

# Analisis Overfitting sin IC

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## Paquetes

```
library(keras) # for deep learning  
library(tidyverse) # general utility functions
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.6      v purrr   0.3.4  
## v tibble  3.1.6      v dplyr  1.0.9  
## v tidyr   1.2.0      v stringr 1.4.0  
## v readr   2.1.2      v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()
```

```
library(caret) # machine learning utility functions
```

```
## Loading required package: lattice
```

```
##
```

```
## Attaching package: 'caret'
```

```
## The following object is masked from 'package:purrr':
```

```
##
```

```
## lift
```

```
library(tibble)  
library(readr)  
library(ggplot2)  
library(tensorflow)
```

```
##
```

```
## Attaching package: 'tensorflow'
```

```
## The following object is masked from 'package:caret':
```

```
##
```

```
## train
```

```
library(neuralnet)
```

```
##  
## Attaching package: 'neuralnet'  
  
## The following object is masked from 'package:dplyr':  
##  
##      compute
```

## Datos

```
load("C:/Users/usuario1/Desktop/CIMPA/Github_CIMPA/PRACTICA_CIMPA/base_cantones.RData")
```

```
Alajuela <- basecantons %>% filter(Canton == "Alajuela") %>%
```

```
  dplyr::select(Year,Month,Nino12SSTA, Nino3SSTA, Nino4SSTA,Nino34SSTA,Nino34SSTA1, Nino34SSTA2, Nino34SSTA3)
```

```
  arrange(Year,Month) %>% ungroup() %>% mutate(Month=as.numeric(Month))
```

```
if(anyNA(Alajuela)){  
  Alajuela <- na.omit(Alajuela)  
}
```

```
#Escala
```

```
normalize <- function(x) {  
  return ((x - min(x)) / (max(x) - min(x)))  
}
```

```
max <- apply(Alajuela,2,max)  
min <- apply(Alajuela,2,min)
```

```
Alajuela2 <- apply(Alajuela, 2, normalize)
```

```
#Train y test
```

```
Fechas = c(1, 0.95, 0.90, 0.85, 0.80, 0.75, 0.70, 0.65, 0.60, 0.55, 0.50)  
Eval = NULL  
Evalc = NULL  
Eval3 = NULL
```

```
p1 = list()  
p2 = list()  
p3 = list()
```

```
for (i in 1:length(Fechas)) {
```

```

data_train = as.data.frame(Alajuela2) %>% filter(Year < Fechas[i]) #PARA ENTRENAR HASTA 2018
data_test = as.data.frame(Alajuela2) %>% filter(Year >= Fechas[i])

X_train = as.matrix(data_train[, -ncol(data_train)])
y_train = as.matrix(data_train[, ncol(data_train)])

X_test = as.matrix(data_test[, -ncol(data_test)])
y_test = as.matrix(data_test[, ncol(data_test)])

## Modelo

set.seed(123)
model <- keras_model_sequential()

model %>%
  layer_simple_rnn(units = 100, input_shape = c(ncol(X_train), 1), activation='tanh',
                  kernel_initializer= initializer_constant(0.5),
                  bias_initializer=initializer_zeros()) %>%
  layer_dense(units = 50, activation = "relu") %>%
  layer_dense(units = 50, activation = "relu") %>%
  layer_dense(units = 50, activation = "relu") %>%
  layer_dropout(rate = 0.1) %>%
  layer_dense(units = 25, activation = "relu") %>%
  layer_dense(units = 25, activation = "relu") %>%
  layer_dense(units = 25, activation = "relu") %>%
  layer_dropout(rate = 0.1) %>%
  layer_dense(units = 12, activation = "relu") %>%
  layer_dense(units = 12, activation = "relu") %>%
  layer_dropout(rate = 0.1) %>%
  layer_dense(units = 6, activation = "relu") %>%
  layer_dense(units = 6, activation = "relu") %>%
  layer_dense(units = 1, activation = "sigmoid")

## Entrenar al modelo

model %>% compile(
  optimizer = "adam",
  loss = "mse",
  metrics = "mae")

history <- model %>% fit(
  X_train,
  y_train,
  epochs = 100,
  batch_size = 18,
  validation_split = 0.1,
  shuffle = F
)

denorm <- function(x) {
  return (x*(max(Alajuela$RR) - min(Alajuela$RR))+min(Alajuela$RR))
}

```

```

pred = denorm(model %>% predict(Alajuela2[, -33]))
results = denorm(model %>% predict(X_test))
results3 = denorm(model %>% predict(Alajuela2[233:235, -33]))

#Grafico

data1 = as.data.frame(cbind(pred, Alajuela$RR))
names(data1) = c("fit", "RR")
data2 = as.data.frame(cbind(results, Alajuela$RR[(236-length(results)):235]))
names(data2) = c("fit", "RR")
data3 = as.data.frame(cbind(results3, Alajuela$RR[233:235]))
names(data3) = c("fit", "RR")

Fecha = paste(Alajuela$Year, Alajuela$Month)

everyother1 <- function(x) x[(seq_along(Fecha) + 5)%%12 == 6]

p1[[i]] <- ggplot(data1, aes(x = Fecha, y = RR, group = 1)) + geom_line(colour = "blue") +
  geom_line(aes(x = Fecha, y = fit, colour = "red")) +
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
        panel.background = element_blank(), axis.text.x = element_text(angle = 45), legend.position = "none") +
  scale_x_discrete(breaks = everyother1) + labs(x = "Fecha", y = "Riesgo Relativo") +
  ggtitle(paste("Predicción desde", Fechas[i], sep = ": "))

p3[[i]] <- ggplot(data3, aes(x = Fecha[233:235], y = RR, group = 1)) + geom_line(colour = "blue") +
  geom_line(aes(x = Fecha[233:235], y = fit, colour = "red")) +
  theme(panel.grid.major = element_blank(), panel.grid.minor = element_blank(),
        panel.background = element_blank(), axis.text.x = element_text(angle = 45), legend.position = "none") +
  labs(x = "Fecha", y = "Riesgo Relativo") + ggtitle(paste("Predicción 3 meses training hasta", Fechas[i], sep = ": "))

metricas <- function(tabla){
  NRMSE <- mean((tabla$fit-tabla$RR)^2)/mean(tabla$RR)
  return(data.frame(NRMSE))
}

Eval[i] = as.numeric(metricas(data1))

Evalc[i] = as.numeric(metricas(data2))

Eval3[[i]] = as.numeric(metricas(data3))

k_clear_session()
}

## Loaded Tensorflow version 2.8.0

```

## Resultados

```
NRMSE = cbind (as.numeric(Eval), as.numeric(Evalc), as.numeric(Eval3))
colnames(NRMSE) = c("Total", "Test", "Solo 2021")
rownames(NRMSE) = c("2021", "2020 +", "2019 +", "2018 +", "2017+", "2016+", "2015+", "2014+", "2013+",
as.data.frame(NRMSE)
```

