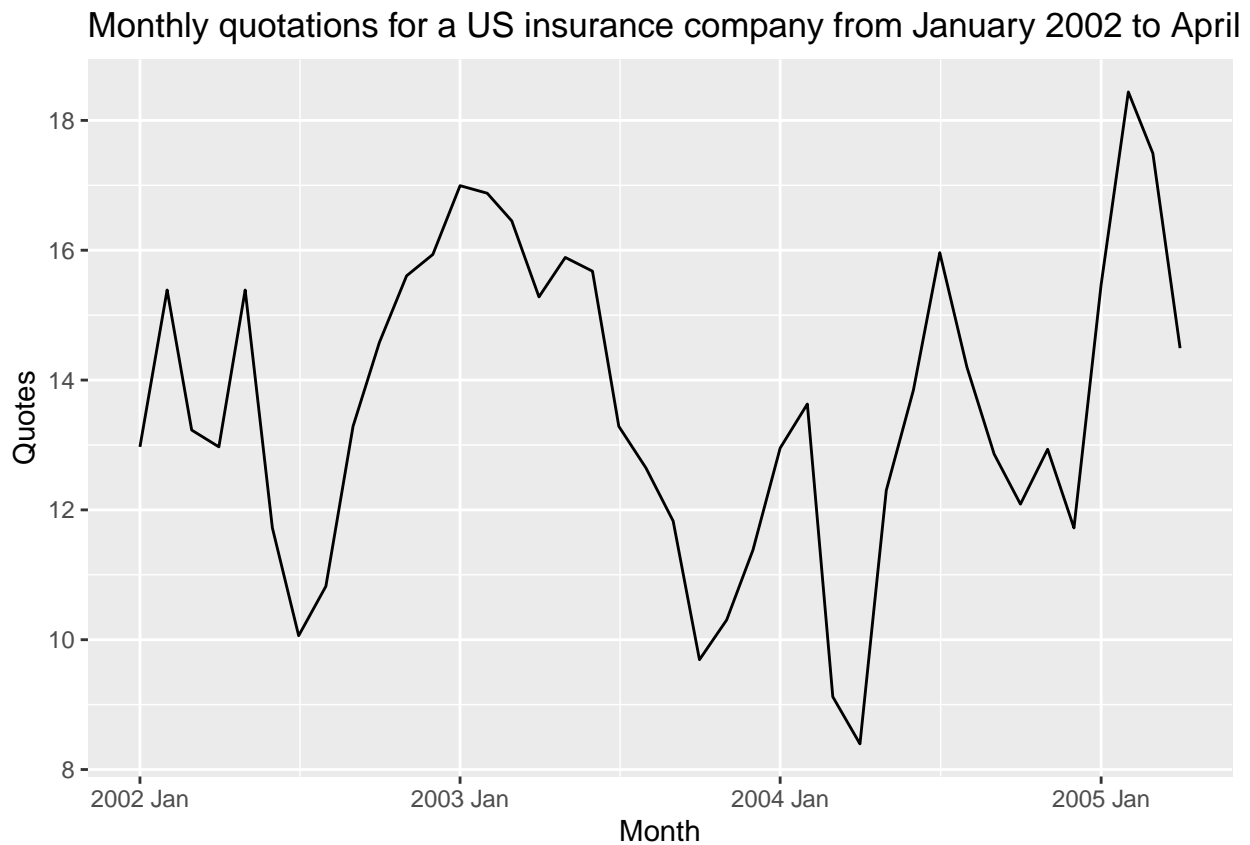


Time Series Project

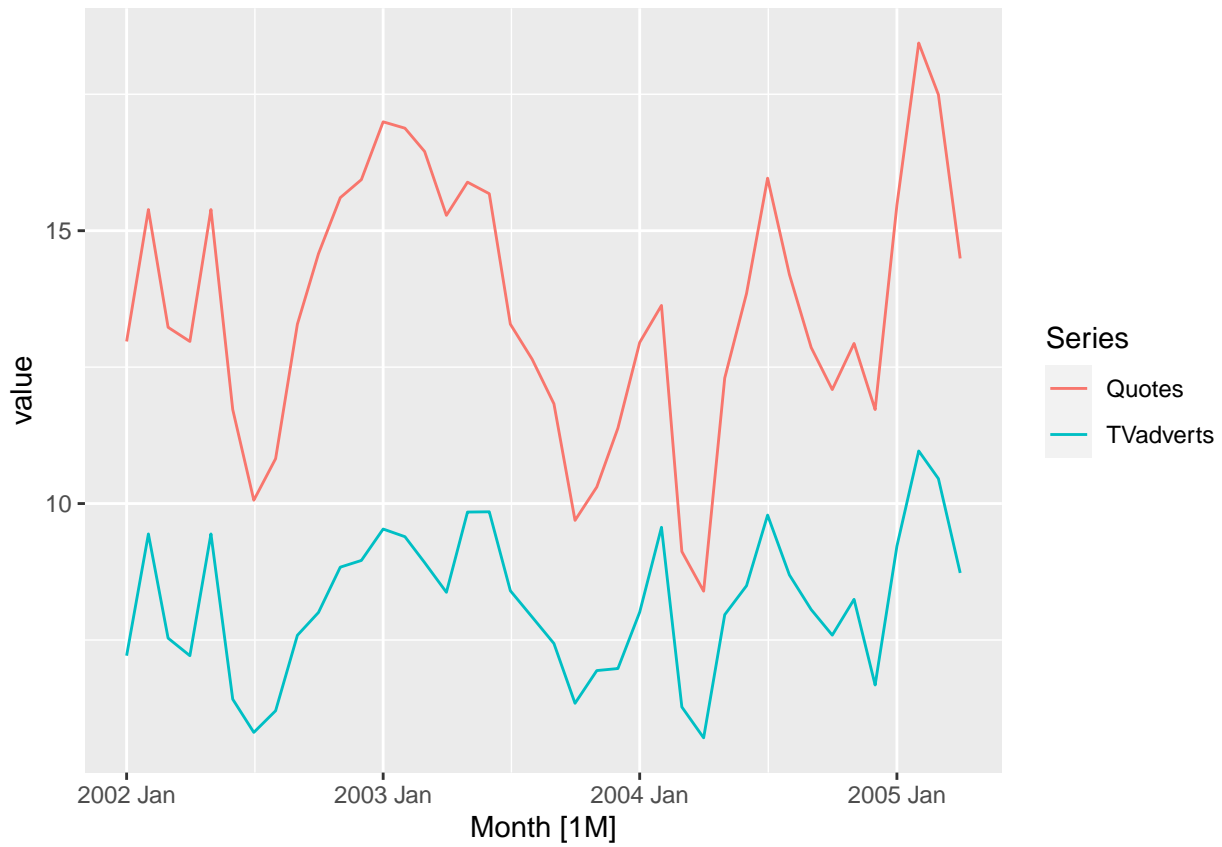
2023-11-17

```
library(fpp3)
library(prophet)
library(fable.prophet)
```

