multiple R-Squared: 0.9673, Adjusted R-squared: 0.9628
##
## Convergence attained in 1 iter. (rel. change -4.8996e-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 1.696
## psi2.days from beg 31.998 2.119
## psi3.days_from_beg 42.622 1.460
## psi4.days_from_beg 45.345 1.767
## psi5.days from beg 50.159 4.259
## Meaningful coefficients of the linear terms:
##
              Estimate Std. Error t value Pr(>|t|)
                  ## (Intercept)
## days_from_beg 0.23940 0.02042 11.724
                                            <2e-16 ***
## U1.days_from_beg -0.12829 0.02475 -5.183
## U3.days from beg 0.27192 0.21975 1.237
## U5.days from beg -0.06629 0.09781 -0.678
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.308 on 63 degrees of freedom
## Multiple R-Squared: 0.9696, Adjusted R-squared: 0.9643
##
## Convergence attained in 13 iter. (rel. change 1.0157e-06)
##
## ----- 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 13.027
## psi2.days_from_beg 32.001 2.158
## psi3.days from beg 43.393 1.408
## psi4.days from beg 45.260 1.130
## psi5.days from beg 54.000 4.216
## psi6.days_from beg 63.014 5.520
## Meaningful coefficients of the linear terms:
              Estimate Std. Error t value Pr(>|t|)
\#\#
                             A 16040 16 ADD - A
```