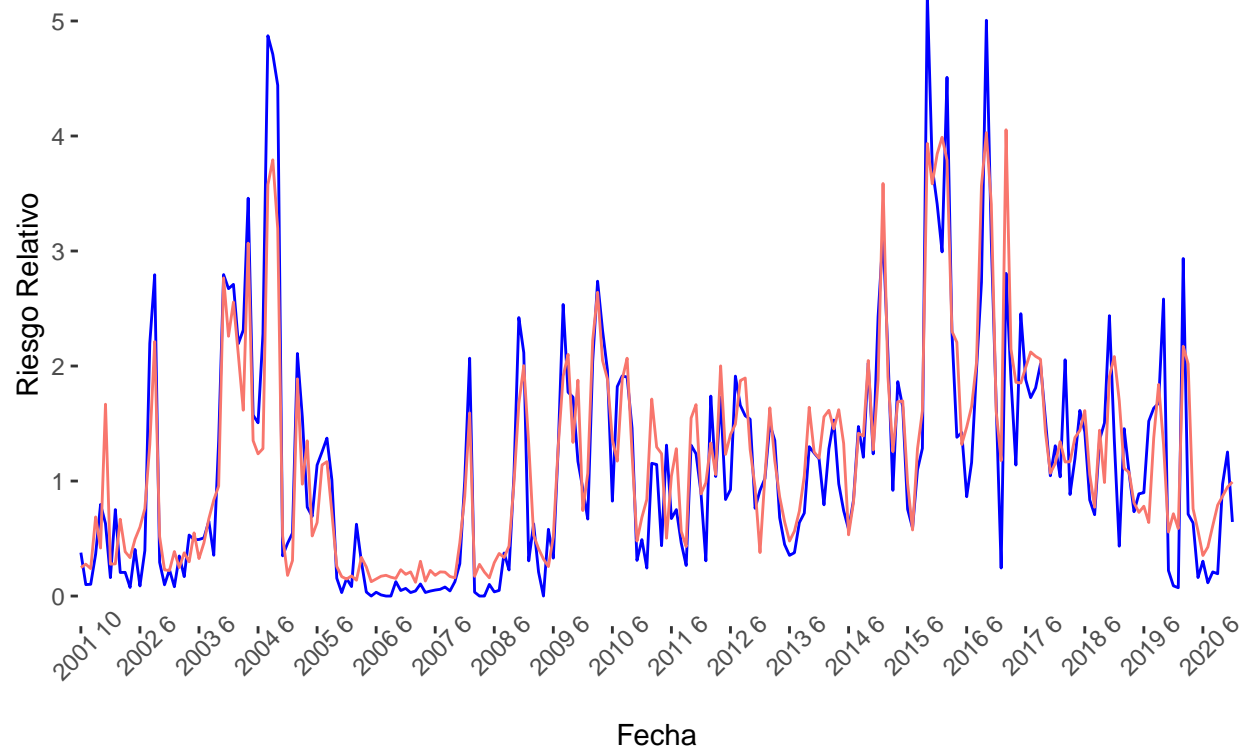
```
##           Total           Test  Solo 2021
## 2021    0.1680727 0.07825733 0.07825733
## 2020 + 0.3600905 0.99871594 0.29005198
## 2019 + 0.3246507 0.59611369 0.18247143
## 2018 + 0.2253654 0.50965104 0.11820007
## 2017+ 0.4836737 0.63771218 0.31077030
## 2016+ 0.3100379 0.56918145 0.13058281
## 2015+ 0.3461211 0.48041430 0.27136931
## 2014+ 0.3533642 0.47620514 0.28888497
## 2013+ 0.4811523 0.56548959 0.27991850
## 2012+ 0.4815832 0.67019586 0.64062580
## 2011+ 0.6134728 0.78794745 0.58607954
```

