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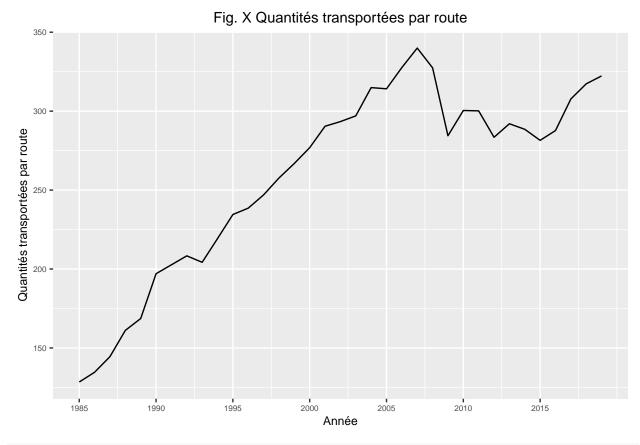
Janvier 2021

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1 Introduction

```
library(usethis)
library(readxl)
library(gitcreds)
library(strucchange)
## Le chargement a nécessité le package : zoo
##
## Attachement du package : 'zoo'
## Les objets suivants sont masqués depuis 'package:base':
##
##
       as.Date, as.Date.numeric
## Le chargement a nécessité le package : sandwich
library(gets)
## Le chargement a nécessité le package : parallel
library(ggplot2)
library(kableExtra)
library(pander)
Transport_France2019 <- read_excel("Transport_France2019_v2.xlsx")</pre>
##Vecteurs des séries
#Qtt_Trsp_route <- Transport_France2019$Qtt_Trsp_route
#Qtt_Trsp_train <- Transport_France2019$Qtt_Trsp_train
#Pdiesel <- Transport_France2019$Pdiesel</pre>
#Qdiesel <- Transport France2019$QDiesel
#GDP <- Transport France2019$GDP
#CPI <- Transport_France2019$CPI</pre>
#QDieselCamion <- Transport_France2019$Qdieselcamion
##Séries temporelles
Qtt_Trsp_route.ts <- ts(Transport_France2019$Qtt_Trsp_route, start=c(1985) , end=c(2019), frequency=1)</pre>
Qtt_Trsp_train.ts <- ts(Transport_France2019$Qtt_Trsp_train, start=c(1985) , end=c(2019), frequency=1)</pre>
Pdiesel.ts <- ts(Transport_France2019$Pdiesel, start=c(1985), end=c(2019), frequency=1)
Qdiesel.ts <- ts(Transport_France2019$QDiesel, start=c(1985), end=c(2019), frequency=1)
GDP.ts <- ts(Transport France2019$"PIB en volume (en milliards d'euros 2014)", start=c(1985), end=c(20
CPI.ts <- ts(Transport_France2019$CPI, start=c(1985) , end=c(2019), frequency=1)</pre>
Qdieselcamion.ts <- ts(Transport_France2019$Qdieselcamion, start=c(1985), end=c(2019), frequency=1)
##Graph en niveau
ggplot() +
  geom_line( aes(x = Transport_France2019$Year,y = Transport_France2019$Qtt_Trsp_route))+
labs(x = "Année", y = "Quantités transportées par route", title = "Fig. X Quantités transportées par ro
    theme(axis.text=element_text(size=6),legend.text=element_text(size=7),axis.title=element_text(size=
  scale_x_continuous(breaks=seq(1985,2019,5))
```



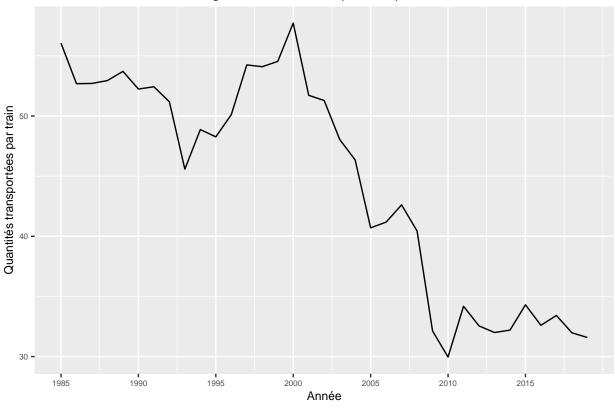


Fig. X Quantités transportées par train

```
ggplot() +
  geom_line( aes(x = Transport_France2019$Year,y = Transport_France2019$Pdiesel))+
labs(x = "Année", y = "Prix du disel", title = "Fig. X Prix du diesel") +
  theme(axis.text=element_text(size=6),legend.text=element_text(size=7),axis.title=element_text(size=scale_x_continuous(breaks=seq(1985,2019,5))
```

