

intervencion

2022-10-08

#A. Posterior en t=19

```
m19 <- c(1.5, 1.8, -0.7)
```

```
C19 <- matrix(c(0.00002, 0.00001, -0.00002, 0.00001, 0.00003, -0.00001, -0.00001, 0.00002), ncol = 3)
```

#Valores conocidos de G20 y W20

```
G20 <- matrix(c(1.001, 0, 0, 0, 1, 0, 0, 0, 1), ncol=3)
```

```
G30 <- matrix(c(1.001, 0, 0, 0, 1, 0, 0, 0, 1), ncol=3)
```

```
W20 <- matrix(c(0.00001, 0, 0, 0, 0.00001, -0.00001, 0, -0.00001, 0.00001), ncol=3)
```

```
W30 <- matrix(c(0.00001, 0, 0, 0, 0.00001, -0.00001, 0, -0.00001, 0.00001), ncol=3)
```

#B. Priori de parámetros en t=20

```
a20 <- G20 %*% m19
```

```
R20 <- G20 %*% C19 %*% t(G20) + W20
```

```
a20
```

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

```
## [1,] 1.501
```

```
## [2,] 1.800
```

```
## [3,] -0.700
```

```
R20
```

```
##      [,1]      [,2]      [,3]
```

```
## [1,] 3.004e-05 1.001e-05 -2.002e-05
```

```
## [2,] 1.001e-05 4.000e-05 -2.000e-05
```

```
## [3,] -2.002e-05 -2.000e-05 7.000e-05
```

```
a20_int <- c(1.501, 2.3, -0.7)
```

```
R20_int <- matrix(c(3.004e-05, 1.001e-05, -2.002e-05, 1.001e-05, 0.005e-05, -2.000e-05, -2.002e-05, -2.000e-05, 7.000e-05), ncol=3)
```

```
F20 <- c(1, 6.06093, 4.51018) #Variables explicativas en t=20. El 1 es el intercepto  
#agregar el intercepto
```

```
S19 <- 0.00005 # Estimación de V en T=19
```

```
n19 <- 19.5 # Grados de libertad
```

```
#C. Pronóstico a un periodo.
```

```
f20 <- as.numeric(t(F20) %*% a20_int)
Q20 <- as.numeric(t(F20) %*% R20_int %*% F20 + S19)
f20
```

```
## [1] 12.28
```

```
Q20
```

```
## [1] 0.184
```

```
c(qst(0.025, nu = n19, mu = f20, sigma = sqrt(Q20)),
  qst(0.975, nu = n19, mu = f20, sigma = sqrt(Q20)))
```

```
## [1] 11.39 13.18
```

```
#Valor observado de Y20:
```

```
Y20 <- 11.6
```

```
#D. Posterior en t=20
```

```
A20 <- R20_int %*% F20 / Q20
```

```
e20 <- Y20-f20
```

```
m20 <- a20_int + A20 %*% e20
```

```
n20 <- n19 + 1
```

```
S20 <- S19 + (S19/n20)*(e20^2/Q20-1)
```

```
C20 <- (S20/S19)*(R20_int-A20 %*% t(A20) * Q20)
```

```
m20
```

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

```
## [1,] 1.5010
```

```
## [2,] 2.1877
```

```
## [3,] -0.7006
```

```
C20
```

```
##           [,1]           [,2]           [,3]
```

```
## [1,] 3.230e-05 1.069e-05 -2.153e-05
```

```
## [2,] 1.069e-05 3.862e-05 -5.232e-05
```

```
## [3,] -2.153e-05 -5.232e-05 7.509e-05
```

Distribuciones filtradas

```

#U20 <- chol(R20_int)
#Z20 <- chol(R20)
#K20 <- U20 %*% solve(Z20)
#G20_int <- K20 %*% G20
#W20_int <- K20 %*% W20 %*% t(K20)

#B19 <- C19 %*% t(G20_int) %*% solve(R20_int)
#a20_menos_1 <- m19 + B19 %*% (m20 - a20_int)
#R20_menos_1 <- C19 - B19 %*% (C20 - R20_int) %*% solve(B19)

#a20_menos_1

#S20/S19 * R20_menos_1

#F19 <- c(1, 6.05563      , 4.53957) #Variables explicativas en t=19. E
                                     #agregar el intercepto
#t(F19) %*% a20_menos_1

#S20/S19 * t(F19) %*% R20_menos_1 %*% F19

```

Mas de una obs.

```

datos_ej <- freeny %>%
  mutate(intercept = 1) %>%
  slice(20:n()) %>%
  dplyr::select(y, intercept, income.level, price.index) %>%
  mutate(y = ifelse(y>=9.6, y+0.5,y))
#Metemos ruido
set.seed(2)
datos_ej$y[11:20] <- datos_ej$y[11:20] + rnorm(10,0,0.05)
datos_ej

```

##		y	intercept	income.level	price.index
##	1967	9.314	1	6.061	4.510
##	1967.25	9.350	1	6.071	4.504
##	1967.5	9.358	1	6.080	4.494
##	1967.75	9.398	1	6.089	4.465
##	1968	9.421	1	6.102	4.449
##	1968.25	9.442	1	6.112	4.440
##	1968.5	9.487	1	6.116	4.420
##	1968.75	9.524	1	6.121	4.411
##	1969	9.540	1	6.122	4.412
##	1969.25	9.581	1	6.131	4.398

