

# Project

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Install the COVID19 Data Hub R package. Select at least two countries to work with of interest to you and make forecasts of the number of cases for the next 12 months.

Build the following models:

Build a linear regression model using TSLM(). Build an appropriate exponential smoothing model.

```
library(pacman)
p_load(fpp3, tidyverse, doParallel)
```

```
covid <- read.csv("1.csv", header = T)
```

```
head(covid)
```

```
##           id           date confirmed deaths recovered tests vaccines
## 1 8320791a 2020-03-13         2      NA         NA      NA         NA
## 2 8320791a 2020-03-14         2      NA         NA      NA         NA
## 3 8320791a 2020-03-15         2      NA         NA      NA         NA
## 4 8320791a 2020-03-16         2      NA         NA      NA         NA
## 5 8320791a 2020-03-17         8      NA         NA      NA         NA
## 6 8320791a 2020-03-18         9      NA         NA      NA         NA
##  people_vaccinated people_fully_vaccinated hosp icu vent school_closing
## 1                NA                    NA    NA  NA  NA              NA
## 2                NA                    NA    NA  NA  NA              NA
## 3                NA                    NA    NA  NA  NA              NA
## 4                NA                    NA    NA  NA  NA              NA
## 5                NA                    NA    NA  NA  NA              NA
## 6                NA                    NA    NA  NA  NA              NA
##  workplace_closing cancel_events gatherings_restrictions transport_closing
## 1                NA                NA                    NA              NA
## 2                NA                NA                    NA              NA
## 3                NA                NA                    NA              NA
## 4                NA                NA                    NA              NA
## 5                NA                NA                    NA              NA
## 6                NA                NA                    NA              NA
##  stay_home_restrictions internal_movement_restrictions
## 1                NA                    NA
## 2                NA                    NA
## 3                NA                    NA
## 4                NA                    NA
## 5                NA                    NA
## 6                NA                    NA
##  international_movement_restrictions information_campaigns testing_policy
## 1                                NA                                NA              NA
```

```

## 2          NA          NA          NA
## 3          NA          NA          NA
## 4          NA          NA          NA
## 5          NA          NA          NA
## 6          NA          NA          NA
##   contact_tracing facial_coverings vaccination_policy elderly_people_protection
## 1          NA          NA          NA          NA
## 2          NA          NA          NA          NA
## 3          NA          NA          NA          NA
## 4          NA          NA          NA          NA
## 5          NA          NA          NA          NA
## 6          NA          NA          NA          NA
##   government_response_index stringency_index containment_health_index
## 1          NA          NA          NA
## 2          NA          NA          NA
## 3          NA          NA          NA
## 4          NA          NA          NA
## 5          NA          NA          NA
## 6          NA          NA          NA
##   economic_support_index administrative_area_level administrative_area_level_1
## 1          NA          1          Grand Princess
## 2          NA          1          Grand Princess
## 3          NA          1          Grand Princess
## 4          NA          1          Grand Princess
## 5          NA          1          Grand Princess
## 6          NA          1          Grand Princess
##   administrative_area_level_2 administrative_area_level_3 latitude longitude
## 1          NA          NA          NA          NA
## 2          NA          NA          NA          NA
## 3          NA          NA          NA          NA
## 4          NA          NA          NA          NA
## 5          NA          NA          NA          NA
## 6          NA          NA          NA          NA
##   population iso_alpha_3 iso_alpha_2 iso_numeric iso_currency key_local
## 1      3533          NA          NA          NA          NA
## 2      3533          NA          NA          NA          NA
## 3      3533          NA          NA          NA          NA
## 4      3533          NA          NA          NA          NA
## 5      3533          NA          NA          NA          NA
## 6      3533          NA          NA          NA          NA
##   key_google_mobility key_apple_mobility key_jhu_csse key_nuts key_gadm
## 1          NA
## 2          NA
## 3          NA
## 4          NA
## 5          NA
## 6          NA

```

```

covid$id[covid$id == "59a13ceb"] <- "United States"
covid$id[covid$id == "78833522"] <- "China"

```

```

covid <- covid %>%
  filter(id == "United States" | id == "China")
head(covid)