```
## (Intercept)
                   2.64450
                             U.1624/ 16.2// < 2e-16 ***
                             0.02298 10.422 3.59e-15 ***
## days_from_beg
                  0.23946
## U1.days_from_beg -0.12901
                              0.02640 -4.886
                                                    NA
## U2.days_from_beg -0.10386
                              0.02900 -3.582
                                                    NA
                                       0.908
## U3.days_from_beg 0.39854
                              0.43914
                                                    NA
## U4.days_from_beg -0.40752
                              0.44098 -0.924
                                                    NA
## U5.days_from_beg -0.07297
                              0.06236 -1.170
                                                    NA
## U6.days from beg 0.05096
                                       1.068
                              0.04771
                                                    NA
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.31 on 61 degrees of freedom
## Multiple R-Squared: 0.9702, Adjusted R-squared: 0.9639
## Convergence attained in 7 iter. (rel. change 4.2113e-06)
```

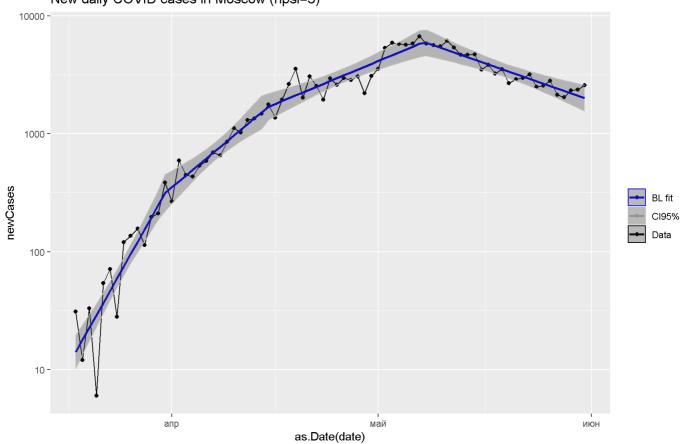
#### New daily COVID cases in Moscow (npsi=1)



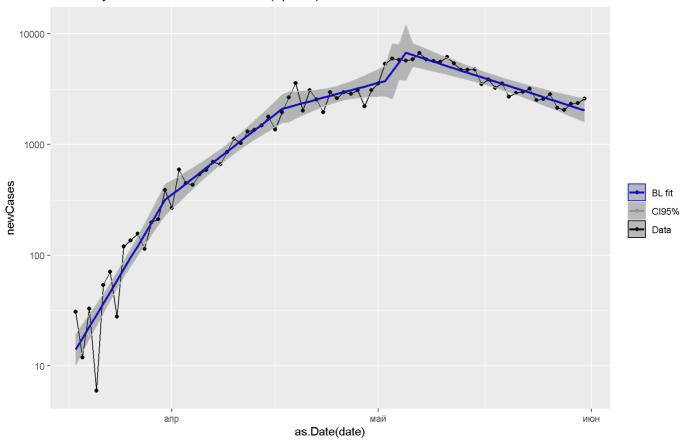
### New daily COVID cases in Moscow (npsi=2)



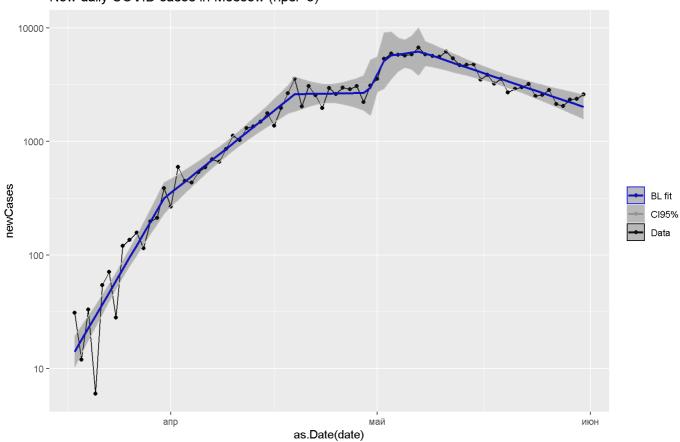
### New daily COVID cases in Moscow (npsi=3)



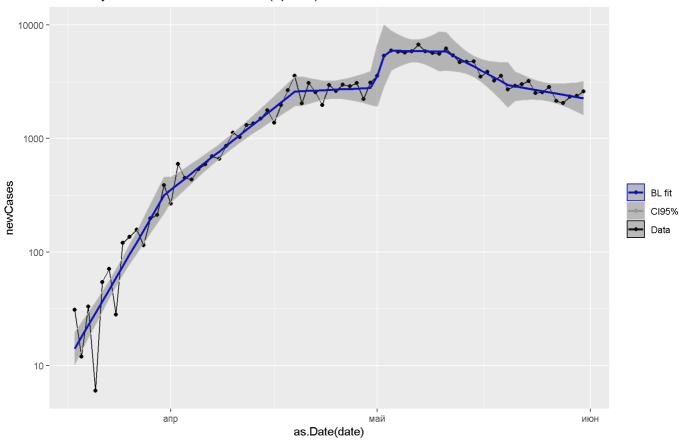
### New daily COVID cases in Moscow (npsi=4)



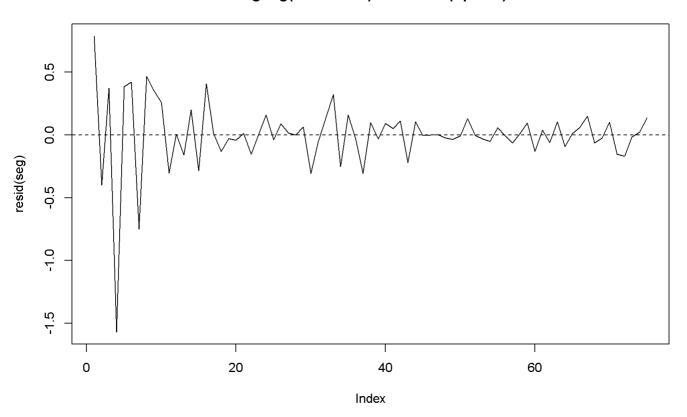
### New daily COVID cases in Moscow (npsi=5)



#### 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%).l CI(95%).u
## slope1 0.1595800 0.0083645 19.07800 0.142900 0.1762600
## slope2 -0.0037765 0.0053687 -0.70342 -0.014481 0.0069284
```

```
## Doubling of cases:
## slope1 slope2
   4.343572 -183.542216
##
##
## ----
      ----- N. Breaking points = 2 =-----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.236090 0.0175510 13.4510 0.201070 0.271100