## Gráficos

p1

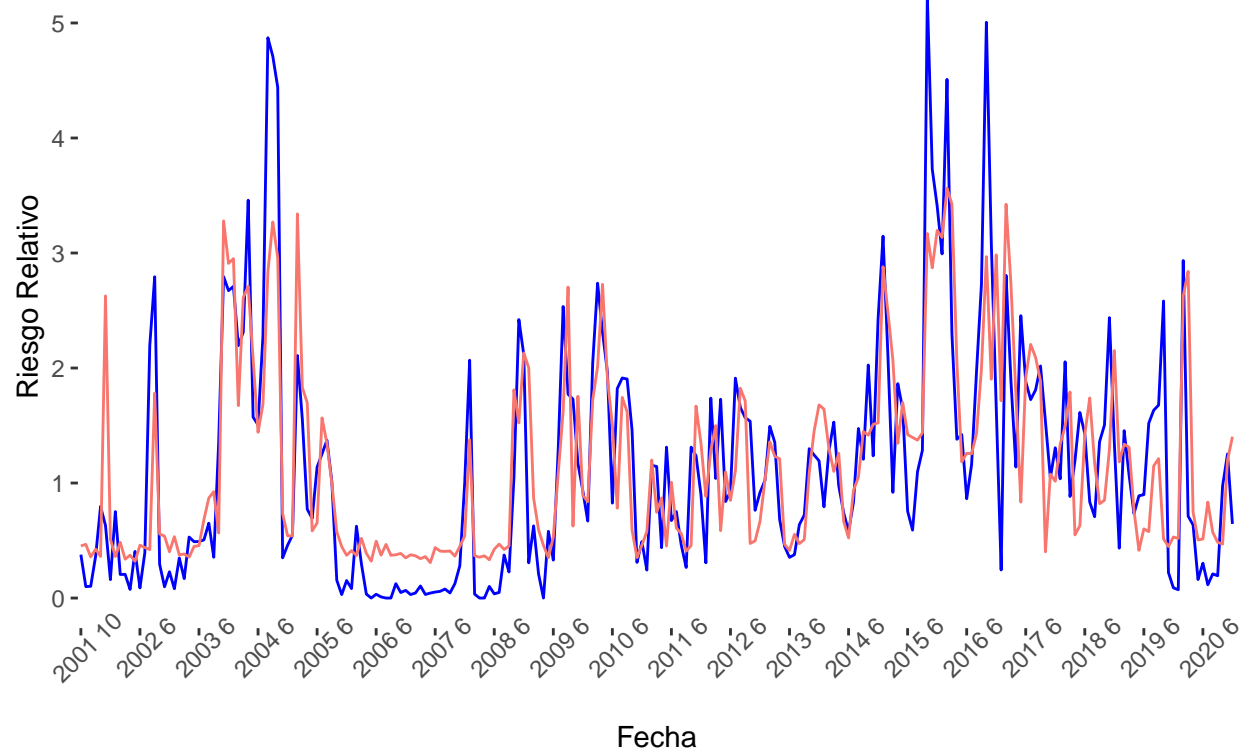
```
## [[1]]
```

Predicción desde: 1



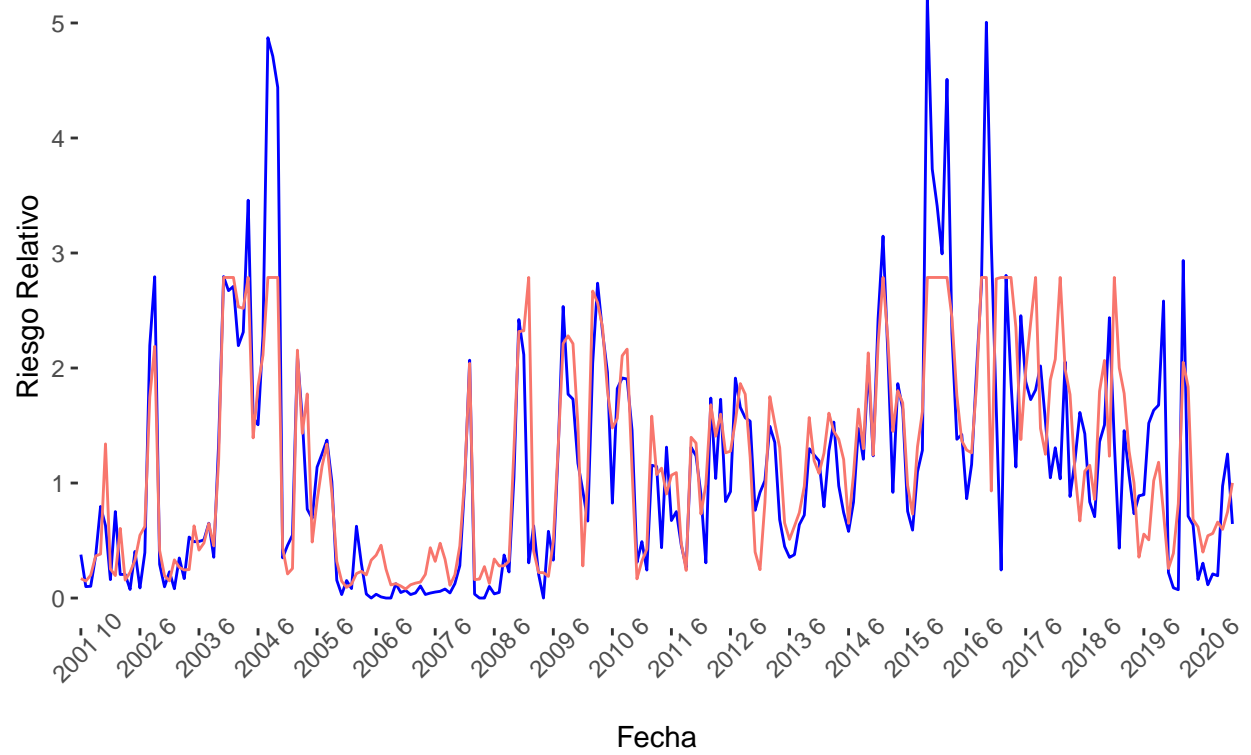
```
##  
## [[2]]
```

Predicción desde: 0.95



```
##  
## [[3]]
```

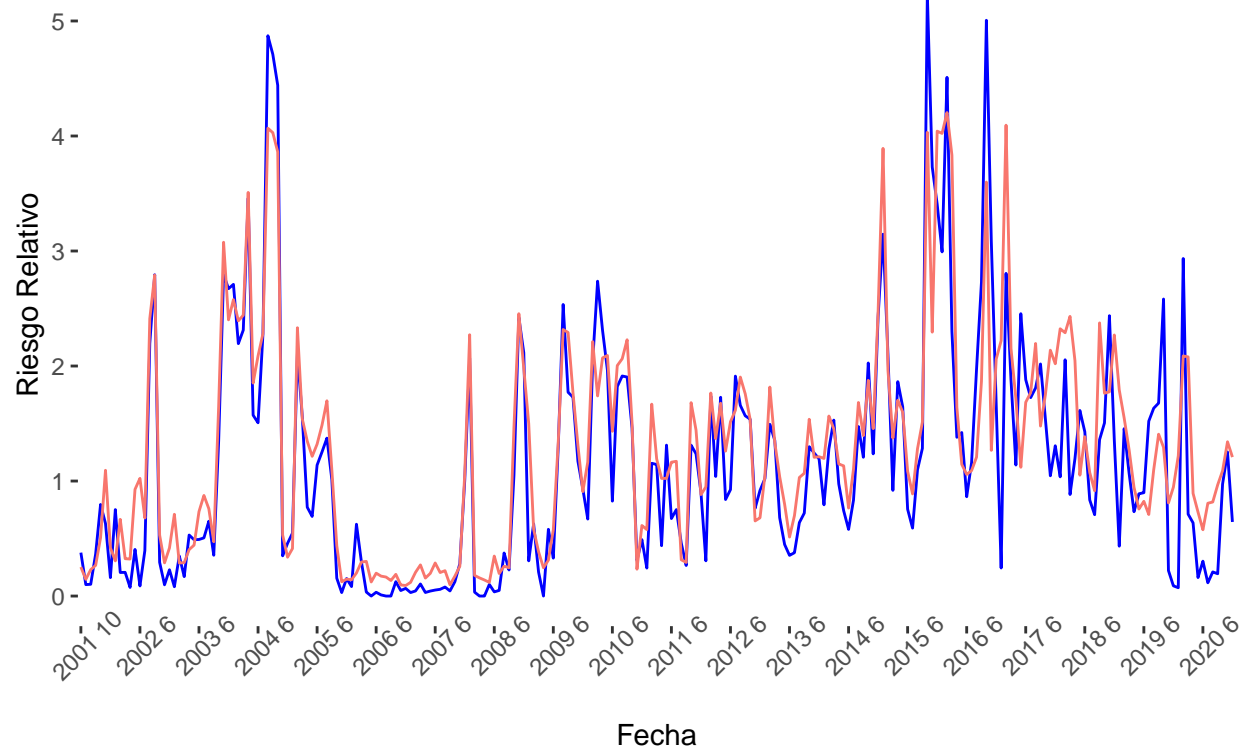
Predicción desde: 0.9



```
##  
## [[4]]
```