##	1969.5	10.056	1	6.147	4.385
##	1969.75	10.154	1	6.153	4.373
##	1970	10.223	1	6.156	4.328
##	1970.25	10.138	1	6.163	4.320
##	1970.5	10.196	1	6.174	4.309
##	1970.75	10.193	1	6.161	4.309
##	1971	10.253	1	6.182	4.306
##	1971.25	10.237	1	6.188	4.296
##	1971.5	10.375	1	6.194	4.278
##	1971.75	10.287	1	6.200	4.278

#Se asume que G_t , V_t y W_t son ctes conocidas para toda t .

actualizacion_V_desc <- function(datos, m0, C0, G, W, S0, n0, lista_in

```
error_varianza <- function(x) {
  if(any (diag(x)<0)) stop ('Elementos del parametro de escala o varian
```

```

  mt_menos_1 <- m0
  Ct_menos_1 <- C0
  St_menos_1 <- S0
  nt_menos_1 <- n0
  lista_at <- list()
  lista_Rt <- list()
  lista_ft <- list()
  lista_Qt <- list()
  lista_mt <- list()
  lista_Ct <- list()
  lista_CI <- list()
  lista_CI_inf <- list()
  lista_CI_sup <- list()
  lista_St <- list()
  interv <- F
  for(t in 1:length(datos$y)){

    if(t %in% lista_interv$t_int){
      at <- lista_interv$at_int[[match(t,list_interv$t_int)]]
      Rt <- lista_interv$Rt_int[[match(t,list_interv$t_int)]]
      interv <- T
    } else {
      at <- G %*% mt_menos_1
      Rt <- G %*% Ct_menos_1 %*% t(G) + W
    }
  }
}
```

```

Ft <- as.numeric(datos[t, 2:4])
ft <- t(Ft) %*% at
Qt <- t(Ft) %*% Rt %*% Ft + St_menos_1
CI <- c(qst(0.025, nu = nt_menos_1, mu = ft, sigma = sqrt(Qt)),
        qst(0.975, nu = nt_menos_1, mu = ft, sigma = sqrt(Qt)))
CI_inf <- CI[1]
CI_sup <- CI[2]
Yt <- datos[t,1]
At <- Rt %*% Ft %*% solve(Qt)
et <- Yt-ft
mt <- at + At %*% et
nt <- nt_menos_1 + 1
St <- as.numeric(St_menos_1 + (St_menos_1/nt) * (et^2 %*% solve(Qt)))
Ct <- (St/St_menos_1)*(Rt - At %*% Qt %*% t(At))

error_varianza(Rt)
error_varianza(St)
error_varianza(Ct)

lista_at[[t]] <- at
lista_Rt[[t]] <- Rt
lista_ft[[t]] <- ft
lista_Qt[[t]] <- Qt
lista_CI[[t]] <- CI
lista_CI_inf[[t]] <- CI_inf
lista_CI_sup[[t]] <- CI_sup
lista_mt[[t]] <- mt
lista_Ct[[t]] <- Ct
lista_St[[t]] <- St
mt_menos_1 <- mt
Ct_menos_1 <- Ct
nt_menos_1 <- nt
St_menos_1 <- St

interv <- F
}
return(list("at" = lista_at, "Rt" = lista_Rt, "ft" = lista_ft,
           "Qt" = lista_Qt, "CI" = lista_CI, "CI_inf" = lista_CI_inf,
           "CI_sup" = lista_CI_sup, "mt" = lista_mt, "Ct" = lista_Ct,
           "St" = lista_St))
}

```

```
res_dlm_no_int <- actualizacion_V_desc(datos_ej, m19, C19, G20, W20, S
```