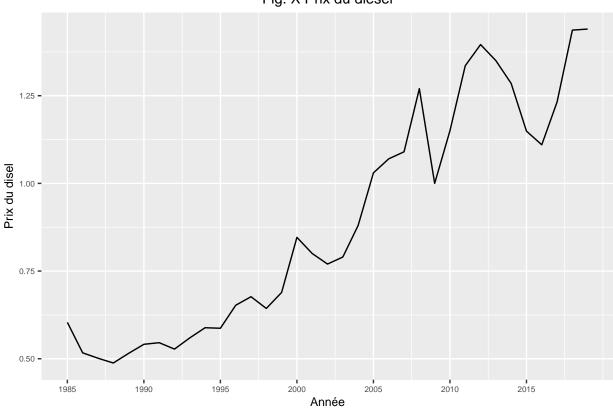


Fig. X Prix du diesel

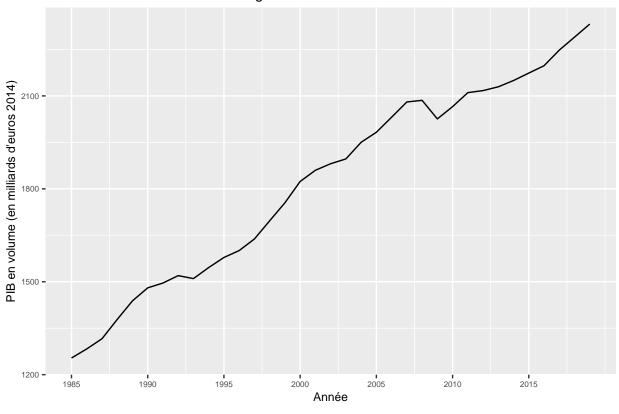


Fig. X Produit intérieur brut

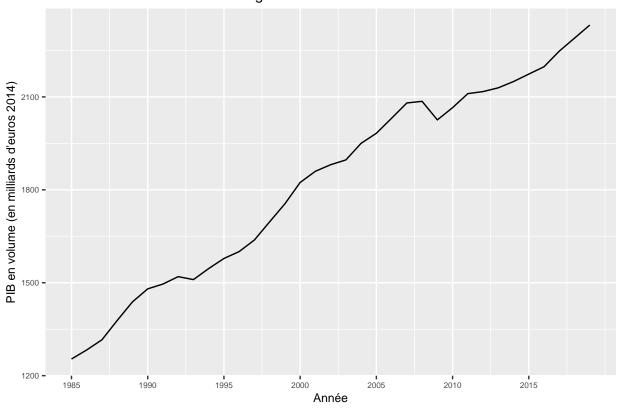
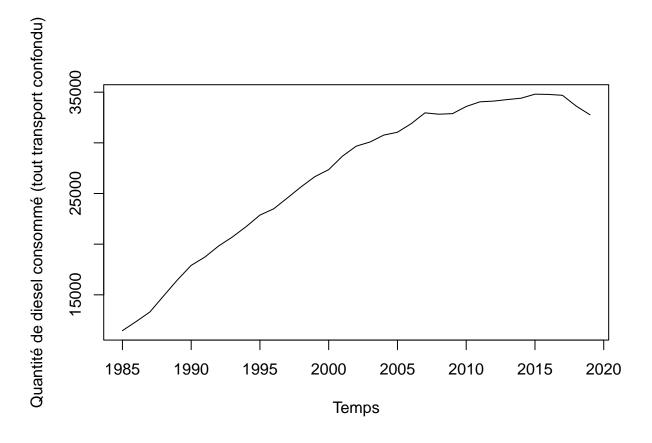


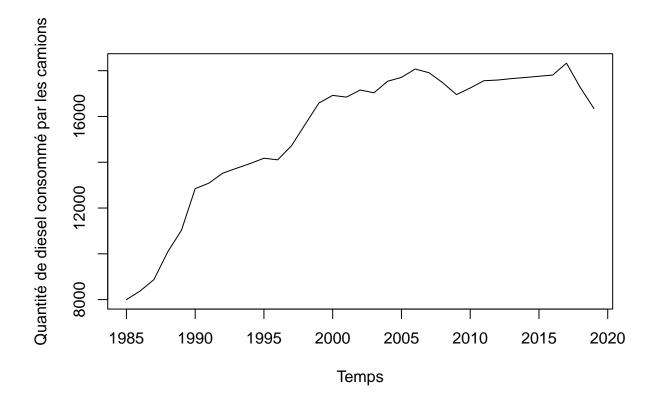
Fig. X Produit intérieur brut

plot(Qdiesel.ts, xlab = "Temps", ylab = "Quantité de diesel consommé (tout transport confondu)")