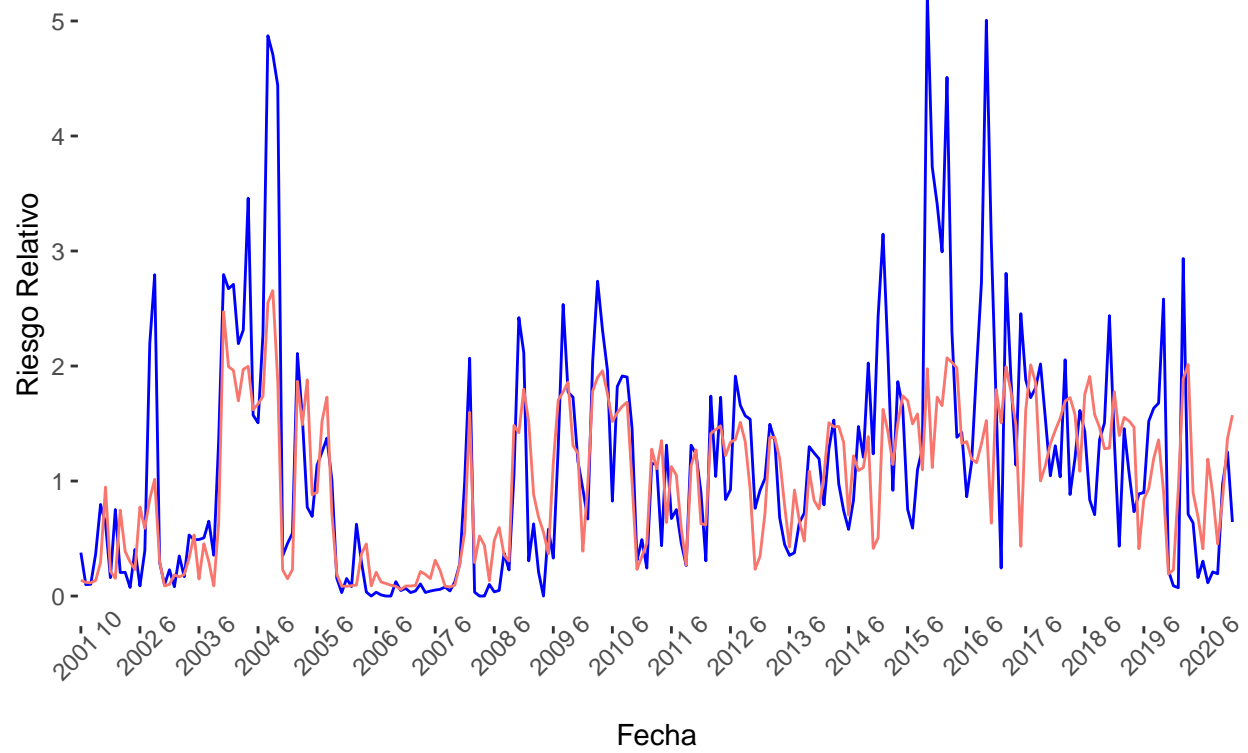


Predicción desde: 0.85



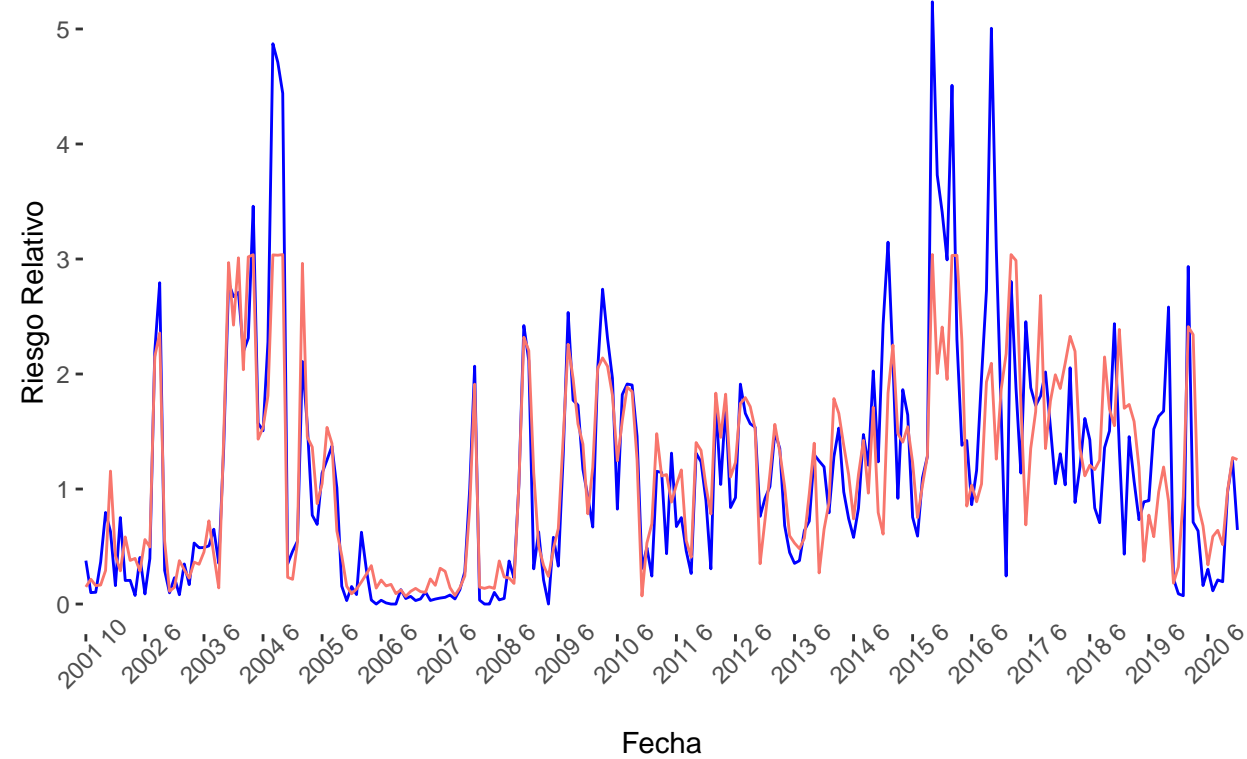
```
##  
## [[5]]
```

Predicción desde: 0.8



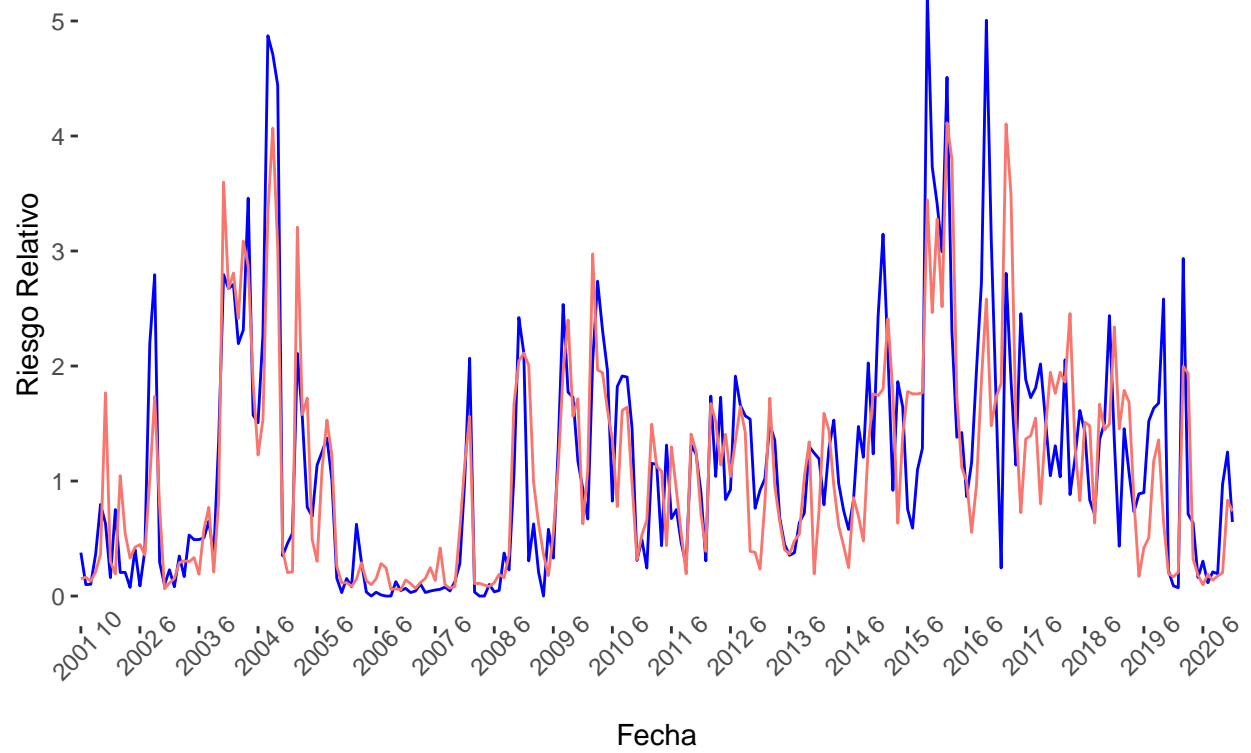
```
##  
## [[6]]
```

Predicción desde: 0.75



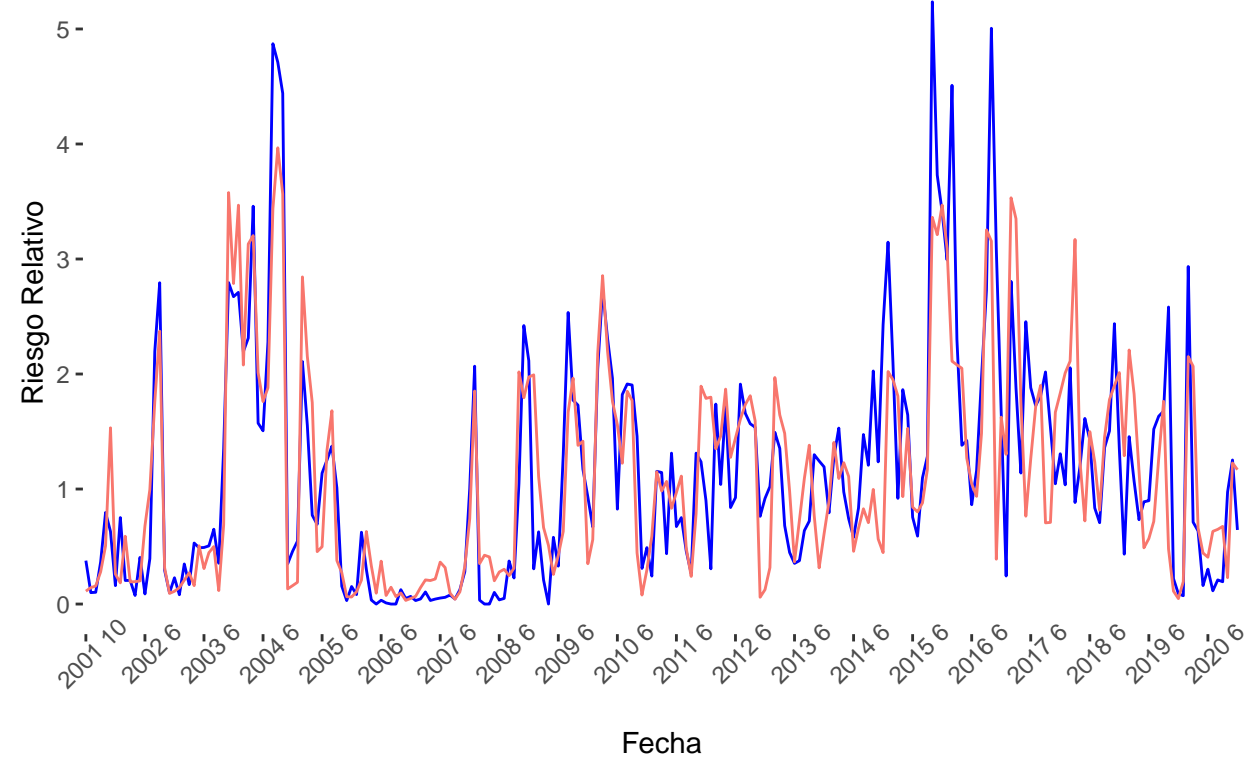
```
##  
## [[7]]
```