## slope2 0.073471 0.0056573 12.9870 0.062185 0.084757
## slope3 -0.047493 0.0089759 -5.2911 -0.065399 -0.029586
##
## Doubling of cases:
## slope1 slope2 slope3
   2.935945 9.434296 -14.594723
##
##
## ----- N. Breaking points = 3 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.239280 0.0208550 11.4730 0.197650 0.280900
## slope2 0.112210 0.0208550 5.3806 0.070586 0.153840
## slope3 0.056000 0.0098881 5.6634 0.036263 0.075737
## slope4 -0.046853 0.0092759 -5.0511 -0.065368 -0.028339
##
## Doubling of cases:
   slope1 slope2 slope3 slope4
2.896804 6.177232 12.377628 -14.794083
##
##
## ----- N. Breaking points = 4 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.239520 0.0208510 11.48700 0.1978800 0.281160
## slope2 0.111300 0.0155700 7.14850 0.0802070 0.142400
## slope3 0.038579 0.0187950 2.05260 0.0010425 0.076115
## slope4 0.195290 0.2223900 0.87815 -0.2488500 0.639420
## slope5 -0.046080 0.0082238 -5.60330 -0.0625040 -0.029656
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.893901 6.227737 17.966956 3.549322 -15.042257
##
##
## ----- N. Breaking points = 5 =----
## Slopes
## $days from beg
## Est. St.Err. t value CI(95%).l CI(95%).u
## slope1 0.2394000 0.0204190 11.724000 0.198600 0.280200
## slope2 0.1111100 0.0139920 7.941200 0.083152 0.139070
## slope3 0.0023568 0.0293650 0.080261 -0.056324 0.061038
## slope4 0.2742800 0.2177700 1.259400 -0.160910 0.709460
## slope5 0.0191410 0.0973920 0.196540 -0.175480 0.213760
## slope6 -0.0471460 0.0090818 -5.191300 -0.065295 -0.028998
##
## Doubling of cases:
## slope1 slope2 slope3 slope4 slope5
   2.895352 6.238387 294.105219 2.527152 36.212694 -14.702142
##
##
## ----- N. Breaking points = 6 =----
## Slopes
## $days_from_beg
## Est. St.Err. t value CI(95%).1 CI(95%).u
## slope1 0.2394600 0.022977 10.422000 0.193520 0.2854100
## slope2 0.1104600 0.012989 8.503800 0.084484 0.1364300
## slope3 0.0065973 0.025923 0.254500 -0.045239 0.0584330
## slope4 0.4051400 0.438380 0.924170 -0.471450 1.2817000
## slope5 -0.0023867 0.047831 -0.049898 -0.098030 0.0932570
## slope6 -0.0753560 0.040018 -1.883000 -0.155380 0.0046656
## slope7 -0.0243980 0.025936 -0.940710 -0.076261 0.0274640
##
## Doubling of cases:
    slope1 slope2 slope3 slope4
                                                slope5
```