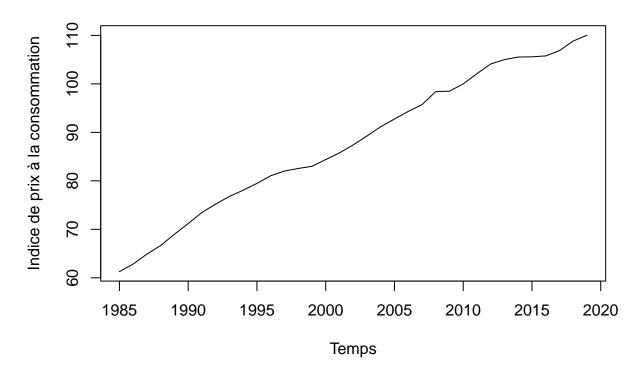


plot(Qdieselcamion.ts, xlab = "Temps", ylab = "Quantité de diesel consommé par les camions")

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plot(CPI.ts, xlab = "Temps", ylab = "Indice de prix à la consommation")



```
n=length(Transport_France2019$Qdieselcamion)

vec <- c(Transport_France2019$Pdiesel/Transport_France2019$CPI,Transport_France2019$"PIB en volume (en :
X <- matrix( vec, ncol=3)
Y=matrix(Transport_France2019$Qdieselcamion,n,1)
q=ncol(Y);
k=ncol(X);
K=k+1

y=Y
x=X

nobs=cbind(1:n)

OLS=lm(formula = y ~ x)

summary(OLS) %>% pander
```

	Estimate	Std. Error	t value	$\Pr(> t)$
(Intercept)	1767	790.8	2.235	0.03277
$\mathbf{x}1$	-378013	120267	-3.143	0.003667
$\mathbf{x2}$	4.242	1.423	2.98	0.005559
x3	36.89	5.615	6.57	2.449e-07

Table 2: Fitting linear model: $y \sim x$

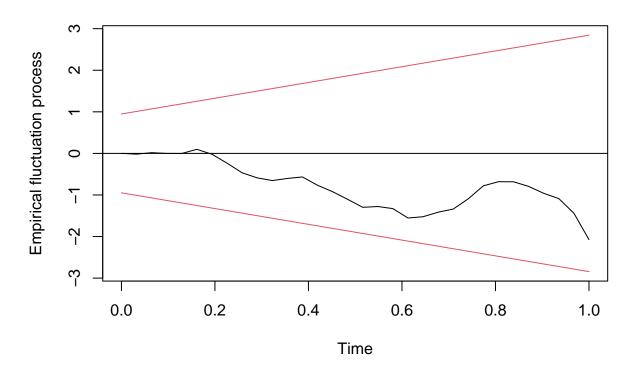
Observations	Residual Std. Error	R^2	Adjusted \mathbb{R}^2
35	721.2	0.9475	0.9424