Predicción desde: 0.7



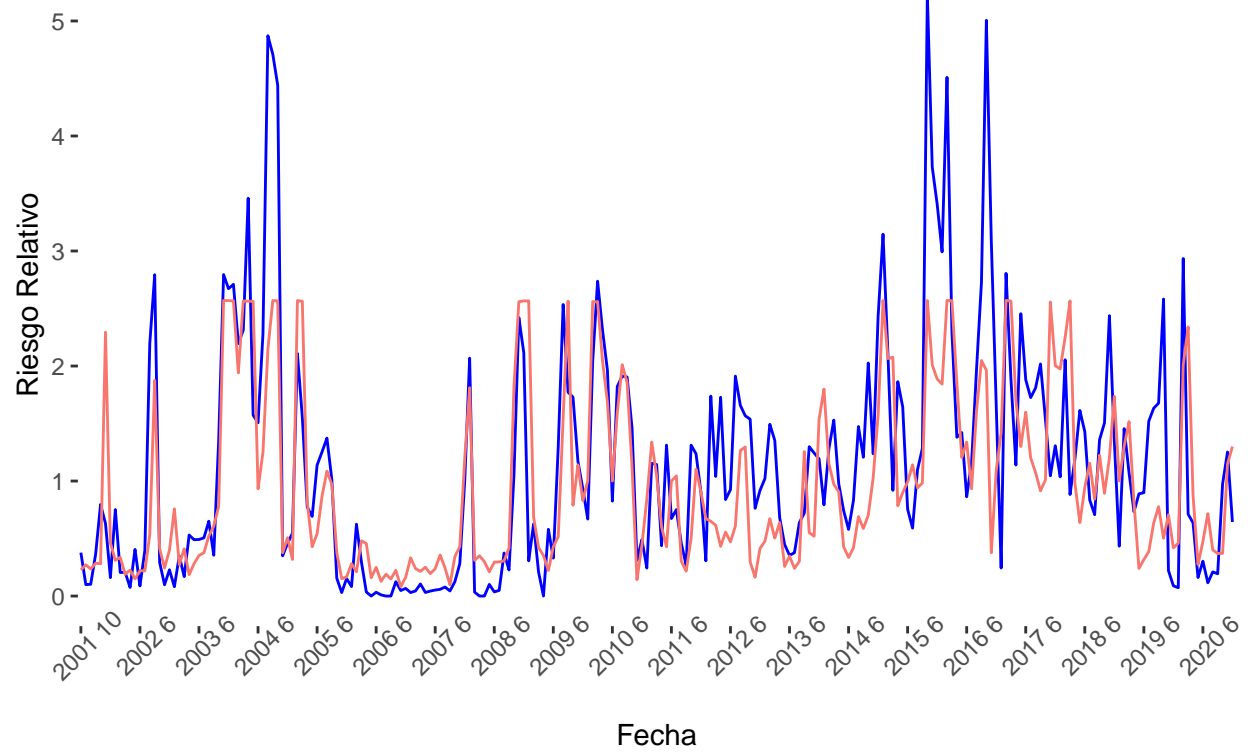
```
##  
## [[8]]
```

Predicción desde: 0.65



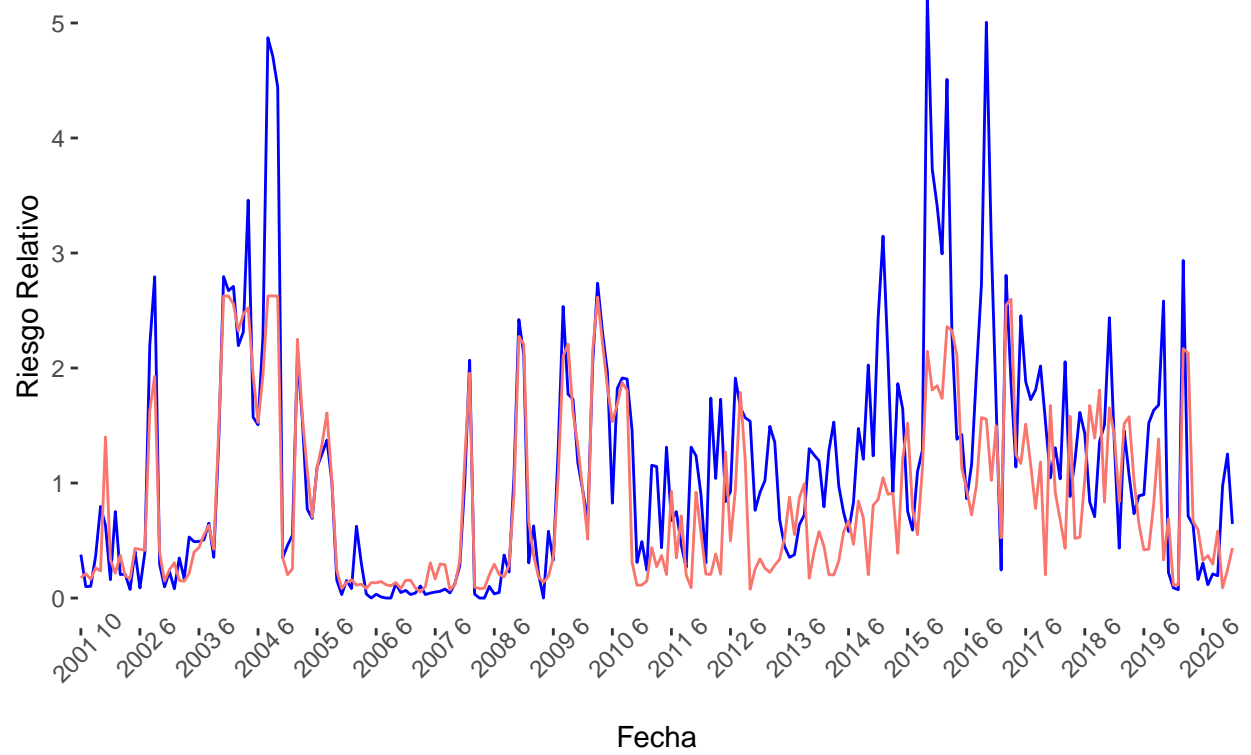
```
##  
## [[9]]
```

Predicción desde: 0.6



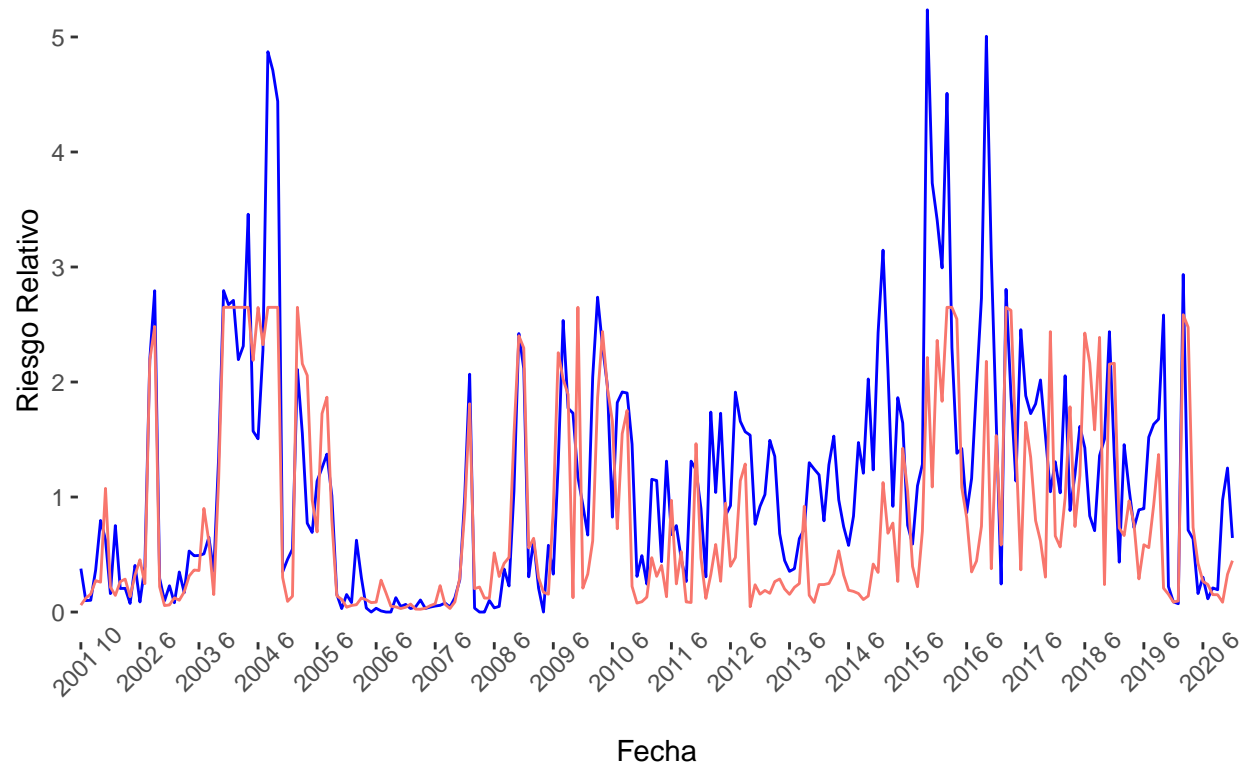
```
##  
## [[10]]
```

Predicción desde: 0.55



