Nonparametric Audio Signal Modeling

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Set Up

Functions

```
load libraries = function(){
  library(tuneR)
  library(tidyverse)
  library(broom)
  library(ggpubr)
  library(fields)
# audio functions
signal_load_process = function(path, start, stop){
  # loads signal from provided path of an audio file
  # start is starting point of song in seconds
  # stop is ending point inn seconds
  # converts stereo signal to mono and normalizes it
  audio = readWave(path, from = start, to = stop, units = 'seconds')
  ymono = mono(audio, which = 'both')
  ymono = normalize(ymono, unit = '1')
  str(ymono)
  return(ymono)
signal_subsamples = function(y, samp_length){
  # extracts signal array and splits it evenly by
  # the length you provide
  s = y@left
  ysplit = seq_along(s)
  samples = split(s, ceiling(ysplit/samp_length))
  return(samples)
}
# data functions
data_signal = function(y){
  # creates dataset from signal consisting of
  \# y = sampled \ signal, \ n = discrete \ time \ index
  df = tibble(n=1:length(y), y=y)
data_every_n = function(n, y){
  # creates new dataset that takes every nth
  # data point in the signal dataset
 n1 = n[seq(1, length(n), 5)]
```

```
y1 = y[seq(1, length(y), 5)]
  df = tibble(n=n1, y=y1)
 df %>% print()
 return(df)
}
# plotting predictions and confidence bands
df preds loess = function(fit, df pred, conf lev){
 # create dataset to plot the curve including values for
  # predictions, confidence bands, standard errors, original signal
  # fit: fitted curve
  # data: df for predictions
  # conf lev: confidence level for intervals
  alpha = 1-conf_lev
  confidence = 1 - (alpha/2)
  crit_val = qnorm(p = confidence)
 n = df_pred$n; y = df_pred$y
 pse = predict(fit, df_pred, se = T)
 p = pse$fit
  s = pse$se.fit
 lb = p - crit_val * s
 ub = p + crit_val * s
  df = data.frame(y=y, n=n, pred=p, se=s, low_cb=lb, upp_cb=ub)
 df = as_tibble(df)
 return(df)
}
df_preds_spline = function(fit, df_pred, conf_lev){
  # create dataset to plot the spline including values for
  # predictions, confidence bands, standard errors, original signal
  # fit: fitted curve
  # data: df for predictions
  # conf_lev: confidence level for intervals
  alpha = 1-conf_lev
  confidence = 1 - (alpha/2)
  crit_val = qnorm(p = confidence)
 n = df_pred$n; y = df_pred$y
 p = predict(fit, x = n)
  s = predictSE(fit, x = n)
  lb = p - crit_val * s
 ub = p + crit_val * s
  df = data.frame(y=y, n=n, pred=p, se=s, low_cb=lb, upp_cb=ub)
  df = as tibble(df)
 return(df)
curve_plot = function(df, type = c('loess', 'tp'), sub = ''){
  if (type == 'loess'){
   title = 'Loess Regression Curve'
  }else{
   title = 'Thin-Plate Smoothing Spline'
  df %>% ggplot(aes(x=n)) + geom_point(aes(y=y), color='cyan') +
```

```
geom_line(aes(y=pred), color='red') +
    geom_line(aes(y=low_cb), color='red', lty=2) +
    geom_line(aes(y=upp_cb), color='red', lty=2) +
    ggtitle(title, subtitle = sub) + theme_classic() +
    labs(x='Discrete Time Index', y='Amplitude')
curve_only_plot = function(df, type = c('loess', 'tp'), sub = ''){
  if (type == 'loess'){
   title = 'Loess Regression Curve'
  }else{
   title = 'Thin-Plate Smoothing Spline'
  df %>% ggplot(aes(x=n)) +
    geom_line(aes(y=pred), color='blue') +
    geom_line(aes(y=low_cb), color='red', lty=2) +
   geom_line(aes(y=upp_cb), color='red', lty=2) +
    ggtitle(title, subtitle = sub) + theme_classic() +
   labs(x='Discrete Time Index', y='Amplitude')
```

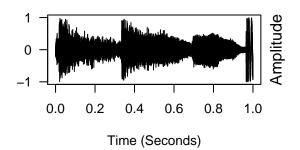
Load Libraries and Audio

```
load_libraries()
xn = signal_load_process('Closer.wav', start = 70, stop = 71)
## Formal class 'Wave' [package "tuneR"] with 6 slots
                 : num [1:44100] -0.0648 -0.0642 -0.0533 -0.0532 -0.1051 ...
     ..@ left
##
     ..@ right
                 : num(0)
##
    ..@ stereo
                : logi FALSE
    ..@ samp.rate: int 44100
##
##
    ..@ bit
                : num 32
##
     ..@ pcm
                 : logi TRUE
```