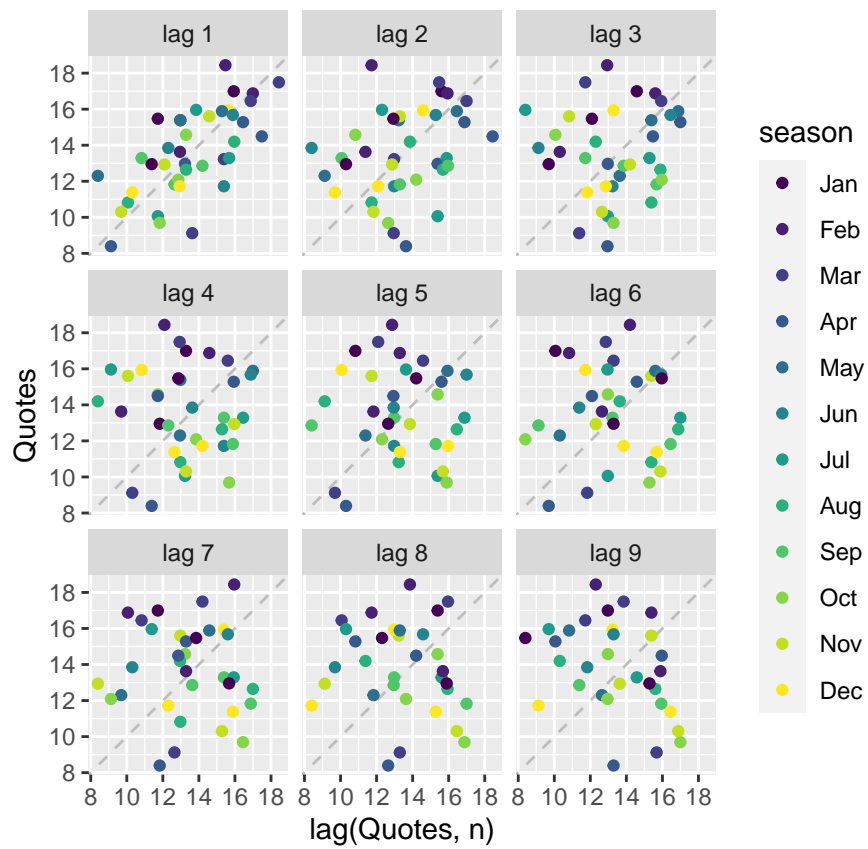
```
insurance |> autoplot(Quotes)+
  labs(y = "Quotes", x = "Month",
       title = "Monthly quotations for a US insurance company from January 2002 to April 2005")
```



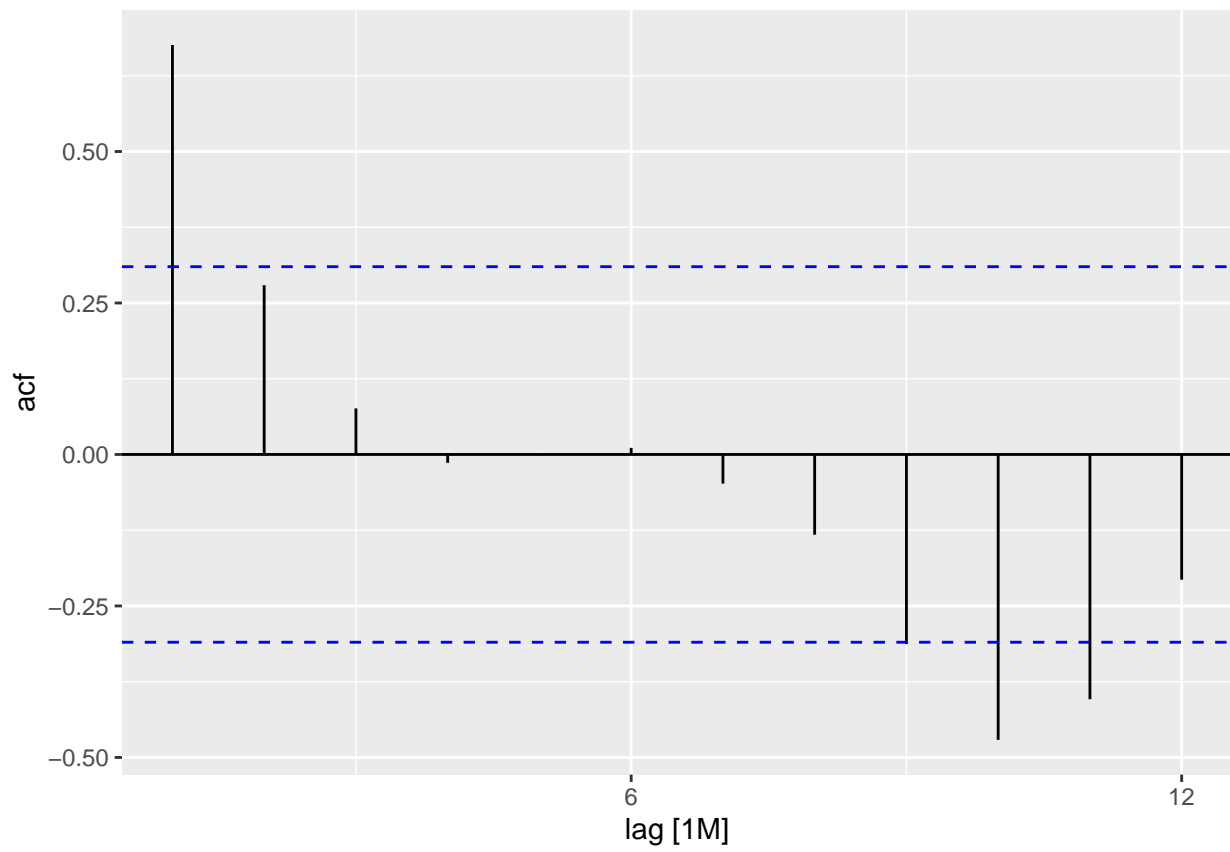
```
insurance |>
  pivot_longer(c(Quotes, TVadverts), names_to="Series") |>
  autoplot(value)
```



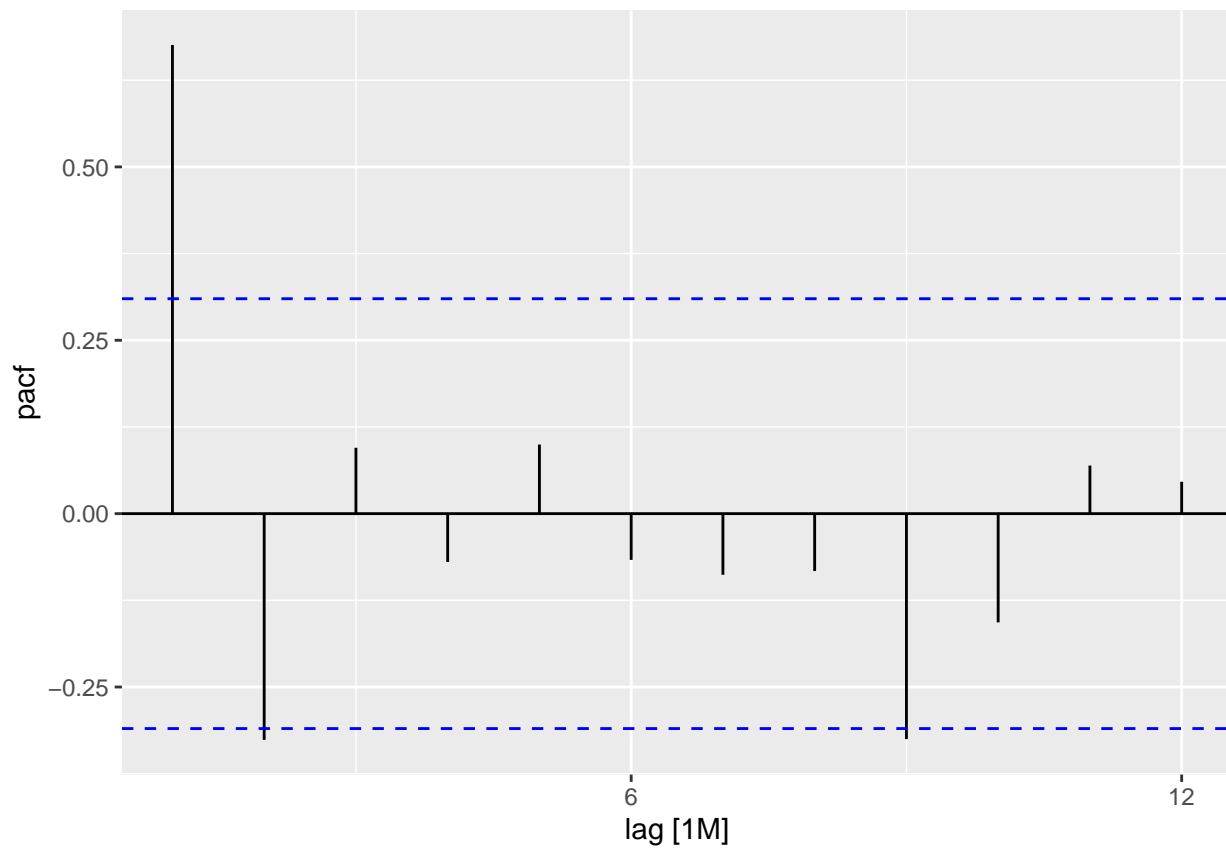
```
insurance |> gg_lag(Quotes, geom='point')
```



```
insurance |> ACF(Quotes, lag_max = 12) |> autoplot()
```



```
insurance |> PACF(Quotes, lag_max = 12) |> autoplot()
```