kable

```
## function (x, format, digits = getOption("digits"), row.names = NA,
##
       col.names = NA, align, caption = NULL, label = NULL, format.args = list(),
##
       escape = TRUE, ...)
## {
##
       format = kable_format(format)
##
       if (!missing(align) && length(align) == 1L && !grepl("[^lcr]",
##
##
           align = strsplit(align, "")[[1]]
##
       if (inherits(x, "list")) {
##
           format = kable format latex(format)
##
           res = lapply(x, kable, format = format, digits = digits,
##
               row.names = row.names, col.names = col.names, align = align,
##
               caption = NA, format.args = format.args, escape = escape,
##
##
           return(kables(res, format, caption, label))
##
##
       caption = kable_caption(label, caption, format)
##
       if (!is.matrix(x))
##
           x = as.data.frame(x)
       if (identical(col.names, NA))
##
##
           col.names = colnames(x)
##
       m = ncol(x)
       isn = if (is.matrix(x))
##
##
           rep(is.numeric(x), m)
##
       else sapply(x, is.numeric)
##
       if (missing(align) || (format == "latex" && is.null(align)))
           align = ifelse(isn, "r", "l")
##
##
       digits = rep(digits, length.out = m)
##
       for (j in seq_len(m)) {
##
           if (is_numeric(x[, j]))
##
               x[, j] = round(x[, j], digits[j])
##
       }
##
       if (any(isn)) {
##
           if (is.matrix(x)) {
               if (is.table(x) && length(dim(x)) == 2)
##
                   class(x) = "matrix"
##
##
               x = format_matrix(x, format.args)
##
           else x[, isn] = format_args(x[, isn], format.args)
##
##
       }
       if (is.na(row.names))
##
##
           row.names = has_rownames(x)
##
       if (!is.null(align))
##
           align = rep(align, length.out = m)
       if (row.names) {
##
           x = cbind(`` = rownames(x), x)
##
```

```
##
           if (!is.null(col.names))
               col.names = c(" ", col.names)
##
           if (!is.null(align))
##
##
               align = c("l", align)
##
       }
##
       n = nrow(x)
##
       x = replace_na(to_character(x), is.na(x))
       if (!is.matrix(x))
##
##
           x = matrix(x, nrow = n)
##
       x = trimws(x)
##
       colnames(x) = col.names
       if (format != "latex" && length(align) && !all(align %in%
##
           c("l", "r", "c")))
##
           stop("'align' must be a character vector of possible values 'l', 'r', and 'c'")
##
##
       attr(x, "align") = align
##
       if (format == "simple" && nrow(x) == 0)
##
           format = "pipe"
       res = do.call(paste("kable", format, sep = "_"), list(x = x,
##
##
           caption = caption, escape = escape, ...))
       structure(res, format = format, class = "knitr_kable")
##
## }
## <bytecode: 0x00000001fcf99f0>
## <environment: namespace:knitr>
names(OLS)
## [1] "coefficients" "residuals"
                                                         "rank"
                                         "effects"
## [5] "fitted.values" "assign"
                                         "qr"
                                                         "df.residual"
## [9] "xlevels"
                                         "terms"
                        "call"
                                                         "model"
xc = cbind(1,x)
bhat = OLS$coefficients
yf = xc %*% bhat
res = y - yf
scr = t(res) %*% res
d1 = t(res) %*% res
d2 = t(res[2:n]-res[1:n-1]) %*% (res[2:n]-res[1:n-1])
dw = d2/d1
print (dw)
##
             [,1]
## [1,] 0.5667082
Wr <- efp(y ~ x, type = "Rec-CUSUM")
plot(Wr)
```

Recursive CUSUM test



```
#
# Test Cusum Square
#
rr <- (recresid(y ~ x))
rr <- rr^2
cumrr <- cumsum(rr)/scr</pre>
```

Warning in cumsum(rr)/scr: Le recyclage d'un tableau (array) de longueur 1 dans un calcul arithmétiq tableau est obsolète.