```

##	id	date	confirmed	deaths	recovered	tests	vaccines	people_vaccinated
## 1	China	2020-01-03	0	0	NA	NA	NA	NA
## 2	China	2020-01-04	1	0	NA	NA	NA	NA
## 3	China	2020-01-05	1	0	NA	NA	NA	NA
## 4	China	2020-01-06	4	0	NA	NA	NA	NA
## 5	China	2020-01-07	4	0	NA	NA	NA	NA
## 6	China	2020-01-08	4	0	NA	NA	NA	NA
##	people_fully_vaccinated	hosp	icu	vent	school_closing	workplace_closing		
## 1	NA	NA	NA	NA	0	0		
## 2	NA	NA	NA	NA	0	0		
## 3	NA	NA	NA	NA	0	0		
## 4	NA	NA	NA	NA	0	0		
## 5	NA	NA	NA	NA	0	0		
## 6	NA	NA	NA	NA	0	0		
##	cancel_events	gatherings_restrictions	transport_closing					
## 1	0		0		0			
## 2	0		0		0			
## 3	0		0		0			
## 4	0		0		0			
## 5	0		0		0			
## 6	0		0		0			
##	stay_home_restrictions	internal_movement_restrictions						
## 1	0				0			
## 2	0				0			
## 3	0				0			
## 4	0				0			
## 5	0				0			
## 6	0				0			
##	international_movement_restrictions	information_campaigns	testing_policy					
## 1		0	0		1			
## 2		0	0		1			
## 3		0	-1		1			
## 4		0	-1		1			
## 5		0	-1		1			
## 6		0	-1		1			
##	contact_tracing	facial_coverings	vaccination_policy	elderly_people_protection				
## 1	1	-1	0		0			
## 2	1	-1	0		0			
## 3	2	-1	0		0			
## 4	2	-1	0		0			
## 5	2	-1	0		0			
## 6	2	-1	0		0			
##	government_response_index	stringency_index	containment_health_index					
## 1	5.99		0.00		6.85			
## 2	5.99		0.00		6.85			
## 3	10.68		2.78		12.20			
## 4	10.68		2.78		12.20			
## 5	10.68		2.78		12.20			
## 6	10.68		2.78		12.20			
##	economic_support_index	administrative_area_level	administrative_area_level_1					
## 1	0		1		China			
## 2	0		1		China			
## 3	0		1		China			
## 4	0		1		China			

```
## 5          0          1          China
## 6          0          1          China
##   administrative_area_level_2 administrative_area_level_3 latitude longitude
## 1          NA          NA 32.82838 111.6491
## 2          NA          NA 32.82838 111.6491
## 3          NA          NA 32.82838 111.6491
## 4          NA          NA 32.82838 111.6491
## 5          NA          NA 32.82838 111.6491
## 6          NA          NA 32.82838 111.6491
##   population iso_alpha_3 iso_alpha_2 iso_numeric iso_currency key_local
## 1 1392730000     CHN      CN        156        CNY      NA
## 2 1392730000     CHN      CN        156        CNY      NA
## 3 1392730000     CHN      CN        156        CNY      NA
## 4 1392730000     CHN      CN        156        CNY      NA
## 5 1392730000     CHN      CN        156        CNY      NA
## 6 1392730000     CHN      CN        156        CNY      NA
##   key_google_mobility key_apple_mobility key_jhu_csse key_nuts key_gadm
## 1                      CN          NA      CHN
## 2                      CN          NA      CHN
## 3                      CN          NA      CHN
## 4                      CN          NA      CHN
## 5                      CN          NA      CHN
## 6                      CN          NA      CHN
```

```
covid$date <- as.Date(covid$date)
```

```
covid <- covid %>%
  as_tsibble(key = "id", index = "date")
head(covid)
```

```
## # A tsibble: 6 x 47 [1D]
## # Key:      id [1]
##   id   date      confirmed deaths recover~1 tests vacci~2 peopl~3 peopl~4 hosp
##   <chr> <date>      <int>  <int>    <int> <dbl>    <dbl>    <int>  <int> <int>
## 1 China 2020-01-03      0      0      NA    NA      NA      NA      NA    NA
## 2 China 2020-01-04      1      0      NA    NA      NA      NA      NA    NA
## 3 China 2020-01-05      1      0      NA    NA      NA      NA      NA    NA
## 4 China 2020-01-06      4      0      NA    NA      NA      NA      NA    NA
## 5 China 2020-01-07      4      0      NA    NA      NA      NA      NA    NA
## 6 China 2020-01-08      4      0      NA    NA      NA      NA      NA    NA
## # ... with 37 more variables: icu <int>, vent <int>, school_closing <int>,
## #   workplace_closing <int>, cancel_events <int>,
## #   gatherings_restrictions <int>, transport_closing <int>,
## #   stay_home_restrictions <int>, internal_movement_restrictions <int>,
## #   international_movement_restrictions <int>, information_campaigns <int>,
## #   testing_policy <int>, contact_tracing <int>, facial_coverings <int>,
## #   vaccination_policy <int>, elderly_people_protection <int>, ...
```

```
class(covid)
```

```
## [1] "tbl_ts"      "tbl_df"      "tbl"         "data.frame"
```

```
covid <- covid %>%
  dplyr::select(id, date, confirmed)
```

```
covid <- covid %>%
  drop_na()
```