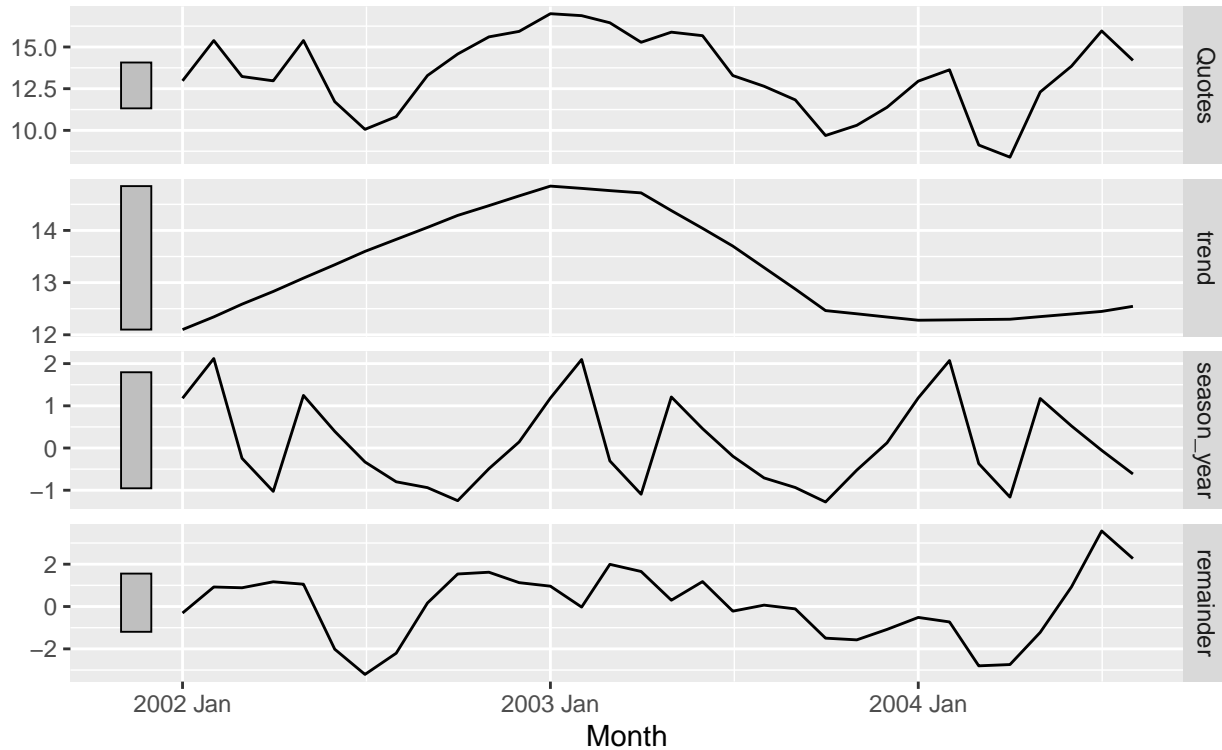
```
Train <- insurance[0:32,]
Test <- insurance[33:40,]
```

```
ins_stl <- Train |>
  model(stl = STL(Quotes))
```

```
ins_stl |>
  components() |>
  autoplot()
```

STL decomposition

Quotes = trend + season_year + remainder



Fitting Models

```
fit1 <- Train |> model(
  lm      = TSLM(Quotes),
  lm2     = TSLM(Quotes ~ TVadverts),
  lm3     = TSLM(Quotes ~ lag(TVadverts)),
  lm4     = TSLM(Quotes ~ TVadverts + lag(TVadverts)),
  ARIMA   = ARIMA(Quotes),
  ARIMA2  = ARIMA(Quotes ~ TVadverts),
  ARIMA3  = ARIMA(Quotes ~ lag(TVadverts)),
  ARIMA4  = ARIMA(Quotes ~ TVadverts + lag(TVadverts)),
  Mean    = MEAN(Quotes),
  SNaive  = SNAIVE(Quotes),
  Naive   = NAIVE(Quotes),
  Drift   = RW(Quotes ~ drift()),
  ETS     = ETS(Quotes),
  NNET    = NNETAR(Quotes),
  Prophet = prophet(Quotes)
)
```

```
## n.changepoints greater than number of observations. Using 24
```

```
## code for methods in class "Rcpp_model_base" was not checked for suspicious field assignments (recomm
```

```
## code for methods in class "Rcpp_model_base" was not checked for suspicious field assignments (recomm
```

```
## code for methods in class "Rcpp_stan_fit" was not checked for suspicious field assignments (recommen
```

```
## code for methods in class "Rcpp_stan_fit" was not checked for suspicious field assignments (recommen
```

```
acc1 <- accuracy(fit1) |> arrange(RMSE)
acc1
```

```
## # A tibble: 15 x 10
##   .model .type      ME  RMSE  MAE    MPE  MAPE  MASE  RMSSE  ACF1
##   <chr>  <chr>    <dbl> <dbl> <dbl>  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 NNET   Trai~ -6.15e- 5 0.0261 0.0165 -3.38e-3 0.119 0.00487 0.00683 -0.555
## 2 ARIMA4 Trai~ -2.04e- 2 0.510 0.406 -1.73e-2 3.17 0.120 0.133 0.00799
## 3 ARIMA2 Trai~ -1.75e- 2 0.536 0.416 -1.38e-1 3.23 0.122 0.140 0.0227
## 4 Proph~ Trai~ -7.94e- 5 0.723 0.565 -3.83e-1 4.24 0.166 0.189 0.319
## 5 lm4    Trai~ -1.11e-16 0.926 0.804 -5.51e-1 6.24 0.237 0.242 0.744
## 6 lm2    Trai~ 0      0.948 0.842 -5.79e-1 6.50 0.248 0.248 0.653
## 7 ARIMA3 Trai~ -1.35e- 2 1.48 1.18 -1.69e+0 9.42 0.348 0.388 0.0336
## 8 ARIMA  Trai~ 2.25e- 2 1.56 1.21 -1.40e+0 9.56 0.358 0.409 -0.0168
## 9 ETS    Trai~ 3.83e- 2 1.83 1.45 -8.65e-1 11.7 0.428 0.478 0.0618
## 10 Drift Trai~ 7.02e-16 1.86 1.50 -1.20e+0 12.1 0.442 0.486 0.0626
## 11 Naive Trai~ 3.96e- 2 1.86 1.50 -8.92e-1 12.1 0.442 0.486 0.0626
## 12 lm3    Trai~ -2.65e-16 2.09 1.75 -2.82e+0 14.2 0.517 0.546 0.395
## 13 lm     Trai~ 0      2.33 1.92 -3.46e+0 15.6 0.566 0.609 0.689
## 14 Mean   Trai~ 5.90e-16 2.33 1.92 -3.46e+0 15.6 0.566 0.609 0.689
## 15 SNaive Trai~ -9.17e- 1 3.82 3.40 -1.26e+1 28.8 1      1      0.755
```

```
print(c("The best mdoel on Train is: ", acc1$.model[1]))
```