```
##  
## [[11]]
```

Predicción desde: 0.5

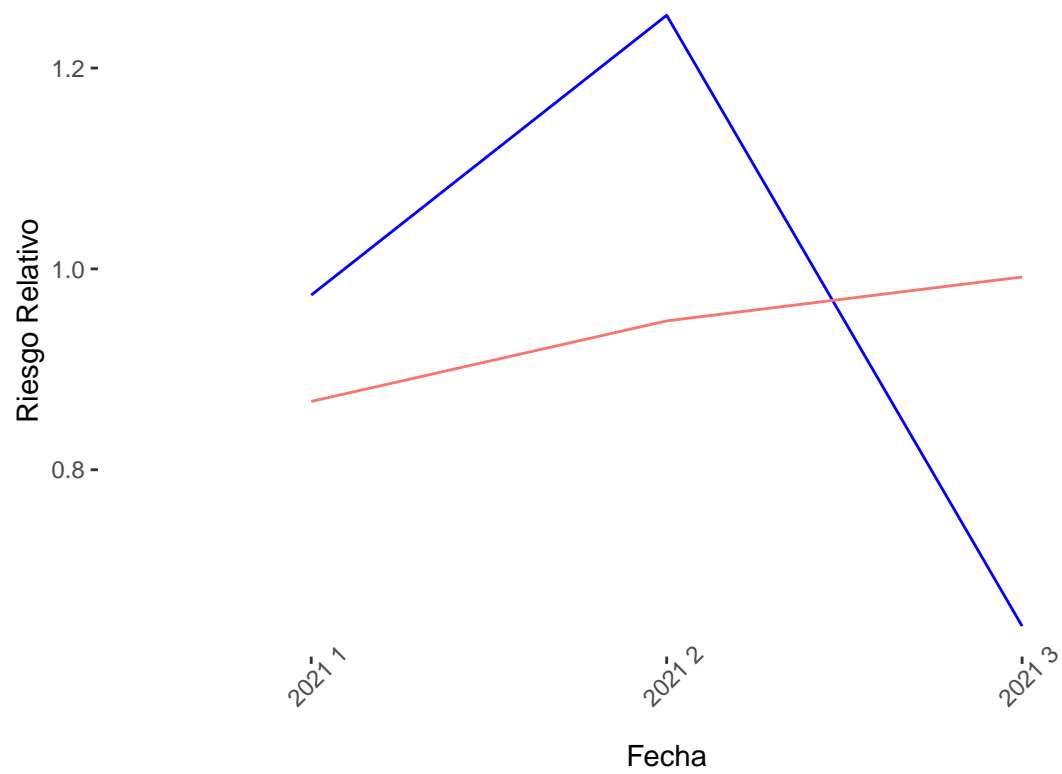


p3

## [[1]]

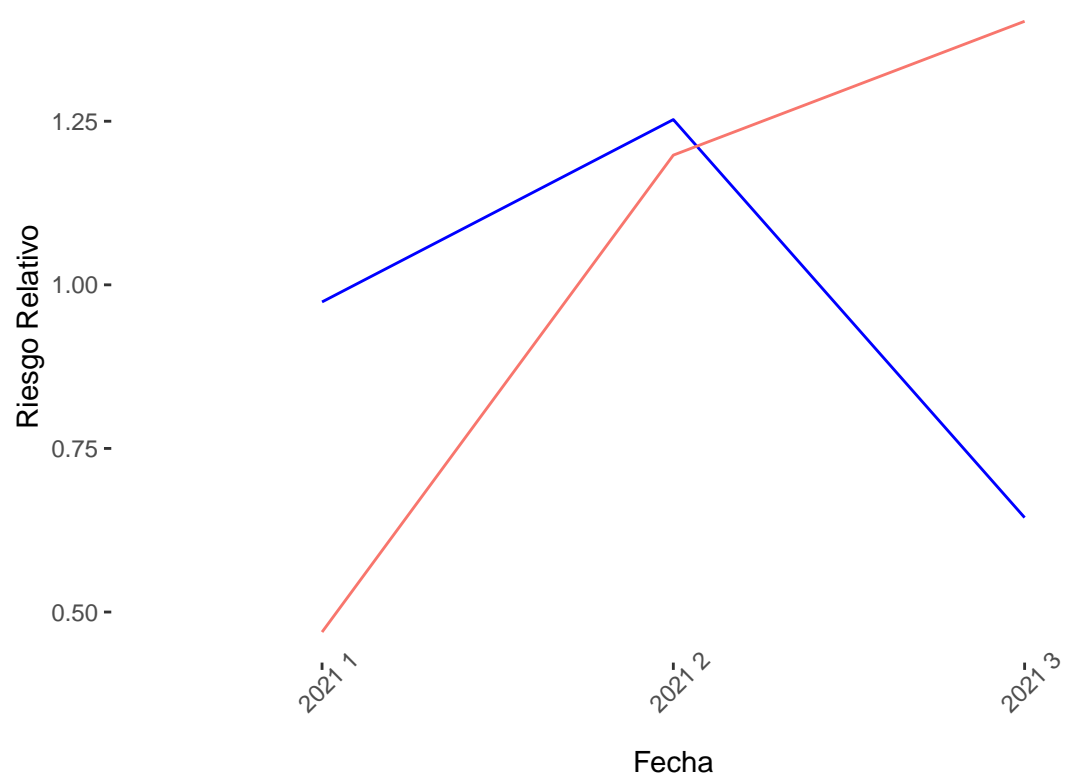


Predicción 3 meses training hasta: 1



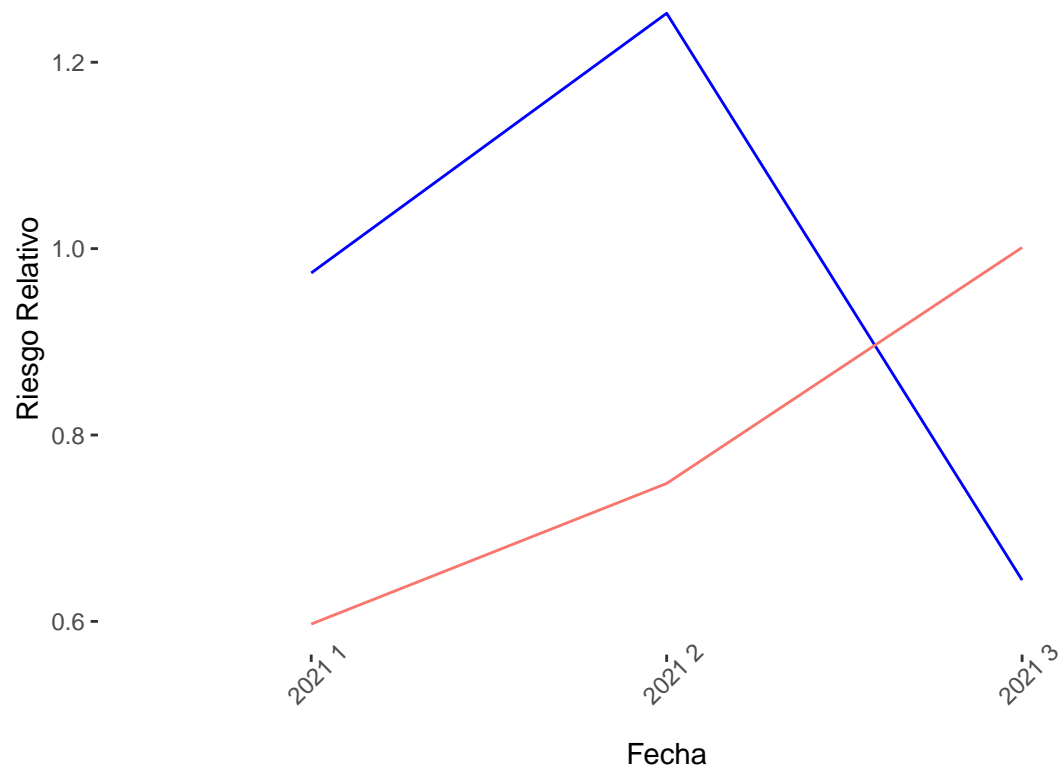
```
##  
## [[2]]
```

Predicción 3 meses training hasta: 0.95