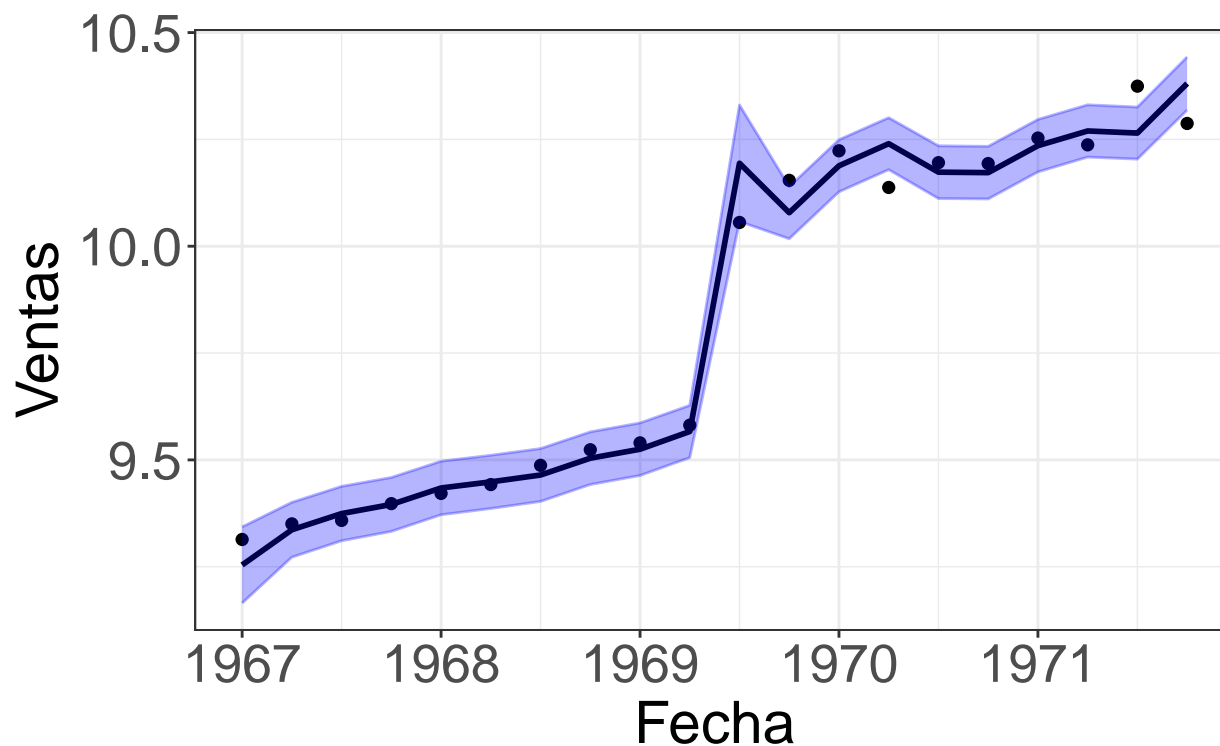
```
a30_int <- res_dlm_no_int$at[[11]] %>% as_vector()
a30_int[2] <- 1.9
R30_int <- res_dlm_no_int$Rt[[11]]
R30_int[2,2] <- 0.0002
list_interv <- list("t_int" = list(11), "at_int" = list(a30_int),
  "Rt_int" = list(R30_int))
```

```
res_dlm <- actualizacion_V_desc(datos_ej, m19, C19, G20, W20, S19, n19)
list_interv$at_int[[match(11,list_interv$t_int)]]
```

```
##           [,1]
## [1,]  1.5172
## [2,]  1.9000
## [3,] -0.6847
```

```
df_graficas <- data.frame("fecha" = datos_ej %>% row.names(), "y_real" =
  "y_pronostico" = res_dlm$ft %>% unlist(), "CI_inf" = res_dlm$CI_inf,
  "CI_sup" = res_dlm$CI_sup %>% unlist()) %>%
  mutate(fecha = as.numeric(fecha))
```

```
ggplot(data = df_graficas, aes(x = fecha)) +
  geom_point(aes(y = y_real, shape = "Observaciones"), size = 2) +
  geom_line(aes(y = y_pronostico, color = 'Pronósticos'), size = 1) +
  geom_line(aes(y = CI_inf), color = "blue", alpha = 0.3) +
  geom_line(aes(y = CI_sup), color = "blue", alpha = 0.3) +
  geom_ribbon(aes(ymax = CI_sup, ymin = CI_inf, fill = 'Intervalo al 95%'),
  theme_bw() +
  scale_colour_manual(
    name = "", values = c("Intervalo al 95%" = "transparent",
      "Pronósticos" = "black")) +
  scale_fill_manual(
    name = "", values = c("Intervalo al 95%" = "blue",
      "Pronósticos" = "transparent")) +
  theme(legend.position = "bottom") +
  labs(shape = "") +
  ylab('Ventas') +
  xlab('Fecha') +
  theme(axis.text.x = element_text(size = 20),
    axis.text.y = element_text(size = 20),
    axis.title = element_text(size = 22),
    legend.text = element_text(size=20))
```