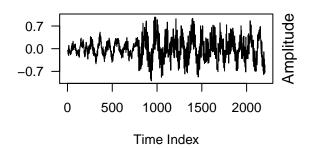
Explore and Create Data

Explore Audio

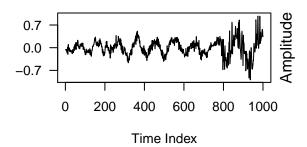
Waveform of Signal



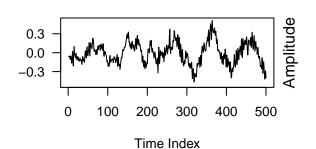
Subsample of Signal: n = 2205



Subsample of Signal: n = 1000



Subsample of Signal: n = 500



par(mfrow=c(1,1))

Subsamples of Signal

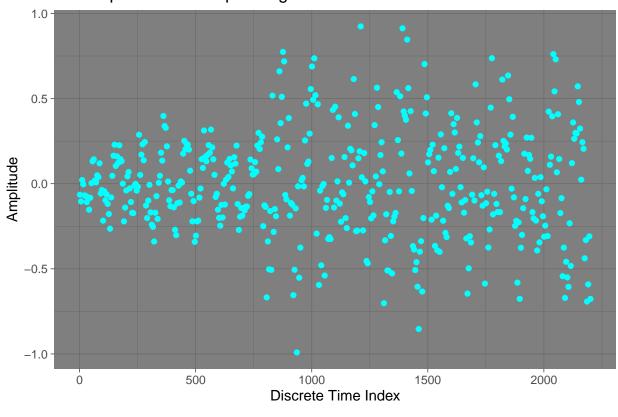
```
xsamps = signal_subsamples(xn, 2205)
xdf_samps = data_signal(xsamps[[1]])
xdf = data_every_n(xdf_samps$n, xdf_samps$y)
## # A tibble: 441 x 2
##
          n
                    У
##
      <int>
                <dbl>
##
    1
          1 -0.0648
##
    2
          6 -0.104
##
    3
         11 0.0215
##
         16 -0.00360
         21 -0.0685
##
    5
##
    6
         26 -0.0716
         31 -0.108
##
    7
##
    8
         36 -0.0806
         41 -0.153
##
    9
         46 -0.0826
##
   10
         with 431 more rows
```

Scatterplot

```
xdf %>% ggplot(aes(x=n, y=y)) + geom_point(color='cyan') +
    ggtitle('Scatterplot of Subsampled Signal') +
    labs(x='Discrete Time Index', y='Amplitude') +
```

theme_dark()