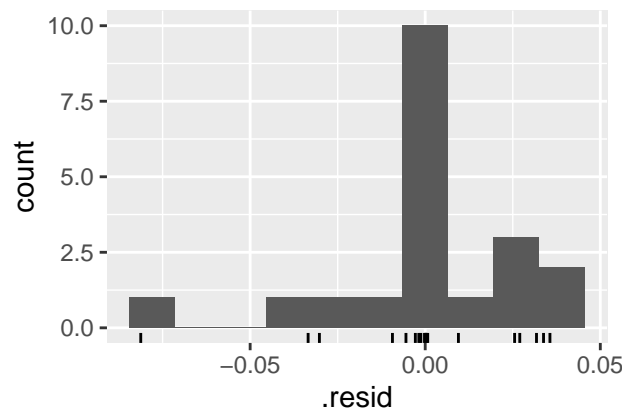
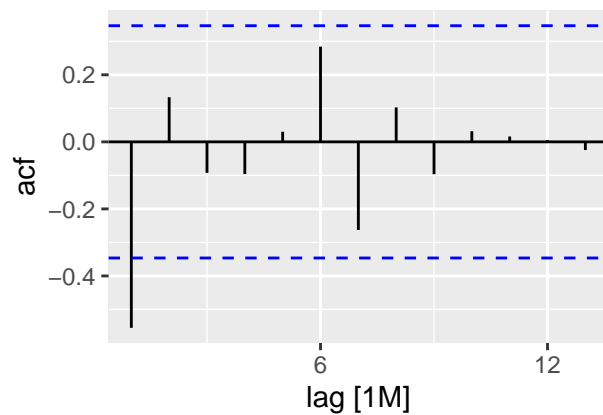
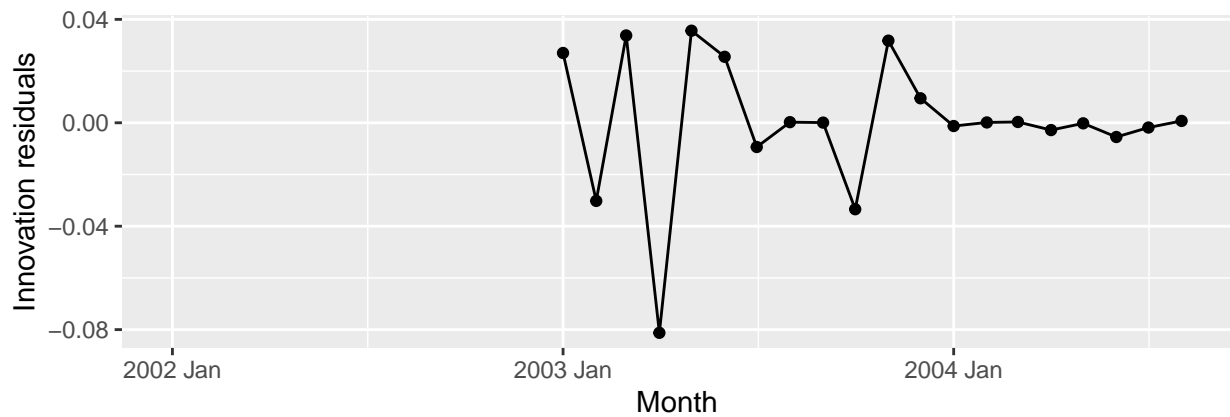
```
## [1] "The best mdoel on Train is: " "NNET"
```

```
fit1[,acc1$.model[1]] |> gg_tsresiduals()
```

```
## Warning: Removed 12 rows containing missing values (`geom_line()`).
```

```
## Warning: Removed 12 rows containing missing values (`geom_point()`).
```

```
## Warning: Removed 12 rows containing non-finite values (`stat_bin()`).
```

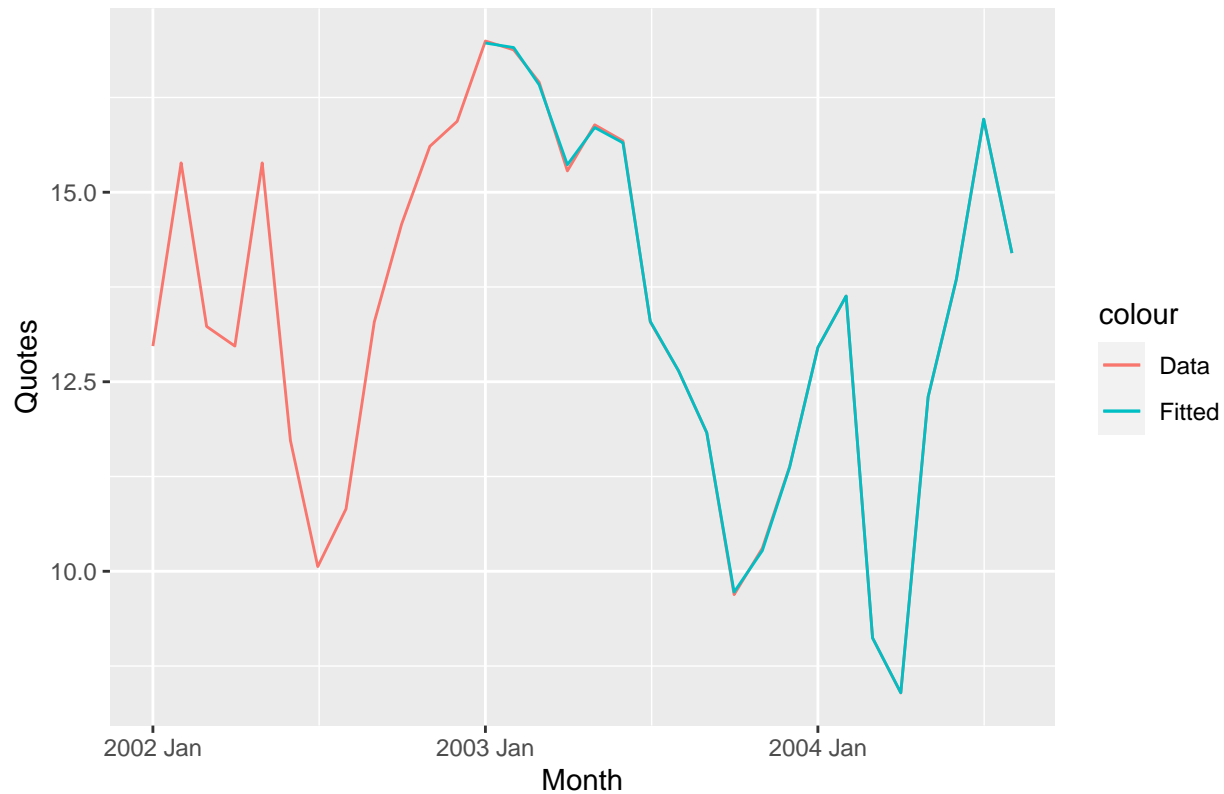


```
# augment(fit1) |> filter(.model == acc1$.model[1]) |>
#   features(.resid, ljung_box, lag=12, dof=0)
#
# augment(fit1) |> filter(.model == acc1$.model[1]) |>
#   features(.innov, box_pierce, lag = 12, dof = 0)
```

```
augment(fit1) |> filter(.model == acc1$.model[1]) |>
  ggplot(aes(x = Month)) +
  geom_line(aes(y = Quotes , colour = "Data")) +
  geom_line(aes(y = .fitted, colour = "Fitted"))+
  ggtitle("The best model of training and actual data")
```

```
## Warning: Removed 12 rows containing missing values (`geom_line()`).
```