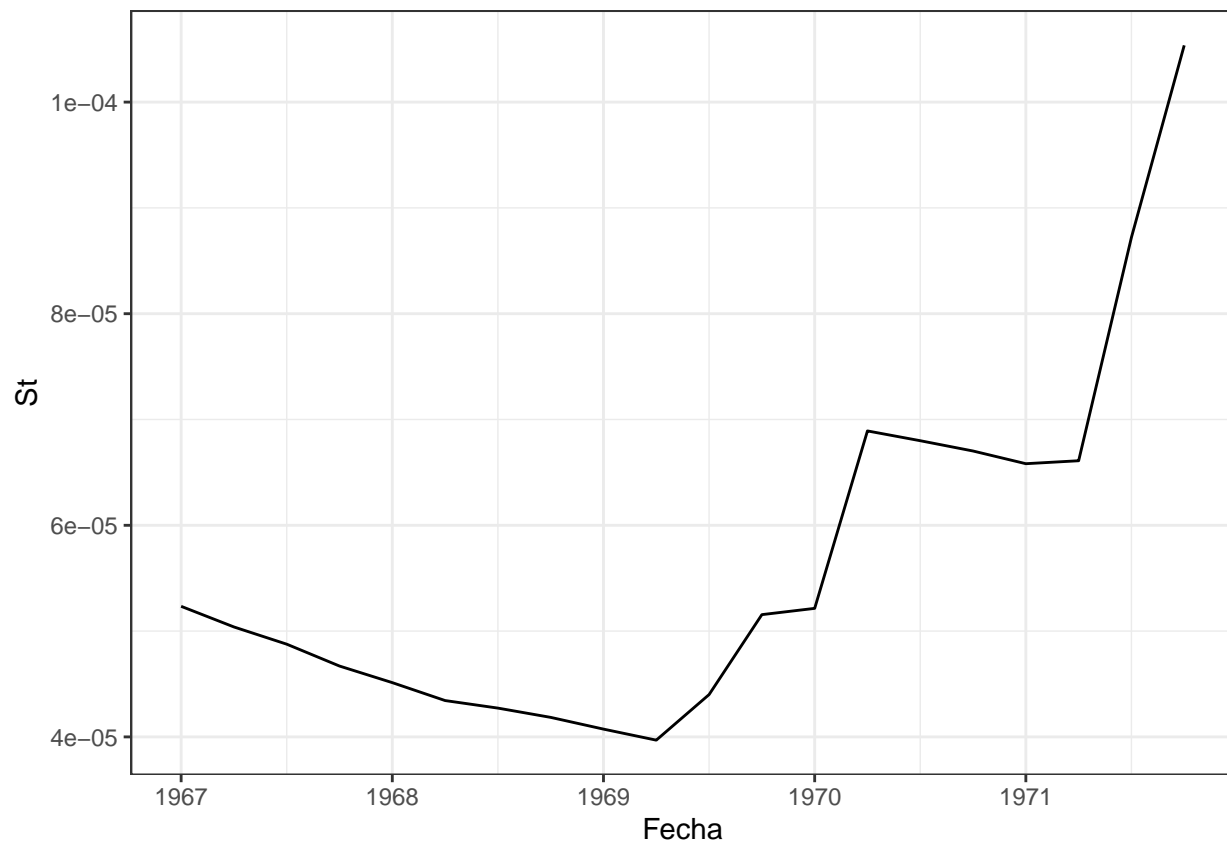


- Observaciones ■ Intervalo al 95% — Pronós

```
ggsave(filename = "graphs/teoria/intervencion/actualizacion_interv.png")
```

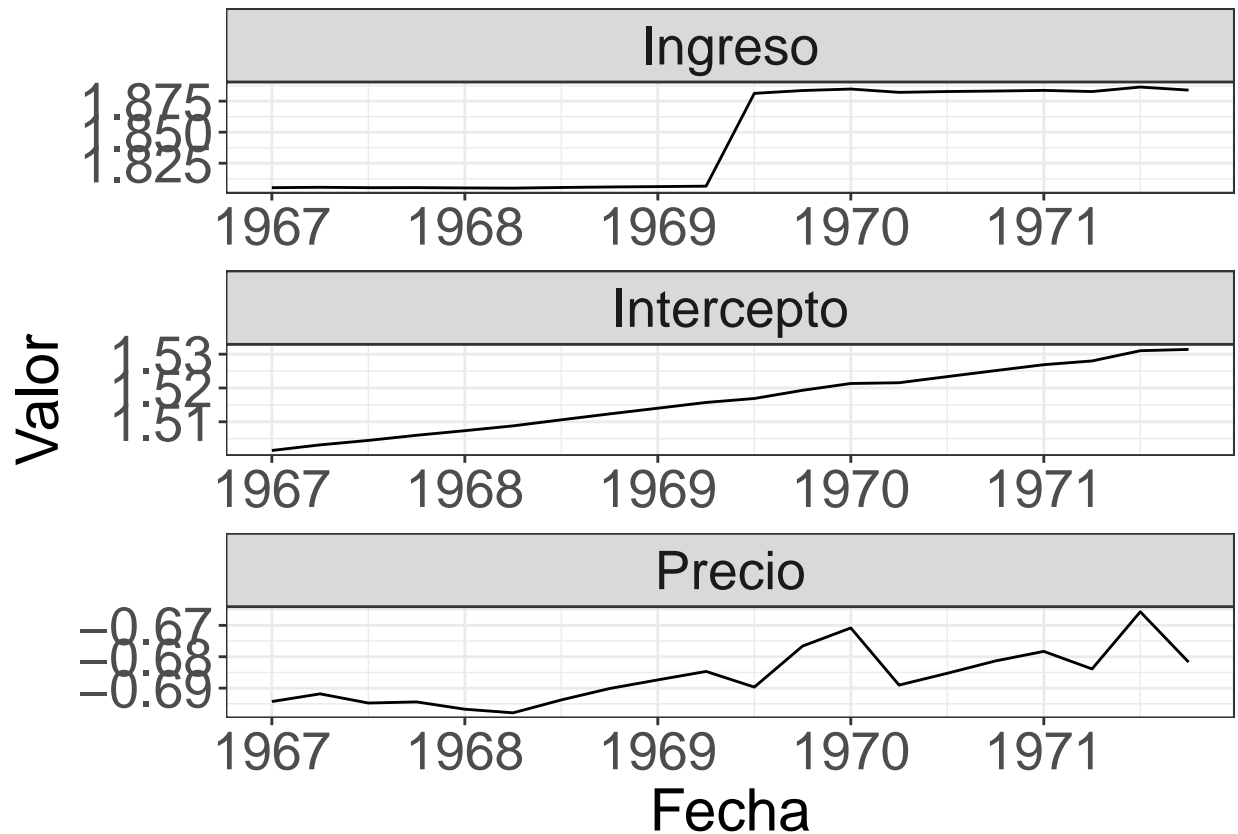
```
df_St <- data.frame(res_dlm$St %>% unlist(), fecha = df_graficas$fecha)
rename(df_St, St = 1)
```

```
ggplot(df_St, aes(x=fecha, y = St)) +
  geom_line() +
  theme_bw() +
  ylab("St") +
  xlab("Fecha")
```



```
df_params <- data.frame(reduce(res_dlm$mt, cbind) %>% t(), fecha = df_g
  rename(Intercepto = X1, Ingreso = X2, Precio = X3) %>%
  pivot_longer(names_to = "parametro", values_to = "valor",
               cols = c(Intercepto, Ingreso, Precio))
```

```
ggplot(df_params, aes(x=fecha, y = valor)) +
  geom_line() +
  facet_wrap(~parametro, nrow = 3, scales = "free") +
  theme_bw() +
  ylab("Valor") +
  xlab("Fecha") +
  theme(axis.text.x = element_text(size = 20),
        axis.text.y = element_text(size = 20),
        axis.title = element_text(size = 22),
        strip.text = element_text(size=20))
```

```
#ggsave(filename = "graphs/teoria/actualizacion/parametros.png", width = 1000, height = 1000)
```