## Time Series Linear Model

### China

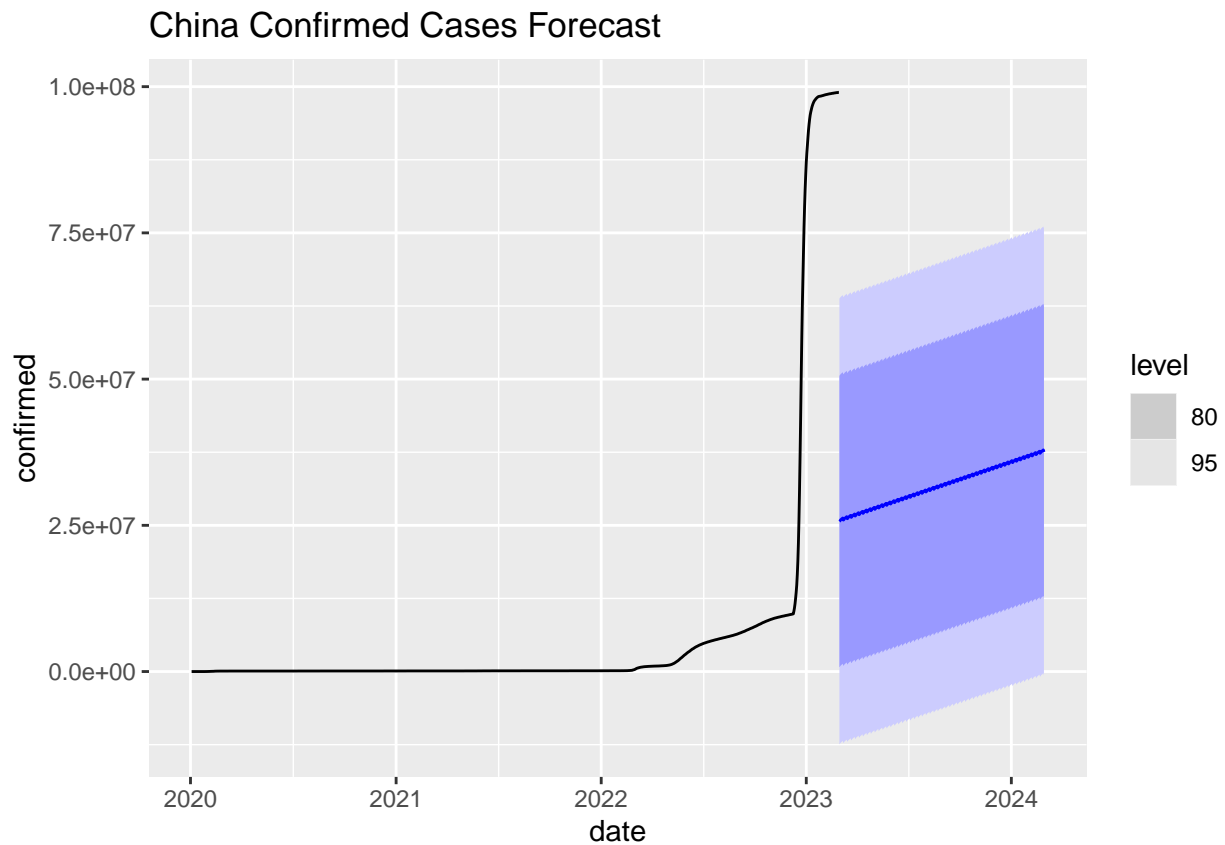
```
china <- covid %>%
  filter(id == "China") %>%
  select(date, confirmed)

head(china)
```

```
## # A tibble: 6 x 2 [1D]
##   date      confirmed
##   <date>      <int>
## 1 2020-01-03         0
## 2 2020-01-04         1
## 3 2020-01-05         1
## 4 2020-01-06         4
## 5 2020-01-07         4
## 6 2020-01-08         4
```

```
china_fit <- china %>%
  model(TSLM(confirmed ~ trend() + season()))

china_fit %>%
  forecast(h = 365) %>%
  autoplot(china) +
  labs(title = "China Confirmed Cases Forecast")
```