The best model of training and actual data



```
fc1 <- fit1 |> forecast(new_data = Test)
```

```
accT1 <- accuracy(fc1, Test) |> arrange(RMSE)
accT1
```

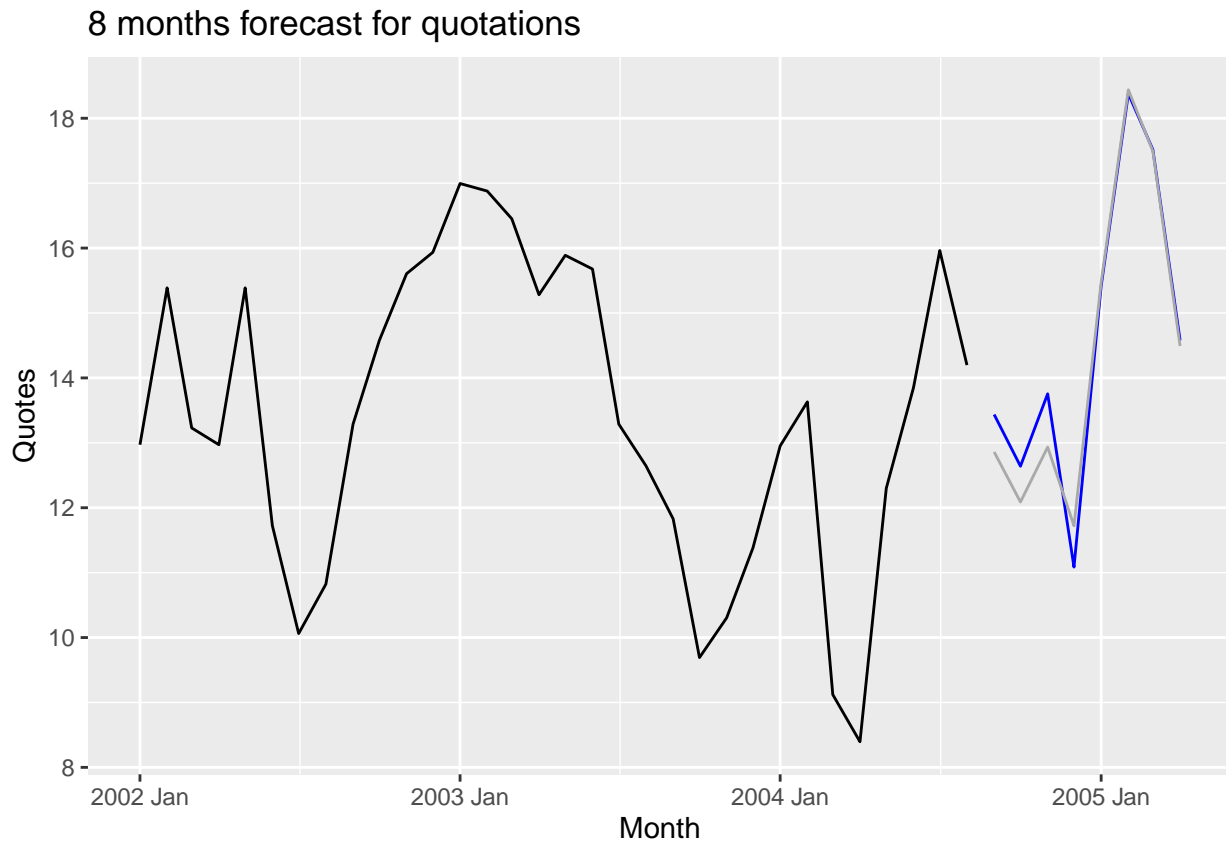
```
## # A tibble: 15 x 10
##   .model .type    ME  RMSE  MAE    MPE  MAPE  MASE  RMSSE  ACF1
##   <chr>  <chr>    <dbl> <dbl> <dbl>  <dbl> <dbl> <dbl> <dbl>
## 1 lm2    Test  -0.163  0.465  0.352  -1.25  2.78  NaN    NaN    0.104
## 2 lm4    Test  -0.153  0.475  0.435  -1.10  3.23  NaN    NaN    0.240
## 3 ARIMA2 Test   0.451  0.935  0.769   2.40  4.93  NaN    NaN    0.693
## 4 ARIMA4 Test   0.650  0.943  0.713   4.00  4.51  NaN    NaN    0.657
## 5 lm3    Test   0.609  2.24   1.87   2.19 12.4   NaN    NaN    0.213
## 6 Drift  Test   0.0617 2.29   1.96  -2.02 13.5   NaN    NaN    0.551
## 7 ETS    Test   0.240  2.36   2.04  -0.831 13.9   NaN    NaN    0.568
## 8 Naive  Test   0.240  2.36   2.04  -0.829 13.9   NaN    NaN    0.568
## 9 Mean   Test   1.04   2.57   2.04   4.86 13.1   NaN    NaN    0.568
## 10 lm    Test   1.04   2.57   2.04   4.86 13.1   NaN    NaN    0.568
## 11 ARIMA Test   1.28   2.59   1.97   6.66 12.5   NaN    NaN    0.542
## 12 ARIMA3 Test   1.45   3.01   2.27   7.56 14.3   NaN    NaN    0.624
## 13 NNET   Test  -1.66   4.06   3.45  -15.4 25.5   NaN    NaN    0.480
## 14 SNaive Test   3.52   4.34   3.52  22.9 22.9   NaN    NaN    0.533
## 15 Prophet Test   4.33  10.5   8.64  31.9 65.6   NaN    NaN   -0.210
```

```
print(c("The best mdoel on Test is: ", accT1$.model[1]))
```

```
## [1] "The best mdoel on Test is: " "lm2"
```

```
fc1 |> filter(.model == accT1$.model[1]) |>
  autoplot(Train, level = NULL) +
  labs(y = "Quotes", x = "Month",
  title = "8 months forecast for quotations") +
  autolayer(Test, colour = "darkgray")
```