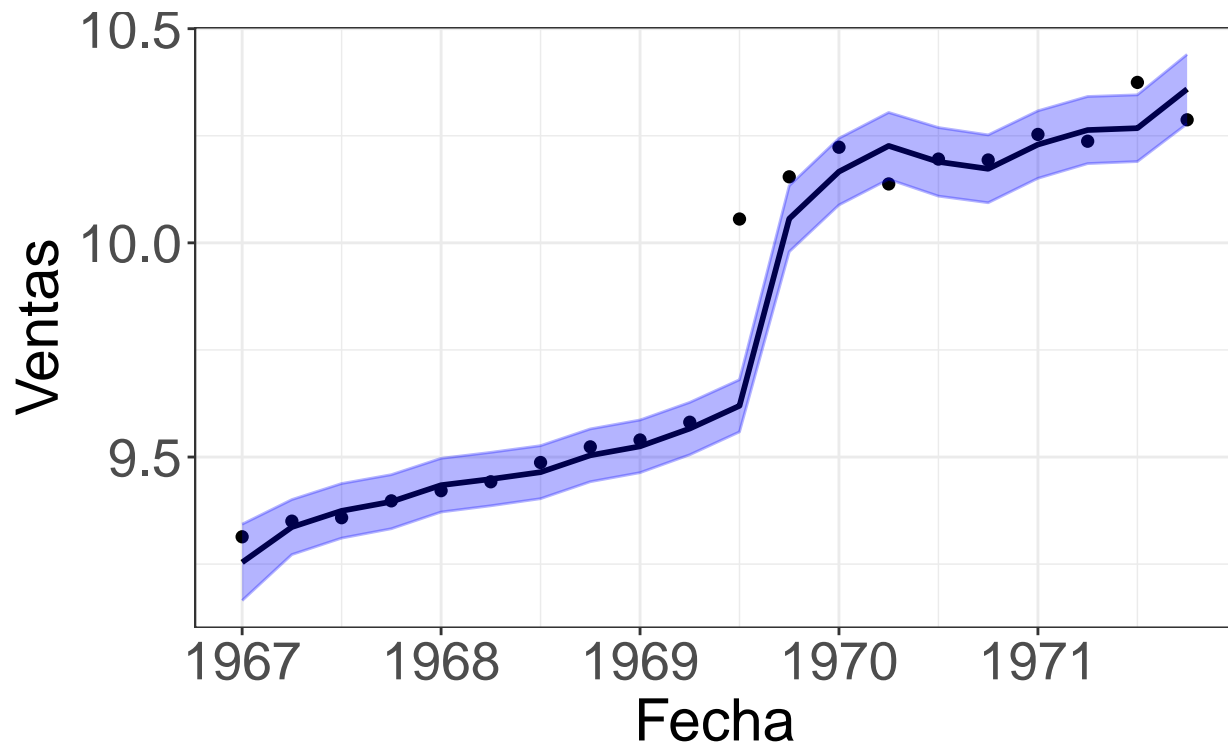
```
df_graficas_no_int <- data.frame("fecha" = datos_ej %>% row.names(), "y_real" = datos_ej[, "y_real"],
  "y_pronostico" = res_dlm_no_int$ft %>% unlist(), "CI_inf" = res_dlm_no_int$CI_inf %>% unlist(),
  "CI_sup" = res_dlm_no_int$CI_sup %>% unlist()) %>%
  mutate(fecha = as.numeric(fecha))
```

```
ggplot(data = df_graficas_no_int, aes(x = fecha)) +
  geom_point(aes(y = y_real, shape = "Observaciones"), size = 2) +
  geom_line(aes(y = y_pronostico, color = 'Pronósticos'), size = 1) +
  geom_line(aes(y = CI_inf), color = "blue", alpha = 0.3) +
  geom_line(aes(y = CI_sup), color = "blue", alpha = 0.3) +
  geom_ribbon(aes(ymax = CI_sup, ymin = CI_inf, fill = 'Intervalo al 95%')) +
  theme_bw() +
  scale_colour_manual(
    name = "", values = c("Intervalo al 95%" = "transparent",
                          "Pronósticos" = "black")) +
  scale_fill_manual(
    name = "", values = c("Intervalo al 95%" = "blue",
                          "Pronósticos" = "transparent")) +
```

```

theme(legend.position = "bottom") +
labs(shape = "") +
ylab('Ventas') +
xlab('Fecha') +
theme(axis.text.x = element_text(size = 20),
      axis.text.y = element_text(size = 20),
      axis.title = element_text(size = 22),
      legend.text = element_text(size=20))