## United States

```
us <- covid %>%
  filter(id == "United States") %>%
  select(date, confirmed)

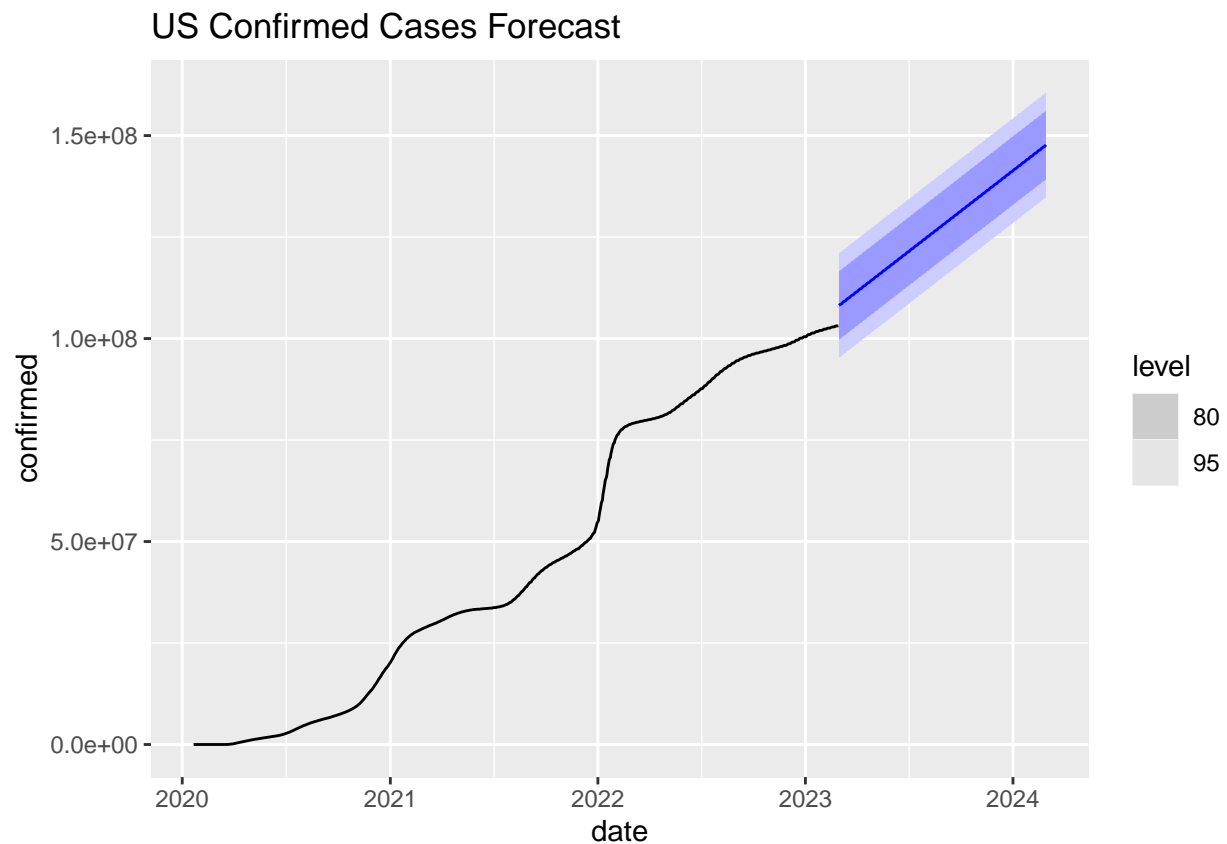
head(us)
```

```
## # A tsibble: 6 x 2 [1D]
##   date      confirmed
##   <date>      <int>
## 1 2020-01-21         1
## 2 2020-01-22         1
## 3 2020-01-23         1
## 4 2020-01-24         2
## 5 2020-01-25         3
## 6 2020-01-26         5
```

```
us_fit <- us %>%
  model(TSLM(confirmed ~ trend() + season()))

us_fit %>%
  forecast(h = 365) %>%
```

```
autoplot(us) +
  labs(title = "US Confirmed Cases Forecast")
```