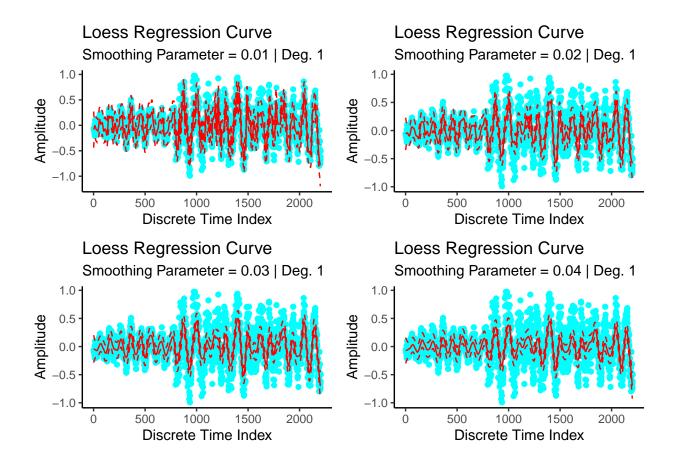
Scatterplot of Subsampled Signal



Loess Regression

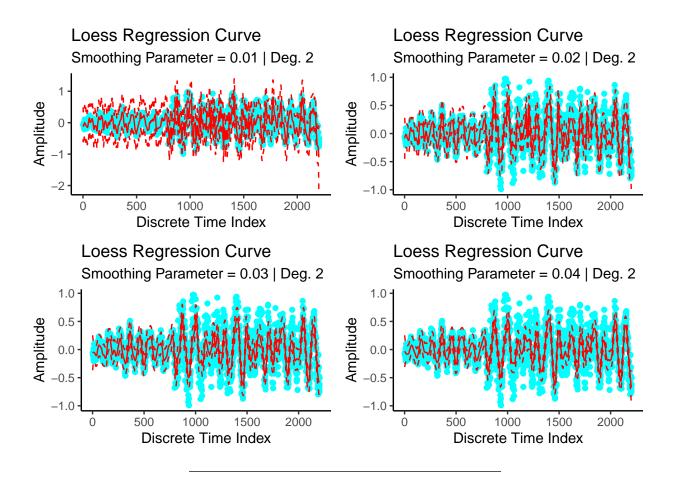
Degree 1

```
sm1 = list()
dl1 = list()
p1 = list()
subs1 = list('Smoothing Parameter = 0.01 | Deg. 1',
             'Smoothing Parameter = 0.02 | Deg. 1',
             'Smoothing Parameter = 0.03 | Deg. 1',
             'Smoothing Parameter = 0.04 | Deg. 1')
for (i in 1:4){
  span = i/100
  s = loess(y ~ n, data = xdf, span = span, degree = 1, family = 'gaussian')
  sm1 = append(sm1, list(s))
  d = df_preds_loess(s, df_pred = xdf_samps, conf_lev = 0.95)
  dl1 = append(dl1, list(d))
  p = curve_plot(d, type = 'loess', sub = subs1[[i]])
  p1 = append(p1, list(p))
ggarrange(p1[[1]], p1[[2]], p1[[3]], p1[[4]], nrow = 2, ncol = 2)
```



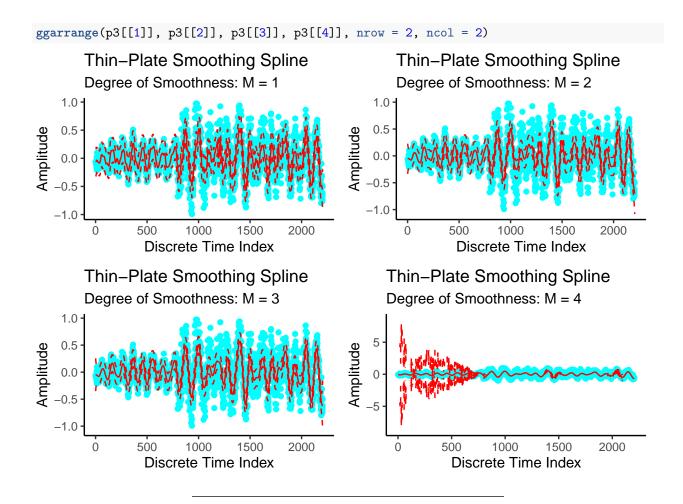
Degree 2

```
sm2 = list()
d12 = list()
p2 = list()
subs2 = list('Smoothing Parameter = 0.01 | Deg. 2',
             'Smoothing Parameter = 0.02 | Deg. 2',
             'Smoothing Parameter = 0.03 | Deg. 2',
             'Smoothing Parameter = 0.04 | Deg. 2')
for (i in 1:4){
  span = i/100
  s = loess(y ~ n, data = xdf, span = span, degree = 2, family = 'gaussian')
  sm2 = append(sm2, list(s))
  d = df_preds_loess(s, df_pred = xdf_samps, conf_lev = 0.95)
  d12 = append(d12, list(d))
  p = curve_plot(d, type = 'loess', sub = subs2[[i]])
  p2 = append(p2, list(p))
ggarrange(p2[[1]], p2[[2]], p2[[3]], p2[[4]], nrow = 2, ncol = 2)
```