```



- Observaciones
- Intervalo al 95% — Pronós

```

ggsave(filename = "graphs/teoria/intervencion/actualizacion_no_interv.")
#Se pasan manualmente las G historicas en G, las de intervencion se
suavizamiento_V_desc <- function(datos, G, nt, lista_interv){

  error_varianza <- function(x) {
    if(any(diag(x)<0)) stop ('Elementos del parametro de escala o varian
  }

  lista_ft_k_filt <- list()
  lista_at_k_filt <- list()

```

```

lista_Rt_k_filt <- list()
lista_resp_med_esc <- list()
lista_CI_inf <- list()
lista_CI_sup <- list()

#Se calculan las distribuciones filtradas para k = T
at_k_filt_mas_1 <- res_dlm$mt[[length(datos$y)]]
Rt_k_filt_mas_1 <- res_dlm$Ct[[length(datos$y)]]
St <- res_dlm$St[[length(datos$y)]]

Ft <- as.numeric(datos[length(datos$y), 2:4])
lista_ft_k_filt[[length(datos$y)]] <- t(Ft) %*% at_k_filt_mas_1
lista_at_k_filt[[length(datos$y)]] <- at_k_filt_mas_1
lista_Rt_k_filt[[length(datos$y)]] <- Rt_k_filt_mas_1
lista_resp_med_esc[[length(datos$y)]] <- t(Ft) %*% Rt_k_filt_mas_1 %
sigma = sqrt(lista_resp_med_esc)
lista_CI_inf[[length(datos$y)]] <- qst(0.025, nu = nt, mu = lista_ft_k_filt,
sigma = sqrt(lista_resp_med_esc))
lista_CI_sup[[length(datos$y)]] <- qst(0.975, nu = nt, mu = lista_ft_k_filt,
sigma = sqrt(lista_resp_med_esc))

interv <- F

error_varianza(Rt_k_filt_mas_1)
error_varianza(St)

for(i in length(datos$y):2){
  #Se calculan las distribuciones filtradas para i-1. De lo mas reciente

  if(i %in% lista_interv$t_int){
    at_int <- lista_interv$at_int[[match(i, lista_interv$t_int)]]
    Rt_int <- lista_interv$Rt_int[[match(i, lista_interv$t_int)]]
    Rt <- res_dlm_no_int$Rt[[i]]
    Ut <- t(chol(Rt_int))
    Zt <- t(chol(Rt))
    Kt <- Ut %*% solve(Zt)
    Gt_int <- Kt %*% G
    interv <- T
  }

  Ct_k <- res_dlm$Ct[[i-1]]
  Rt_k_mas_1 <- res_dlm$Rt[[i]]
  mt_k <- res_dlm$mt[[i-1]]
  at_k_mas_1 <- res_dlm$at[[i]]
  St_k_mas_1 <- res_dlm$St[[i]]
}