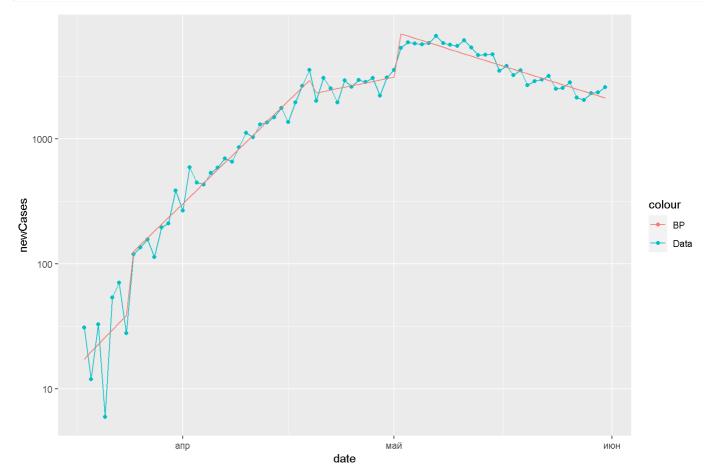
```
## 2.894626 6.275097 105.065281 1.710883 -290.420740 -9.198301

## slope7

## -28.410000
```

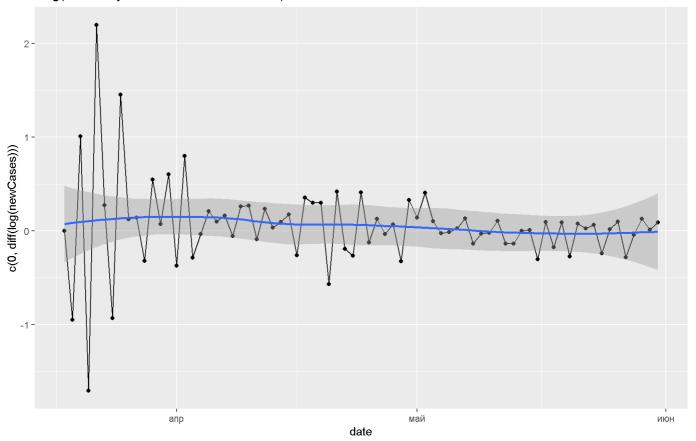
# Struct change

```
##
## 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:
## 0.09333333 0.44 0.6
```

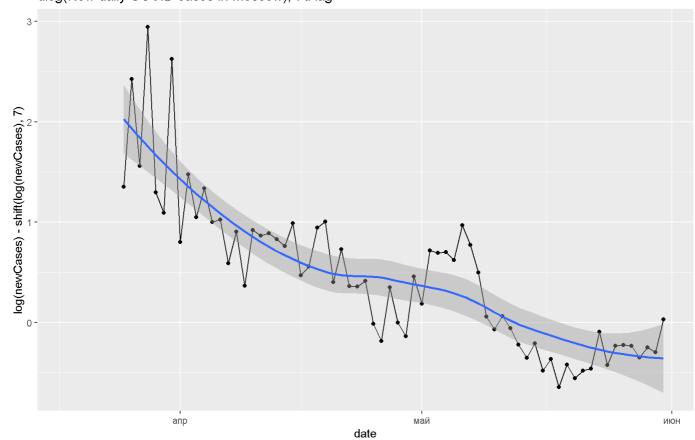


# Дифференциалы

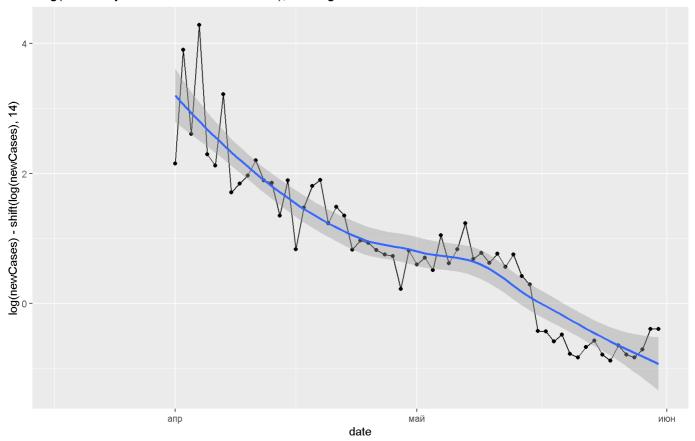
### dlog(New daily COVID cases in Moscow)