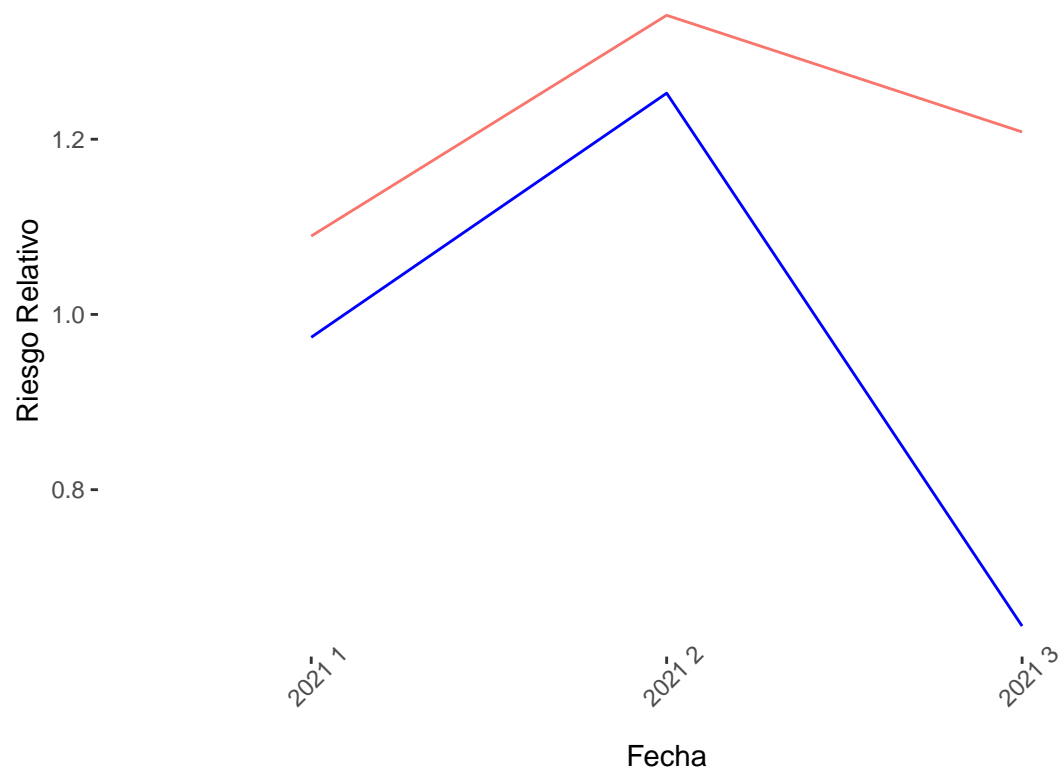
```
##  
## [[3]]
```

Predicción 3 meses training hasta: 0.9

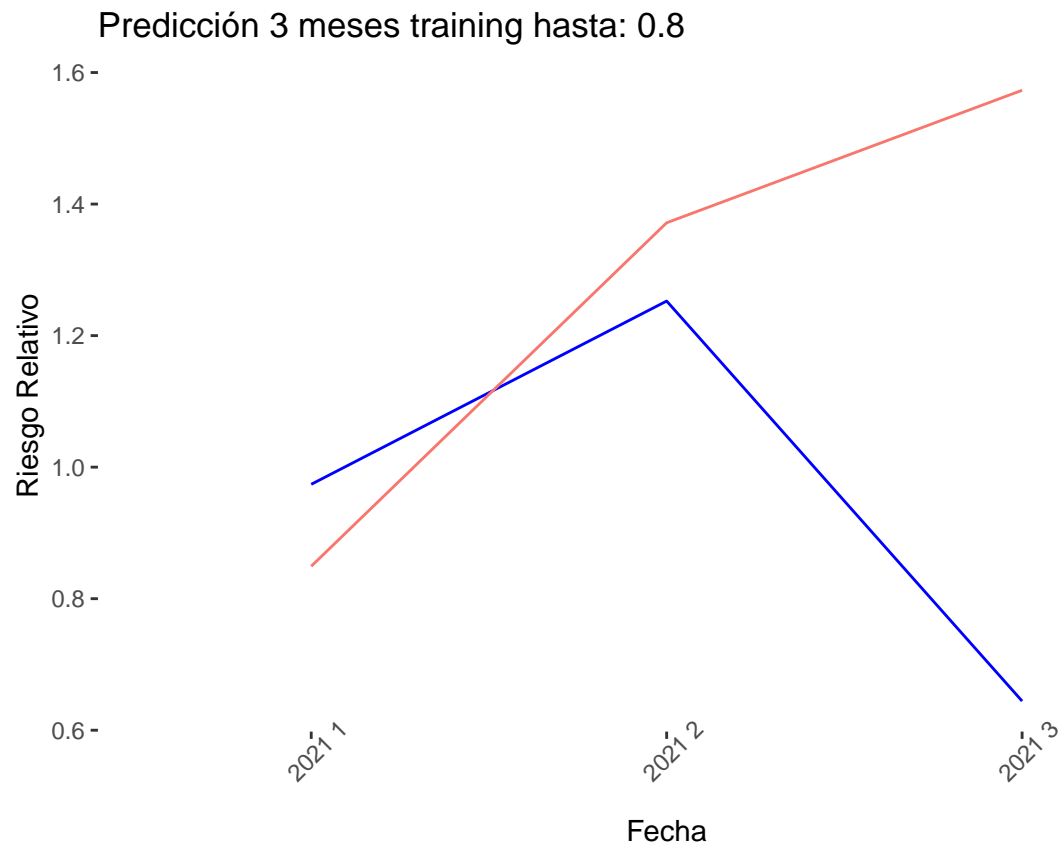


```
##  
## [[4]]
```

Predicción 3 meses training hasta: 0.85

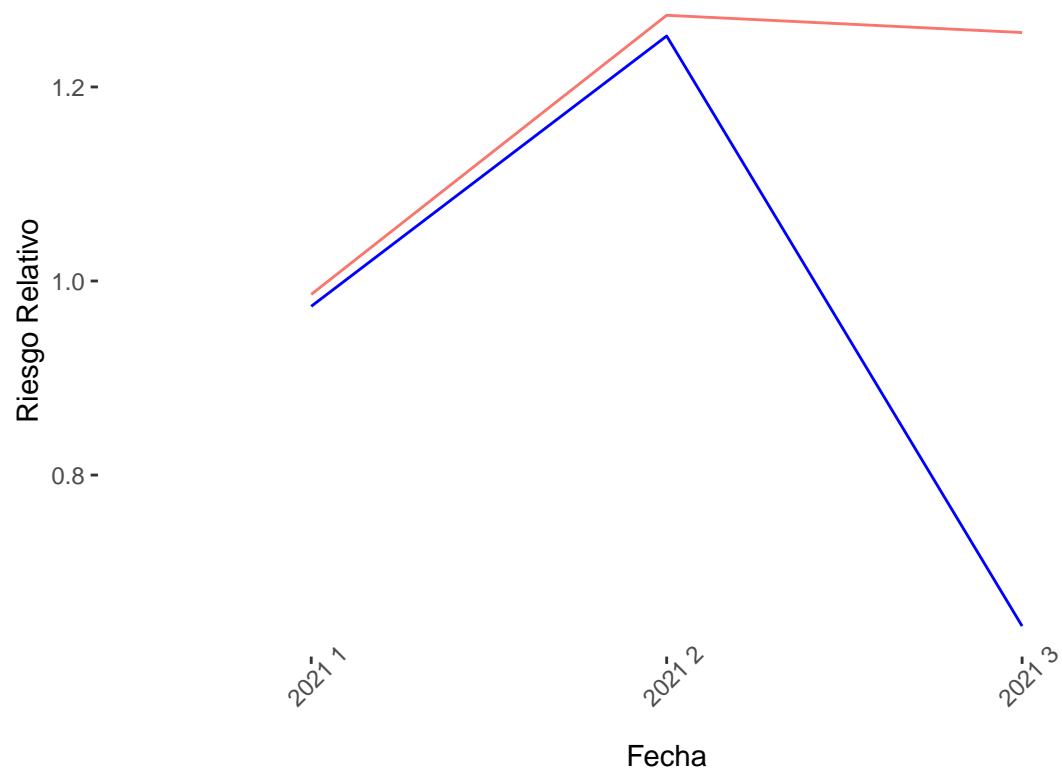


```
##  
## [[5]]
```

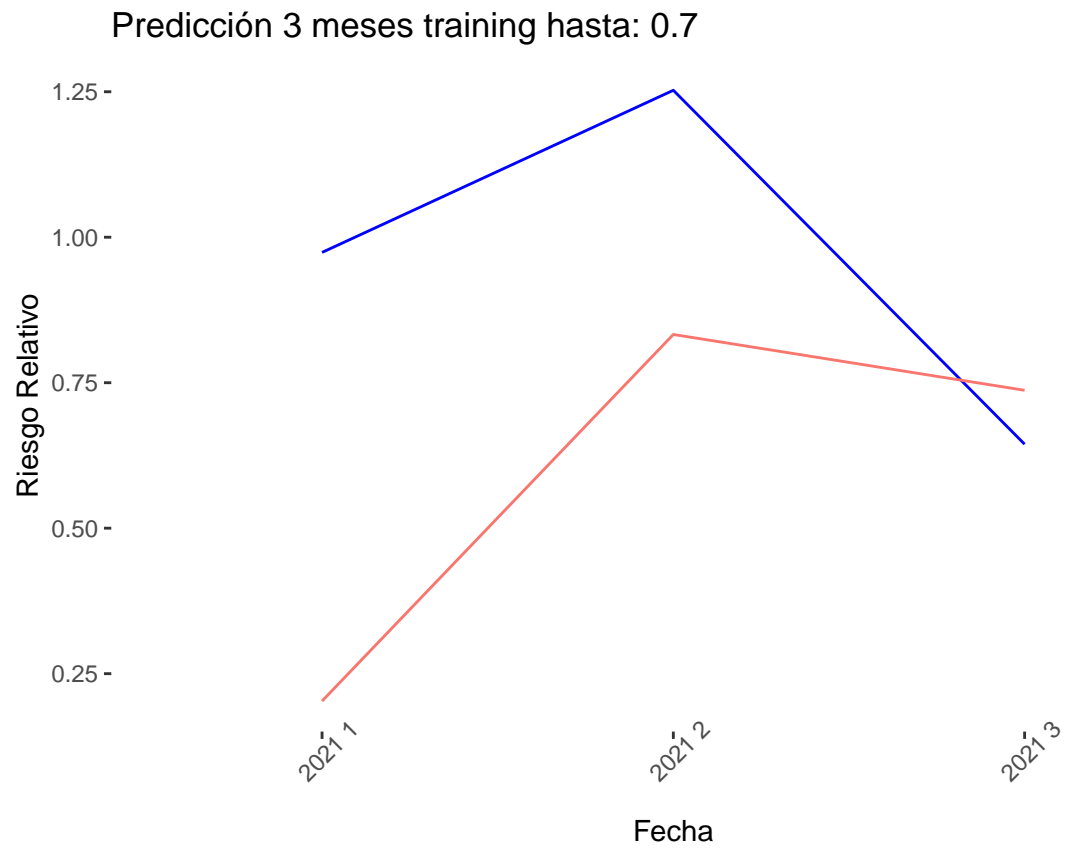