```

```

St_k <- res_dlm$St[[i-1]]

Bt_k <- Ct_k %*% t(`if`(interv,Gt_int, G)) %*% solve(Rt_k_mas_1)
at_k_filt <- mt_k + Bt_k %*% (at_k_filt_mas_1 - at_k_mas_1)
Rt_k_filt <- Ct_k + Bt_k %*% (Rt_k_filt_mas_1 - Rt_k_mas_1) %*% t(
params_esc <- (St/St_k)*Rt_k_filt

Ft_k <- as.numeric(datos[i-1, 2:4])
ft_k_filt <- t(Ft_k) %*% at_k_filt
resp_med_esc <- (St/St_k)*(t(Ft_k) %*% Rt_k_filt %*% Ft_k)
CI <- c(qst(0.025, nu = nt, mu = ft_k_filt, sigma = sqrt(resp_med_esc),
           qst(0.975, nu = nt, mu = ft_k_filt, sigma = sqrt(resp_med_esc)

error_varianza(Rt_k_mas_1)
error_varianza(Ct_k)
error_varianza(St_k_mas_1)
error_varianza(St_k)
error_varianza(Rt_k_filt)

lista_ft_k_filt[[i-1]] <- ft_k_filt
lista_at_k_filt[[i-1]] <- at_k_filt
lista_Rt_k_filt[[i-1]] <- Rt_k_filt
lista_resp_med_esc[[i-1]] <- resp_med_esc
lista_CI_inf[[i-1]] <- CI[1]
lista_CI_sup[[i-1]] <- CI[2]

at_k_filt_mas_1 <- at_k_filt
Rt_k_filt_mas_1 <- Rt_k_filt
interv <- F
}

return(list("ft_k_filt" = lista_ft_k_filt, "at_k_filt" = lista_at_k_filt,
           "Rt_k_filt" = lista_Rt_k_filt, "resp_media_esc" = lista_resp_med_esc,
           "CI_inf" = lista_CI_inf, "CI_sup" = lista_CI_sup))
}

res_dlm_suav <- suavizamiento_V_desc(datos_ej, G20, 39.5, list_interv)

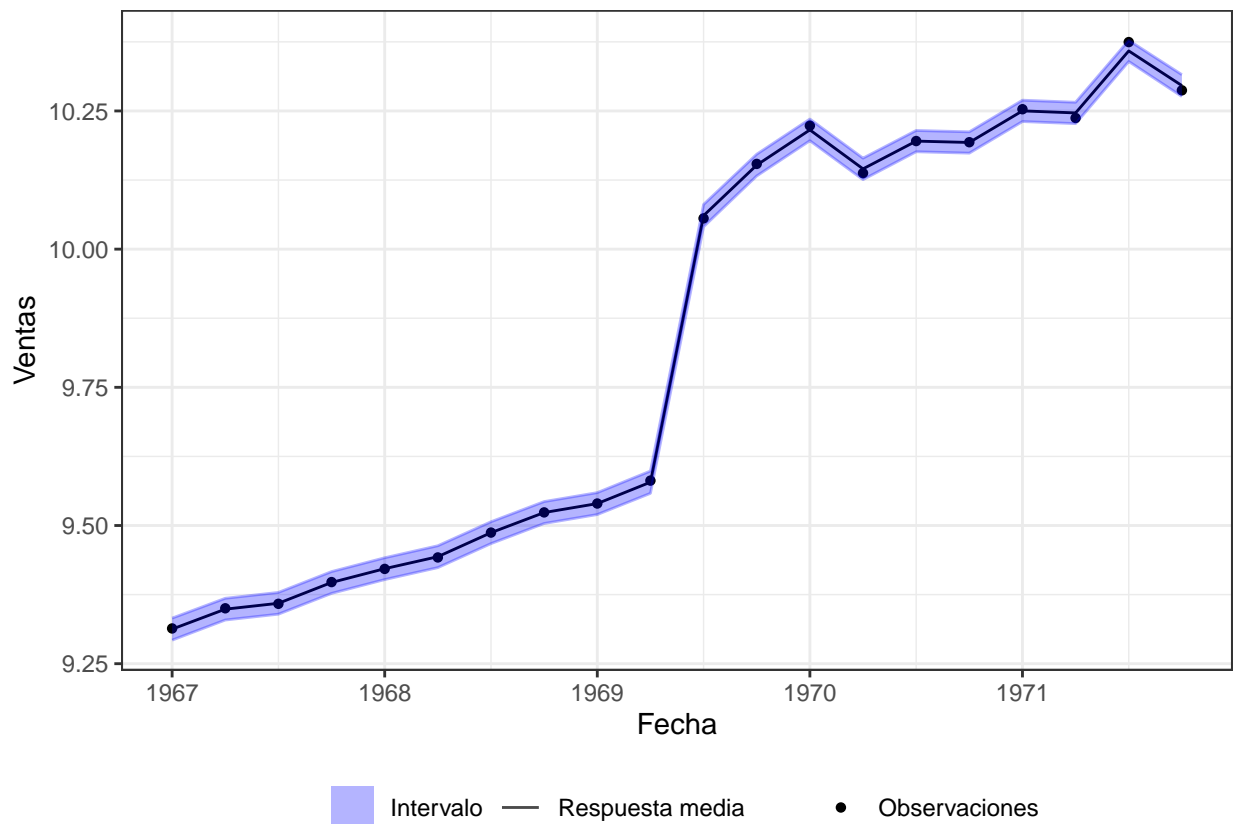
df_graficas_suav <- data.frame("fecha" = datos_ej %>% row.names(), "y_1" =
  "resp_media" = res_dlm_suav$ft_k_filt %>% unlist(), "CI_inf" =
  "CI_sup" = res_dlm_suav$CI_sup %>% unlist()) %>%
  mutate(fecha = as.numeric(fecha))

```

```

ggplot(data = df_graficas_suav, aes(x = fecha)) +
  geom_point(aes(y = y_real, shape = "Observaciones")) +
  geom_line(aes(y = resp_media, color = 'Respuesta media')) +
  geom_line(aes(y = CI_inf), color = "blue", alpha = 0.3) +
  geom_line(aes(y = CI_sup), color = "blue", alpha = 0.3) +
  geom_ribbon(aes(ymax = CI_sup, ymin = CI_inf, fill = 'Intervalo'), alpha = 0.3) +
  theme_bw() +
  scale_colour_manual(
    name = "", values = c("Intervalo" = "transparent",
                          "Respuesta media" = "black")) +
  scale_fill_manual(
    name = "", values = c("Intervalo" = "blue",
                          "Respuesta media" = "transparent")) +
  theme(legend.position = "bottom") +
  labs(shape = "") +
  ylab('Ventas') +
  xlab('Fecha')