Plot variable not specified, automatically selected ``.vars = Quotes``



Fitting Combination Models

```
fit2 <- fit1 |> mutate(
  Combination1 = (lm2 + lm4)/2,
  Combination2 = (lm2 + ARIMA2)/2,
  Combination3 = (lm2 + ARIMA4)/2,
  Combination4 = (lm4 + ARIMA2)/2,
  Combination5 = (lm4 + ARIMA4)/2,
  Combination6 = (ARIMA2 + ARIMA4)/2,
  Combination7 = (NNET + ARIMA4)/2,
  Combination8 = (NNET + ARIMA2)/2,
  Combination8 = (NNET + Prophet)/2,
  Combination8 = (ARIMA4 + ARIMA2)/2,
  Combination8 = (ARIMA4 + Prophet)/2,
  Combination8 = (ARIMA2 + Prophet)/2,
)

acc2 <- accuracy(fit2) |> arrange(RMSE)
```

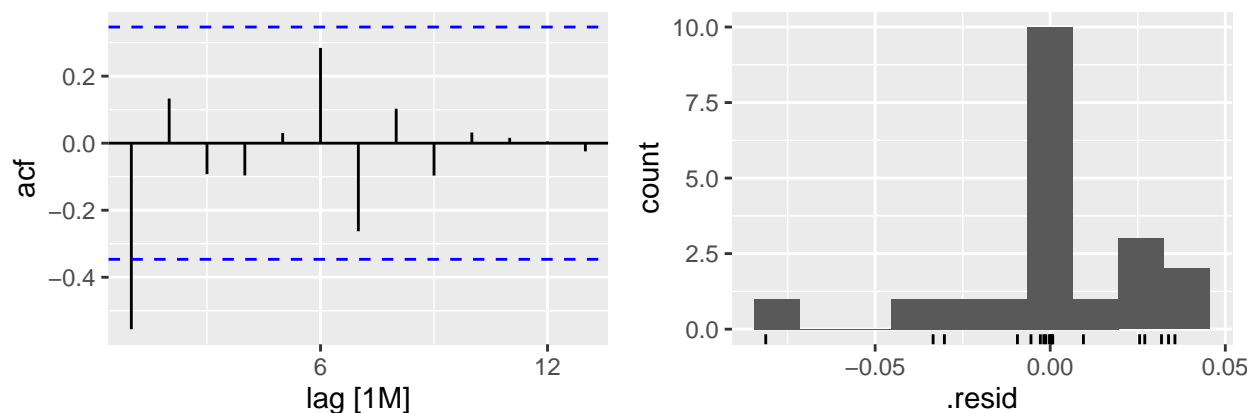
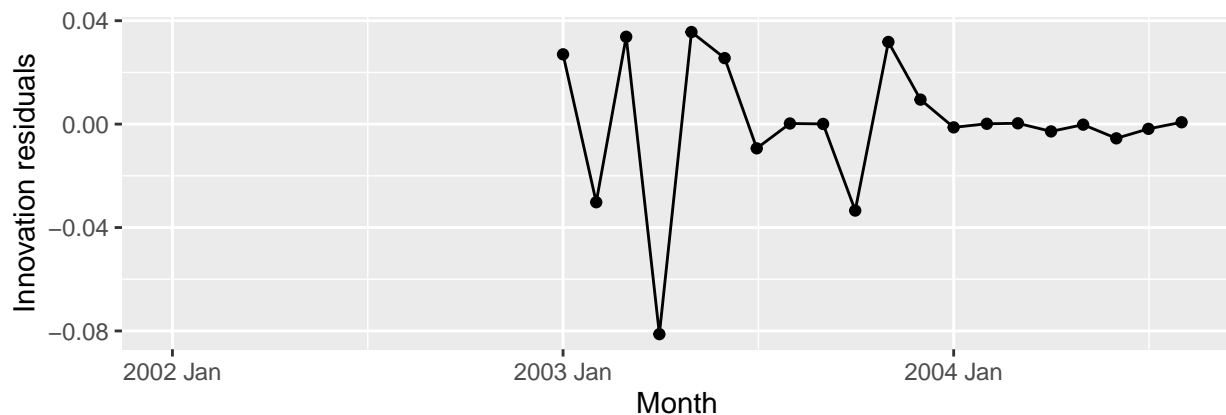
```
acc2
```

```
## # A tibble: 23 x 10
##   .model .type      ME  RMSE  MAE      MPE  MAPE    MASE  RMSSE  ACF1
##   <chr>  <chr>    <dbl> <dbl> <dbl>    <dbl> <dbl>    <dbl> <dbl> <dbl>
## 1 NNET    Trai~ -6.15e-5 0.0261 0.0165 -0.00338 0.119 0.00487 0.00683 -0.555
## 2 Combina~ Trai~ -3.34e-2 0.289 0.234 -0.155 1.82 0.0689 0.0756 0.0559
## 3 Combina~ Trai~ -8.78e-3 0.481 0.385 -0.261 2.95 0.113 0.126 0.156
## 4 ARIMA4   Trai~ -2.04e-2 0.510 0.406 -0.0173 3.17 0.120 0.133 0.00799
## 5 Combina~ Trai~ -1.93e-2 0.518 0.413 -0.0805 3.21 0.122 0.136 0.0251
## 6 ARIMA2   Trai~ -1.75e-2 0.536 0.416 -0.138 3.23 0.122 0.140 0.0227
## 7 Combina~ Trai~ -2.59e-2 0.596 0.486 -0.428 3.69 0.143 0.156 0.429
## 8 Combina~ Trai~ -1.02e-2 0.598 0.484 -0.284 3.70 0.143 0.156 0.478
## 9 Combina~ Trai~ -8.74e-3 0.598 0.509 -0.359 3.94 0.150 0.156 0.426
## 10 Combina~ Trai~ -9.09e-3 0.605 0.508 -0.347 3.95 0.150 0.158 0.505
## # i 13 more rows
```

```
print(c("The best mdoel on Train is: ", acc2$.model[1]))
```

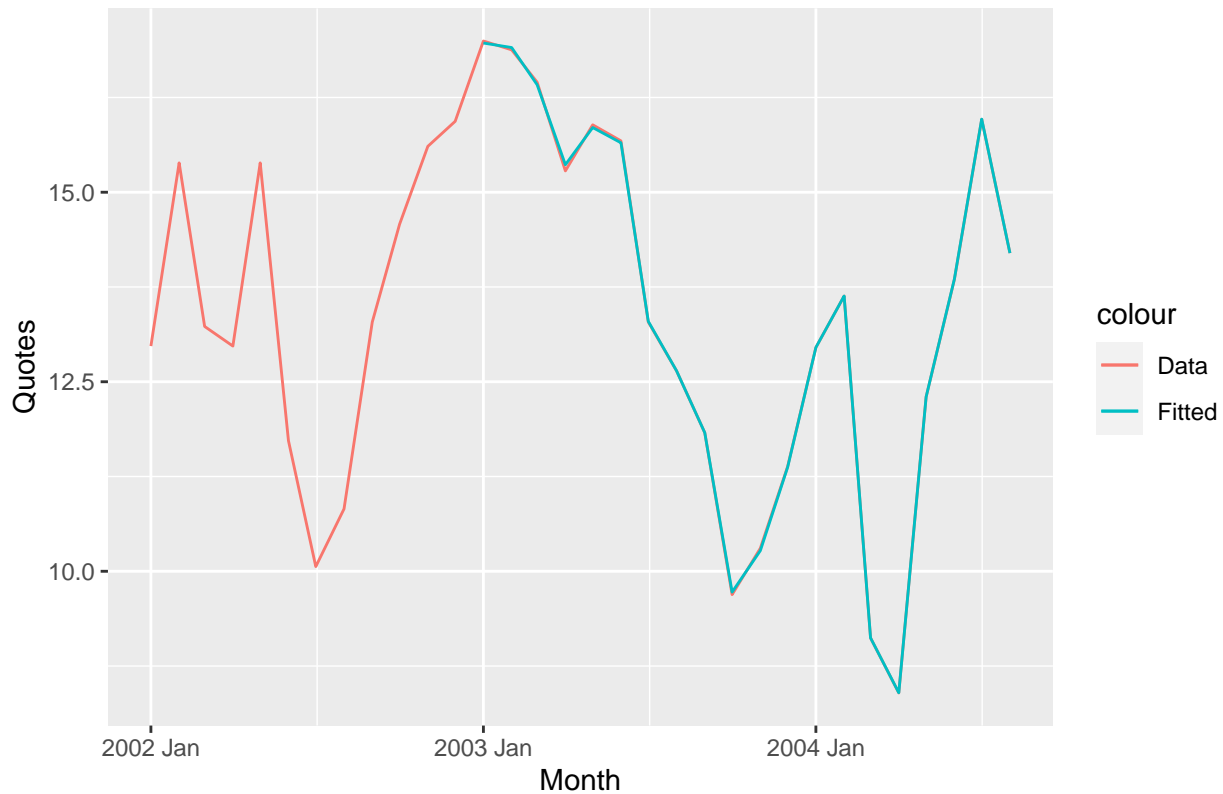
```
## [1] "The best mdoel on Train is: " "NNET"
```

```
fit2[,acc1$.model[1]] |> gg_tsresiduals()
```



```
augment(fit2) |> filter(.model == acc1$.model[1]) |>
  ggplot(aes(x = Month)) +
  geom_line(aes(y = Quotes , colour = "Data")) +
  geom_line(aes(y = .fitted, colour = "Fitted"))+
  ggtitle("The best model of training and actual data")
```