Thin-Plate Splines

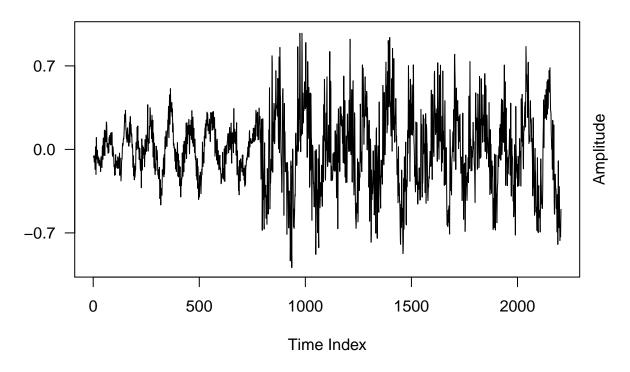
```
attach(xdf)
tpm = list()
dtp = list()
p3 = list()
subs3 = list('Degree of Smoothness: M = 1',
             'Degree of Smoothness: M = 2',
             'Degree of Smoothness: M = 3',
             'Degree of Smoothness: M = 4')
for (i in 1:4){
 m = Tps(x=n, Y=y, m=(i))
  tpm = append(tpm, list(m))
  d = df_preds_spline(m, xdf_samps, 0.95)
  dtp = append(dtp, list(d))
  p = curve_plot(dtp[[i]], type = 'tp', sub = subs3[[i]])
  p3 = append(p3, list(p))
}
## Warning:
## Grid searches over lambda (nugget and sill variances) with minima at the endpoints:
     (GCV) Generalized Cross-Validation
##
     minimum at right endpoint lambda = 8.881789e-16 (eff. df= 55.30496)
```



Comparisons

Real Subsample of Signal

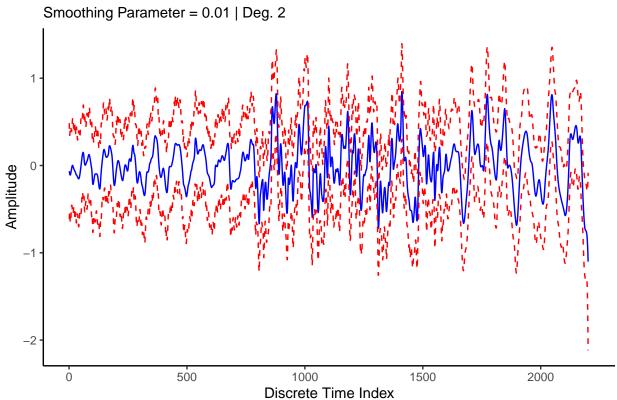
Subsample of Signal: n = 2205



Loess Degree 2

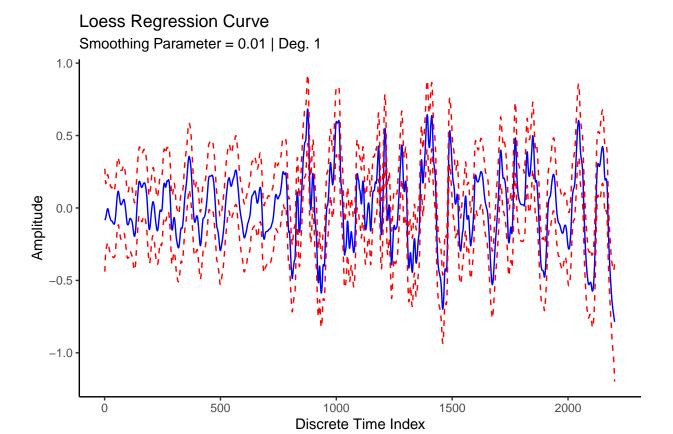
```
curve_only_plot(dl2[[1]], type = 'loess', sub = subs2[[1]])
```

Loess Regression Curve



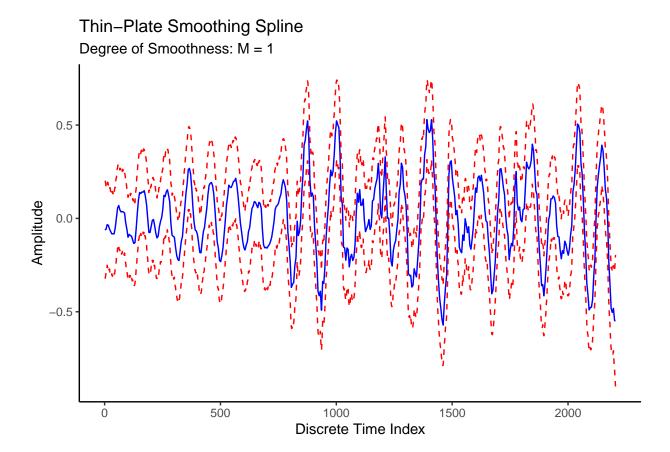
Loess Degree 1

```
curve_only_plot(dl1[[1]], type = 'loess', sub = subs1[[1]])
```



Thin-Plate

```
curve_only_plot(dtp[[1]], type = 'tp', sub = subs3[[1]])
```



Best Result

Based on the actual behavior of the signal, the quadratic loess regression curve with smoothing parameter 0.01 seemed to give the most accurate depiction of the actual signal.