```



Paso a paso

```
#Valores conocidos de G30 y W30 no son necesarios por la intervención
```

```
#B. Priori de parámetros en t=30 intervenidos
```

```
a30_int
```

```
##           [,1]
## [1,]  1.5172
## [2,]  1.9000
## [3,] -0.6847
```

```
R30_int
```

```
##           [,1]      [,2]      [,3]
## [1,]  1.126e-04  6.172e-06 -0.0000318
## [2,]  6.172e-06  2.000e-04 -0.0001445
## [3,] -3.180e-05 -1.445e-04  0.0002463
```

```
F30 <- c(1, 6.14705, 4.38513) #Variables explicativas en t=30. El 1 es  
                                #agregar el intercepto
```

```
S29 <- res_dlm$St[[10]] # Estimación de V en T=29
```

```
n29 <- 29.5 # Grados de libertad
```

```
#C. Pronóstico a un periodo.
```

```
f30 <- as.numeric(t(F30) %*% a30_int)
```

```
Q30 <- as.numeric(t(F30) %*% R30_int %*% F30 + S29)
```

```
f30
```

```
## [1] 10.19
```

```
Q30
```

```
## [1] 0.004455
```

```
c(qst(0.025, nu = n29, mu = f30, sigma = sqrt(Q30)),  
  qst(0.975, nu = n29, mu = f30, sigma = sqrt(Q30)))
```

```
## [1] 10.06 10.33
```

```
#Valor observado de Y30:
```

```
Y30 <- 10.05563427
```

```
#D. Posterior en t=20
```

```

A30 <- R30_int %*% F30 / Q30
e30 <- Y30-f30
m30 <- a30_int + A30 %*% e30
n30 <- n29 + 1
S30 <- S29 + (S29/n30)*(e30^2/Q30-1)
C30 <- (S30/S29)*(R30_int-A30 %*% t(A30) * Q30)
m30

```

```

##           [,1]
## [1,]  1.5169
## [2,]  1.8813
## [3,] -0.6897

```

C30

```

##           [,1]      [,2]      [,3]
## [1,]  1.248e-04  5.178e-06 -3.569e-05
## [2,]  5.178e-06  1.315e-04 -1.842e-04
## [3,] -3.569e-05 -1.842e-04  2.667e-04

```

Distribuciones filtradas

..

```

R30 <- res_dlm_no_int$Rt[[11]]
# La descomposicion de Cholesky en R devuelve una triangular superior
# para hacerla triangular inferior para que coincida con la teoría pr
U30 <- t(chol(R30_int))
Z30 <- t(chol(R30))
K30 <- U30 %*% solve(Z30)
G30_int <- K30 %*% G30
W30_int <- K30 %*% W30 %*% t(K30)

```

```

C29 <- res_dlm$Ct[[10]]
m29 <- res_dlm$mt[[10]]
a39_menos_9 <- res_dlm_suav$at_k_filt[[11]]
R39_menos_9 <- res_dlm_suav$Rt_k_filt[[11]]

B29 <- C29 %*% t(G30_int) %*% solve(R30_int)
a39_menos_10 <- m29 + B29 %*% (a39_menos_9 - a30_int)
R39_menos_10 <- C29 + B29 %*% (R39_menos_9 - R30_int) %*% t(B29)

a39_menos_10

```

```
##           [,1]
## [1,] 1.5160
## [2,] 1.7960
## [3,] -0.6706
```

```
S39 <- res_dlm$St[[20]] # Estimación de V en T=39
```

```
S39/S29 * R39_menos_10
```

```
##           [,1]      [,2]      [,3]
## [1,] 5.839e-04 4.242e-05 -0.0001916
## [2,] 4.242e-05 3.633e-04 -0.0005233
## [3,] -1.916e-04 -5.233e-04 0.0007883
```

```
F29 <- c(1, 6.131, 4.398) #Variables explicativas en t=29.
t(F29) %*% a39_menos_10
```

```
##           [,1]
## [1,] 9.578
```

```
S39/S29 * t(F29) %*% R39_menos_10 %*% F29
```

```
##           [,1]
## [1,] 9.959e-05
```

```
U30 %*% t(U30)
```

```
##           [,1]      [,2]      [,3]
## [1,] 1.126e-04 6.172e-06 -0.0000318
## [2,] 6.172e-06 2.000e-04 -0.0001445
## [3,] -3.180e-05 -1.445e-04 0.0002463
```

```
Z30 %*% t(Z30)
```

```
##           [,1]      [,2]      [,3]
## [1,] 1.126e-04 6.172e-06 -0.0000318
## [2,] 6.172e-06 1.056e-04 -0.0001445
## [3,] -3.180e-05 -1.445e-04 0.0002463
```

```
a30 <- res_dlm_no_int$at[[11]] %>% as_vector()
h30 <- a30_int - K30 %*% a30
K30 %*% a30 + h30
```



```
##          [,1]
## [1,] 1.5172
## [2,] 1.9000
## [3,] -0.6847
```

```
K30 %*% R30 %*% t(K30)
```

```
##          [,1]      [,2]      [,3]
## [1,] 1.126e-04 6.172e-06 -0.0000318
## [2,] 6.172e-06 2.000e-04 -0.0001445
## [3,] -3.180e-05 -1.445e-04 0.0002463
```

```
R30_int
```

```
##          [,1]      [,2]      [,3]
## [1,] 1.126e-04 6.172e-06 -0.0000318
## [2,] 6.172e-06 2.000e-04 -0.0001445
## [3,] -3.180e-05 -1.445e-04 0.0002463
```