## Exponential Smoothing

### China

#### Best Model

According to the RSME value, the appropriate exponential model is the holts method.

```
registerDoParallel(cores = 16)
china %>%
  stretch_tsibble() %>%
  model(dholts = ETS(confirmed ~ error("A") + trend("Ad") + season("N")),
        holts = ETS(confirmed ~ error("A") + trend("A") + season("N")),
        ses = ETS(confirmed ~ error("A") + trend("N") + season("N"))) %>%
  forecast(h = 365) %>%
  accuracy(china)
```

```
## Warning: 6 errors (2 unique) encountered for dholts
## [5] Not enough data to estimate this ETS model.
## [1] only 1 case, but 2 variables
```

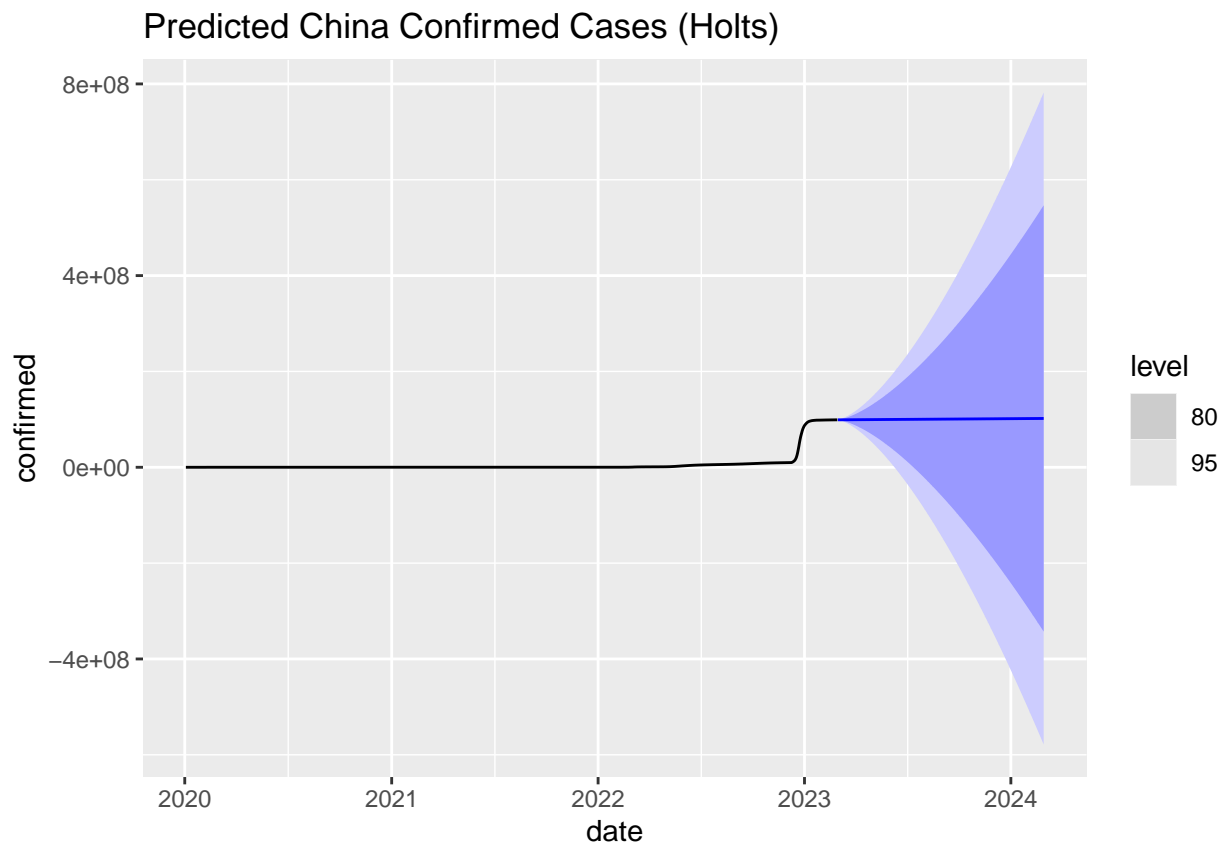
```
## Warning: 5 errors (2 unique) encountered for holts
## [4] Not enough data to estimate this ETS model.
## [1] only 1 case, but 2 variables

## Warning: 3 errors (1 unique) encountered for ses
## [3] Not enough data to estimate this ETS model.

## Warning: The future dataset is incomplete, incomplete out-of-sample data will be treated as missing.
## 365 observations are missing between 2023-03-01 and 2024-02-28

## # A tibble: 3 x 10
##   .model .type      ME      RMSE      MAE    MPE    MAPE    MASE  RMSSE  ACF1
##   <chr>  <chr>    <dbl>    <dbl>    <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 dholts Test  6433955. 22671937. 6849061. 32.8   38.3   11.3   5.94 0.991
## 2 holts  Test  5690838. 22618077. 6637613.  9.96  49.3   11.0   5.92 0.990
## 3 ses    Test  7000446. 22793917. 7000446. 40.3   40.3   11.6   5.97 0.992
```