Utilisez c() ou as.vector() à la place.

```
# Valeurs seuil de la distribution Cusum

# 
c0 = 0.18915

kp2=K+1

c0 = 0.18915 # valeur critique de c0

t2 <- ts(kp2:n)

t3=t2-1

smin <-((t2-k)/(n-k))-c0

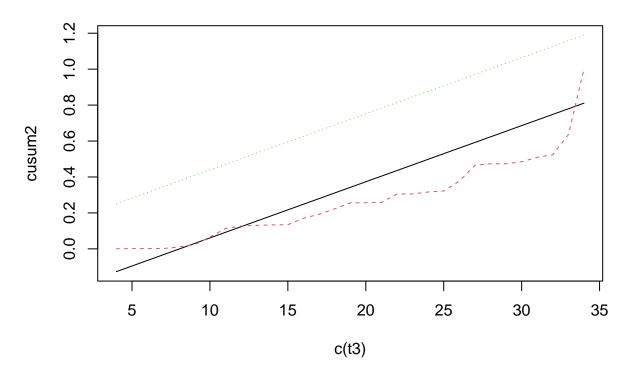
smax <- ((t2-k)/(n-k))+c0

#

vec2 <- c(smin, cumrr, smax)

cusum2 <- matrix(vec2, ncol = 3);

matplot(c(t3), cusum2, type ="1")
```



```
#sctest(y ~ x, type = "Chow", point =)
# pour R, la rupture est test?e non pas sur 1979-1980 mais sur 1979.
#Sur Eviews, la rupture est test?e sur 1980/
for(i in 9:30) {
print(sctest(y ~ x, type = "Chow", point = i) )
}
##
   Chow test
##
##
## data: y ~ x
## F = 4.6272, p-value = 0.005642
##
##
##
   Chow test
##
## data: y ~ x
## F = 4.6428, p-value = 0.005547
##
##
##
   Chow test
##
## data: y ~ x
## F = 4.6513, p-value = 0.005497
##
```

```
##
## Chow test
##
## data: y ~ x
## F = 5.796, p-value = 0.00168
##
##
## Chow test
##
## data: y ~ x
## F = 8.5115, p-value = 0.0001406
##
##
## Chow test
##
## data: y ~ x
## F = 10.764, p-value = 2.389e-05
##
##
## Chow test
##
## data: y ~ x
## F = 11.173, p-value = 1.772e-05
##
##
## Chow test
##
## data: y ~ x
## F = 11.398, p-value = 1.507e-05
##
##
## Chow test
##
## data: y ~ x
## F = 10.585, p-value = 2.728e-05
##
##
## Chow test
##
## data: y ~ x
## F = 10.003, p-value = 4.242e-05
##
##
## Chow test
## data: y ~ x
## F = 9.932, p-value = 4.48e-05
##
##
## Chow test
##
## data: y ~ x
## F = 8.8977, p-value = 0.0001021
##
```

```
##
## Chow test
##
## data: y ~ x
## F = 8.902, p-value = 0.0001017
##
##
## Chow test
##
## data: y ~ x
## F = 9.1596, p-value = 8.247e-05
##
##
## Chow test
##
## data: y ~ x
## F = 8.2241, p-value = 0.0001793
##
##
## Chow test
##
## data: y ~ x
## F = 8.8558, p-value = 0.0001056
##
##
## Chow test
##
## data: y ~ x
## F = 9.8364, p-value = 4.825e-05
##
##
## Chow test
##
## data: y ~ x
## F = 11.003, p-value = 2.004e-05
##
##
## Chow test
##
## data: y ~ x
## F = 9.5052, p-value = 6.256e-05
##
##
## Chow test
##
## data: y ~ x
## F = 7.0616, p-value = 0.0005013
##
##
## Chow test
##
## data: y ~ x
## F = 6.9945, p-value = 0.0005331
##
```

```
##
## Chow test
##
## data: y ~ x
## F = 7.5019, p-value = 0.0003366
n=length(Transport_France2019$Qdieselcamion)
P1 <- replicate(35, 0)
P1[1:16] <- Transport_France2019$Pdiesel[1:16]/Transport_France2019$CPI[1:16]
P2 <- replicate(35, 0)
P2[17:35] <- Transport_France2019$Pdiese1[17:35]/Transport_France2019$CPI[17:35]
vec <- c(P1,P2, Transport_France2019$"PIB en volume (en milliards d'euros 2014)",Transport_France2019$Q
X <- matrix( vec, ncol=4)</pre>
Y=matrix(Transport_France2019$Qdieselcamion,n,1)
q=ncol(Y);
k=ncol(X);
K=k+1
y=Y
X=X
nobs=cbind(1:n)
OLS=lm(formula = y ~ x)
summary(OLS)
##
## Call:
## lm(formula = y \sim x)
## Residuals:
       \mathtt{Min}
                 1Q
                      Median
                                    3Q
                                            Max
## -2393.64 -264.17
                        46.23
                               364.36 1468.79
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.174e+02 1.670e+03 0.070 0.94445
              -2.946e+05 1.410e+05 -2.089 0.04528 *
## x1
              -3.651e+05 1.203e+05 -3.034 0.00495 **
## x2
## x3
               4.871e+00 1.525e+00
                                      3.195 0.00328 **
               3.735e+01 5.607e+00 6.661 2.23e-07 ***
## x4
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 718.3 on 30 degrees of freedom
## Multiple R-squared: 0.9496, Adjusted R-squared: 0.9429
## F-statistic: 141.4 on 4 and 30 DF, p-value: < 2.2e-16
names(OLS)
```

"effects"

"rank"

[1] "coefficients" "residuals"

```
## [5] "fitted.values" "assign"
                                        "qr"
                                                        "df.residual"
## [9] "xlevels"
                       "call"
                                        "terms"
                                                        "model"
xc = cbind(1,x)
bhat = OLS$coefficients
yf = xc %*% bhat
res = y - yf
scr = t(res) %*% res
d1 = t(res) %*% res
d2 = t(res[2:n]-res[1:n-1]) %*% (res[2:n]-res[1:n-1])
dw = d2/d1
print (dw)
             [,1]
## [1,] 0.5232029
```