The best model of training and actual data



```
# augment(fit2) |> filter(.model == acc2$.model[1]) |>
#   features(.resid, ljung_box, lag=12, dof=0)
#
# augment(fit2) |> filter(.model == acc2$.model[1]) |>
#   features(.innov, box_pierce, dof = 0, lag = 12)
```

```
fc2 <- fit2 |> forecast(new_data = Test)
```

```
accT2 <- accuracy(fc2, Test) |> arrange(RMSE)
accT2
```

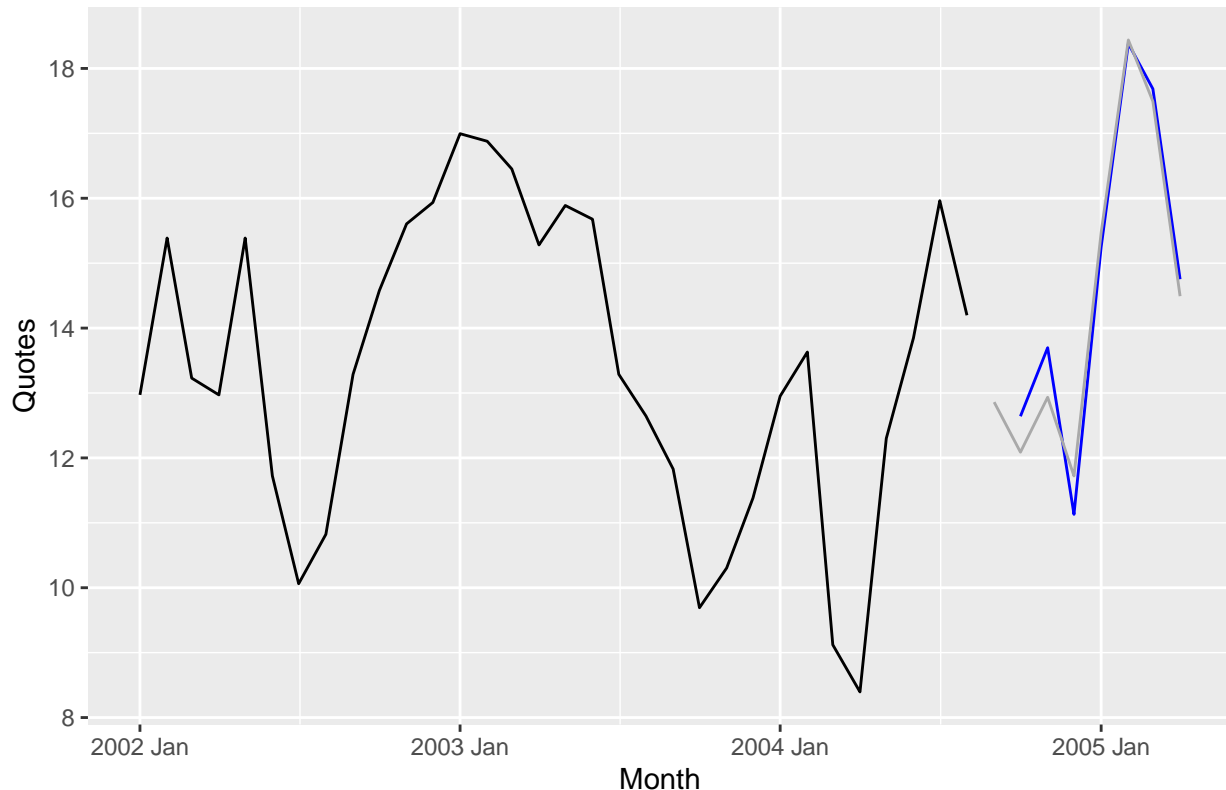
```
## # A tibble: 23 x 10
##   .model      .type      ME RMSE  MAE   MPE  MAPE  MASE RMSSE  ACF1
##   <chr>      <chr>    <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 Combination1 Test  -0.129 0.447 0.377 -0.943 2.88   NaN   NaN 0.0983
## 2 lm2        Test  -0.163 0.465 0.352 -1.25  2.78   NaN   NaN 0.104
## 3 lm4        Test  -0.153 0.475 0.435 -1.10  3.23   NaN   NaN 0.240
## 4 Combination3 Test   0.244 0.564 0.519  1.37  3.55   NaN   NaN 0.570
## 5 Combination4 Test   0.203 0.577 0.521  0.989 3.53   NaN   NaN 0.502
## 6 Combination2 Test   0.144 0.582 0.547  0.573 3.76   NaN   NaN 0.609
## 7 Combination5 Test   0.299 0.584 0.507  1.77  3.43   NaN   NaN 0.457
## 8 Combination6 Test   0.551 0.926 0.741  3.20  4.72   NaN   NaN 0.681
## 9 ARIMA2     Test   0.451 0.935 0.769  2.40  4.93   NaN   NaN 0.693
## 10 ARIMA4    Test   0.650 0.943 0.713  4.00  4.51   NaN   NaN 0.657
## # i 13 more rows
```

```
print(c("The best mdoel on Test is: ", accT2$.model[1]))
```

```
## [1] "The best mdoel on Test is: " "Combination1"
```

```
fc2 |> filter(.model == accT2$.model[1]) |>
  autoplot(Train, level = NULL) +
  labs(y = "Quotes", x = "Month",
  title = "8 months forecast for quotations") +
  autolayer(Test, colour = "darkgray")
```