```
##  
## [[6]]
```

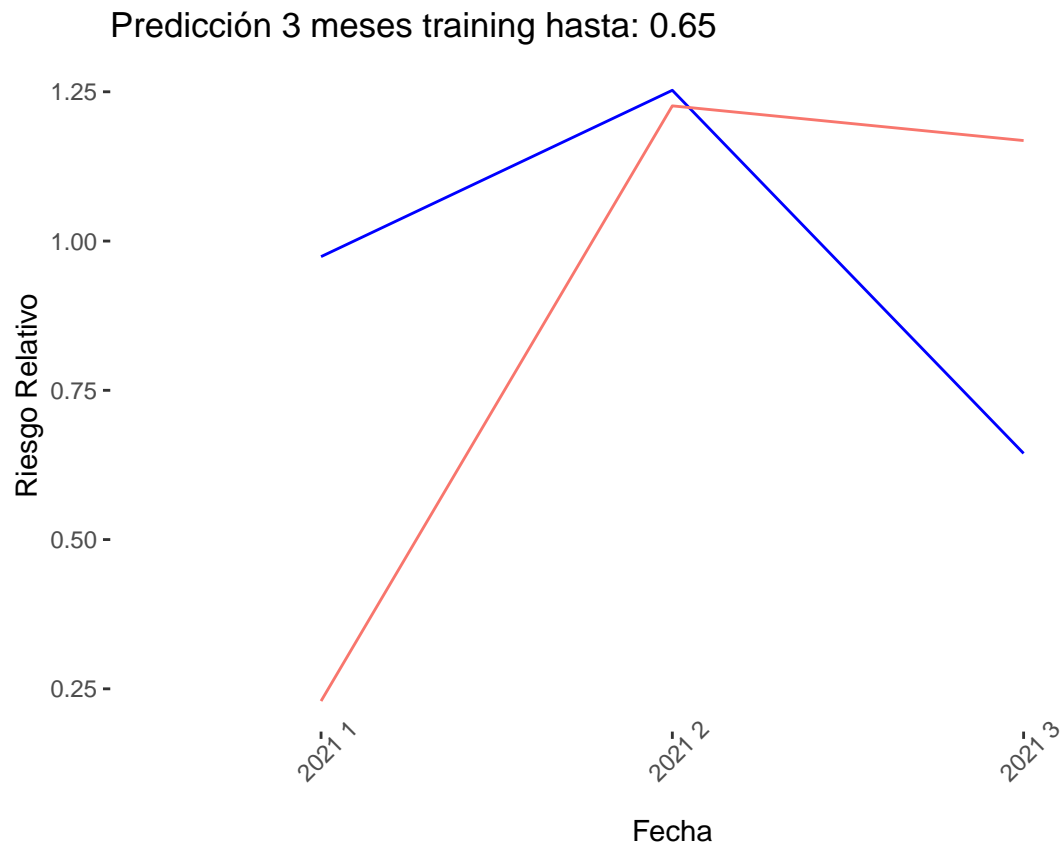
Predicción 3 meses training hasta: 0.75



##  
## [[7]]



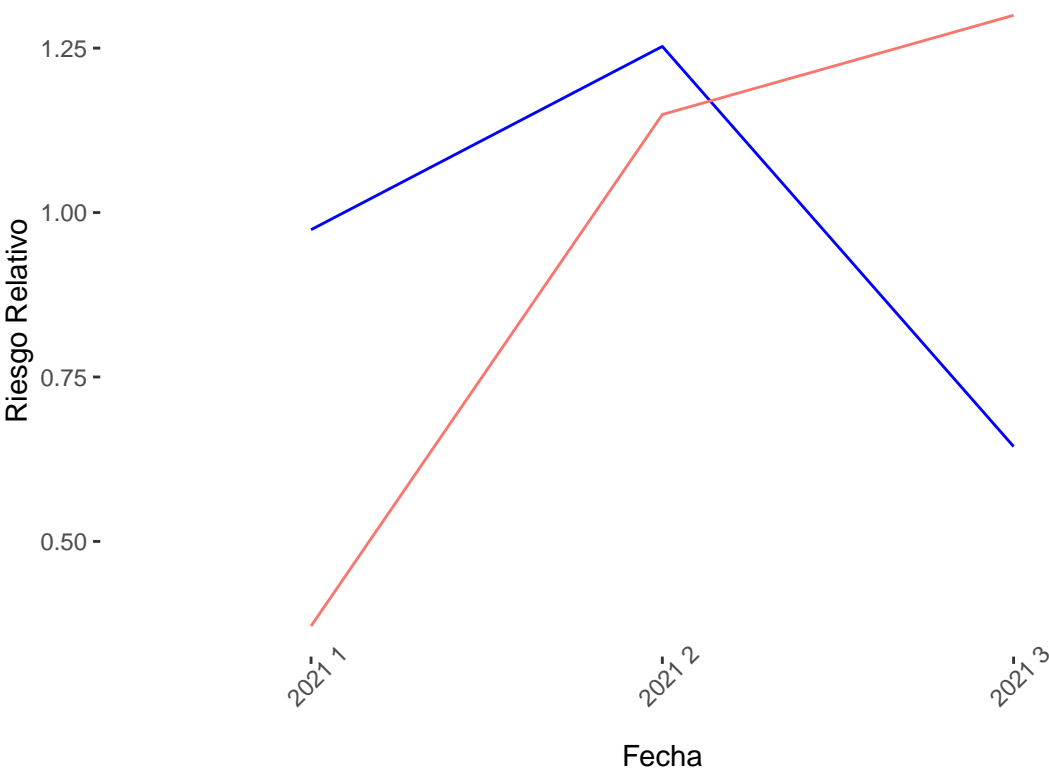
```
##  
## [[8]]
```



```
##  
## [[9]]
```



Predicción 3 meses training hasta: 0.6



##  
## [[10]]



```
##  
## [[11]]
```