```
china %>%
  model(holts = ETS(confirmed ~ error("A") + trend("A") + season("N"))) %>%
  forecast(h = 365) %>%
  autoplot(china) +
  labs(title = "Predicted China Confirmed Cases (Holts)")
```





## US

### Best Model

According to the RSME values, the appropriate exponential model is the damped holts method.

```
us %>%
  stretch_tsibble(.init = 10) %>%
  model(dholts = ETS(confirmed ~ error("A") + trend("Ad") + season("N")),
        holts = ETS(confirmed ~ error("A") + trend("A") + season("N")),
        ses = ETS(confirmed ~ error("A") + trend("N") + season("N"))) %>%
  forecast(h = 365) %>%
  accuracy(us)
```

```
## Warning: The future dataset is incomplete, incomplete out-of-sample data will be treated as missing.
## 365 observations are missing between 2023-03-01 and 2024-02-28
```

```
## # A tibble: 3 x 10
##   .model .type      ME      RMSE      MAE    MPE  MAPE  MASE RMSSE  ACF1
##   <chr>  <chr>    <dbl>    <dbl>    <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 dholts Test  14214508. 20253535. 14686189.  34.4  35.0  23.0  19.8 0.993
## 2 holts  Test   1131641. 26694216. 13556262.  10.8  29.7  21.2  26.1 0.997
## 3 ses    Test   18160490. 23345197. 18160490.  43.2  43.2  28.4  22.9 0.991
```

```
us %>%
  model(dholts = ETS(confirmed ~ error("A") + trend("Ad") + season("N"))) %>%
  forecast(h = 365) %>%
  autoplot(us) +
  labs(title = "Predicted US Confirmed Cases (Damped Holts)")
```

Predicted US Confirmed Cases (Damped Holts)