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



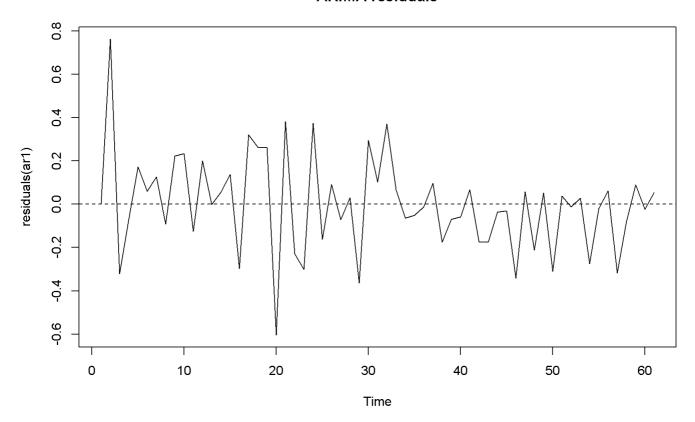
#### dlog(New daily COVID cases in Moscow), 14d lag



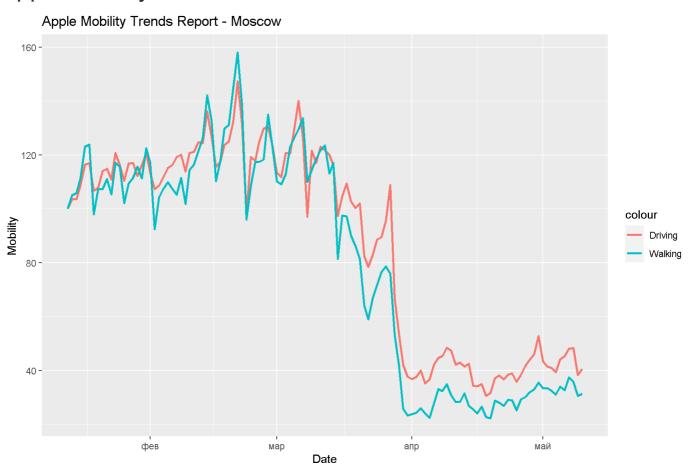
### ARMA errors

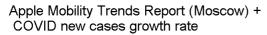
```
## Series: log(d1$newCases)
## Regression with ARIMA(0,1,0) errors
## Coefficients:
##
        0.0379
##
## s.e. 0.0291
##
\#\# sigma^2 estimated as 0.05171: log likelihood=4.23
## AIC=-4.46 AICc=-4.25 BIC=-0.27
## Training set error measures:
                                RMSE
##
                                          MAE
                                                      MPE
                                                              MAPE
                      ME
## Training set 8.289548e-05 0.2236428 0.1664577 0.02941686 2.201469 0.9667602
##
## Training set -0.2954424
```

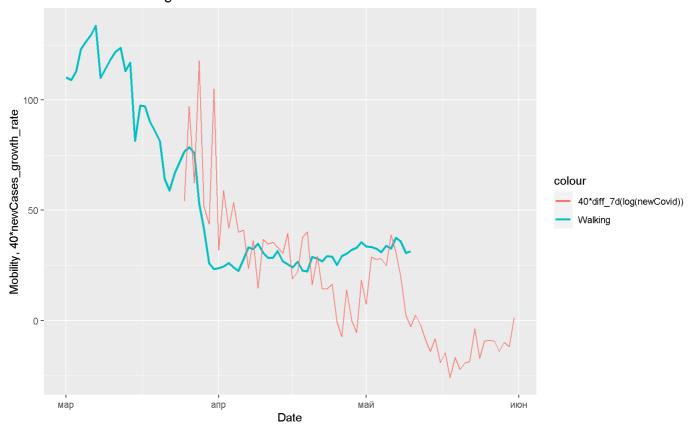
### **ARIMA** residuals



# Apple mobility trends







## Rolling sum over 2 weeks (aka active cases)

14d sum daily COVID-19 cases in Moscow (aka active cases)