8 months forecast for quotations



Cross Validation

```
insurance_stretch <- insurance |>
  stretch_tsibble(.init = 5, .step = 1) |>
  filter(.id != max(.id))

fit_cv1 <- insurance_stretch |>
  model(
    lm      = TSLM(Quotes),
    lm2     = TSLM(Quotes ~ TVadverts),
    lm3     = TSLM(Quotes ~ lag(TVadverts)),
    lm4     = TSLM(Quotes ~ TVadverts + lag(TVadverts)),
    Mean    = MEAN(Quotes),
    SNaive  = SNAIVE(Quotes),
    Naive   = NAIVE(Quotes),
    Drift   = RW(Quotes ~ drift()),
```

```

    ETS    = ETS(Quotes),
    NNET    = NNETAR(Quotes),
    Prophet = prophet(Quotes)
)

fc_cv1 <- fit_cv1 |>
  forecast(new_data = insurance_stretch , h=8)

# Cross-validated
fc_cv1 |> accuracy(insurance) |> arrange(RMSE)

## # A tibble: 11 x 10
##   .model .type      ME RMSE  MAE   MPE  MAPE  MASE RMSSE  ACF1
##   <chr>  <chr>    <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 Prophet Test  -1.92e- 5 0.752 0.383 -0.365 2.96   NaN   NaN 0.317
## 2 lm4     Test  -3.74e-16 0.813 0.684 -0.406 5.19   NaN   NaN 0.743
## 3 lm2     Test  -3.13e-16 0.833 0.714 -0.431 5.40   NaN   NaN 0.564
## 4 lm3     Test   4.20e-16 1.97  1.68 -2.39 13.2   NaN   NaN 0.372
## 5 Mean    Test   1.31e-16 2.19  1.83 -2.92 14.4   NaN   NaN 0.643
## 6 lm      Test   1.48e-16 2.19  1.83 -2.92 14.4   NaN   NaN 0.643
## 7 SNaive  Test   4.31e- 1 3.20  2.53  0.481 19.2   NaN   NaN 0.595
## 8 ETS     Test   1.63e- 2 3.32  2.63 -2.79 20.3   NaN   NaN 0.844
## 9 Naive   Test   1.53e- 3 3.36  2.68 -2.91 20.7   NaN   NaN 0.849
## 10 NNET   Test   1.04e+ 0 4.15  3.10  5.10 23.0   NaN   NaN 0.755
## 11 Drift  Test  -3.32e- 1 4.56  3.69 -5.30 28.4   NaN   NaN 0.920

# ARIMA = ARIMA(Quotes),
# ARIMA2 = ARIMA(Quotes~ TVadverts),
# ARIMA3 = ARIMA(Quotes~ lag(TVadverts)),
# ARIMA4 = ARIMA(Quotes~ TVadverts + lag(TVadverts)),

```

Bootstrapped

```

sim <- ins_stl |>
  generate(new_data = Train, times = 100,
    bootstrap_block_size = 24) |> select(-.model, -Quotes)

fit3 <- sim |> model(
  lm      = TSLM(.sim),
  lm2     = TSLM(.sim ~ TVadverts),
  lm3     = TSLM(.sim ~ lag(TVadverts)),
  lm4     = TSLM(.sim ~ TVadverts + lag(TVadverts)),
  ARIMA   = ARIMA(.sim),
  ARIMA2  = ARIMA(.sim ~ TVadverts),
  ARIMA3  = ARIMA(.sim ~ lag(TVadverts)),
  ARIMA4  = ARIMA(.sim ~ TVadverts + lag(TVadverts)),
  Mean    = MEAN(.sim),
  SNaive  = SNAIVE(.sim),
  Naive   = NAIVE(.sim),
  Drift   = RW(.sim ~ drift()),
  ETS     = ETS(.sim),

```

```

      NNET = NNETAR(.sim),
      Prophet = prophet(.sim)
)

acc3 <- accuracy(fit3) |> arrange(RMSE)

acc3

## # A tibble: 1,500 x 11
##   .rep .model .type      ME      RMSE      MAE      MPE      MAPE      MASE
##   <chr> <chr> <chr>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 40    Prophet Traini~ 3.11e-15 3.11e-13 2.45e-13 -3.71e-14 1.85e-12 1.07e-13
## 2 50    Prophet Traini~ -6.93e-14 3.35e-13 2.58e-13 -5.74e-13 1.99e-12 9.52e-14
## 3 82    Prophet Traini~ -4.00e-15 9.80e-13 7.94e-13 -6.23e-14 6.48e-12 2.71e-13
## 4 58    Prophet Traini~ 2.93e-14 1.09e-12 8.95e-13 6.09e-13 6.91e-12 3.74e-13
## 5 36    Prophet Traini~ -7.22e-13 3.90e-12 3.21e-12 -4.97e-12 2.43e-11 1.44e-12
## 6 14    Prophet Traini~ -1.05e-12 4.42e-12 3.46e-12 -7.57e-12 2.68e-11 1.07e-12
## 7 8      NNET    Traini~ -1.99e- 5 6.64e- 4 5.57e- 4 -2.00e- 4 4.45e- 3 3.29e- 4
## 8 47    NNET    Traini~ 9.75e- 7 9.70e- 4 7.70e- 4 -5.41e- 5 5.78e- 3 2.23e- 4
## 9 55    NNET    Traini~ 3.01e- 3 2.73e- 2 2.09e- 2 1.74e- 2 1.55e- 1 1.42e- 2
## 10 25   NNET    Traini~ 7.61e- 5 3.08e- 2 1.28e- 2 -1.38e- 3 8.71e- 2 3.23e- 3
## # i 1,490 more rows
## # i 2 more variables: RMSSE <dbl>, ACF1 <dbl>

fc3 <- fit3 |> filter(.rep == acc3$.rep[1]) |> select(-.rep) |> forecast(new_data = Test)

names(Test)[names(Test) == "Quotes"] <- ".sim"

accT3 <- accuracy(fc3, Test) |> arrange(RMSE)
accT3

## # A tibble: 15 x 10
##   .model .type      ME RMSE      MAE      MPE      MAPE      MASE RMSSE      ACF1
##   <chr> <chr>    <dbl> <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>    <dbl>
## 1 lm2    Test    0.695 1.61 1.28 3.38 8.18    NaN    NaN    0.632
## 2 lm4    Test    0.810 1.68 1.37 4.04 8.68    NaN    NaN    0.503
## 3 ARIMA2 Test    0.592 2.15 1.94 2.27 13.3    NaN    NaN    0.589
## 4 ARIMA4 Test    0.641 2.20 1.97 2.58 13.5    NaN    NaN    0.598
## 5 SNaive Test    1.04 2.22 1.45 5.71 9.06    NaN    NaN    0.524
## 6 lm3    Test    1.11 2.48 1.96 5.32 12.5    NaN    NaN    0.379
## 7 Mean   Test    1.18 2.63 2.04 5.84 12.9    NaN    NaN    0.568
## 8 lm     Test    1.18 2.63 2.04 5.84 12.9    NaN    NaN    0.568
## 9 ARIMA3 Test    2.67 3.10 2.67 18.0 18.0    NaN    NaN    0.431
## 10 ETS   Test    4.00 4.64 4.00 25.9 25.9    NaN    NaN    0.568
## 11 Naive Test    4.00 4.64 4.00 25.9 25.9    NaN    NaN    0.568
## 12 Drift Test    4.38 5.04 4.38 28.4 28.4    NaN    NaN    0.600
## 13 NNET  Test   -0.634 5.36 4.41 -9.33 32.1    NaN    NaN    0.550
## 14 Prophet Test  -0.286 6.39 4.28 -5.92 32.6    NaN    NaN   -0.170
## 15 ARIMA Test    6.49 6.83 6.49 44.3 44.3    NaN    NaN    0.537

print(c("The best mdoel on Test is: ", accT3$.model[1]))

## [1] "The best mdoel on Test is: " "lm2"

```