# Introduction to change point detection: lab worksheet

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# Question 1: Creating a test statistic to estimate change points

- (a) Without using pre-existing libraries, write a function to calculate the CUSUM statistic vector  $\{\mathcal{T}_{0,k,n}\}_{k=1}^{n-1}$  for a change point in the mean.
- (b) Without using pre-existing libraries, write a function to calculate the MOSUM statistic vector  $\{\mathcal{T}_G(k)\}_{k=G}^{n-G}$  for a change point in the mean, for a given bandwidth G.
- (c) To set a threshold for declaring changes, we need to know the noise level  $\sigma$ . In reality, this is unknown; how could we estimate it?
  - *Hint*: the sample standard deviation would be positively biased by the changes. Medians are more robust than means: how can we transform the data to remove most of the effect of the change points, before using a median-like analogue of the standard deviation? If in doubt, Google is your friend.
- (d) Implement your chosen method of estimating  $\sigma$ , and add the functionality to your CUSUM and MOSUM functions so that the test statistic calculations in parts (a) and (b) are scaled by your estimate  $\widehat{\sigma}$
- (e) (Harder) To find multiple change points with the MOSUM approach, we can calculate all  $\hat{\theta}$  that satisfy:

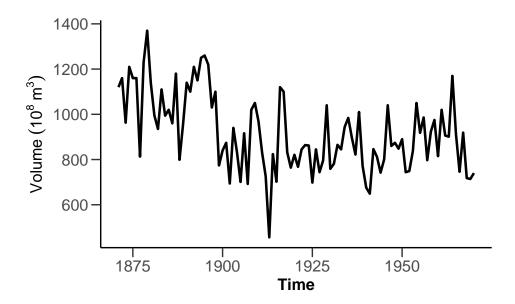
$$\mathcal{T}_{\ell}(G, \widehat{\theta}) > D$$
 and  $\widehat{\theta} = \operatorname{argmax}_{k: |k-\widehat{\theta}| \leq \eta G} \mathcal{T}_{G}(k)$ .

for some window fraction  $\eta \in (0,1)$  and a threshold D. That is,  $\widehat{\theta}$  is declared a change point if it is a local maximiser of  $\mathcal{T}_G(k)$  over a sufficiently large interval of size  $\eta G$ , at which the threshold D is exceeded.

Extend your MOSUM function to return as output the change point estimators satisfying this criterion.

#### Question 2: Nile annual river flow

In this question you will use the code you've written so far to analyse data collected on the river Nile. Data in the file  ${\tt nile\_volume.txt}$  records measurements of the annual volume (in units  $10^8m^3$ ) of discharge from the Nile River at Aswan for the years 1871 to 1970. The measurements are of meteorological importance as evidence of a possible abrupt change in the rainfall levels around the turn of the 20th century.



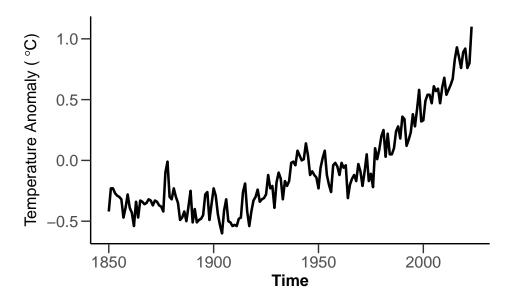
- (a) Load the data into R from the nile\_volume.txt file, and plot it. By eye, where does it look like there could be a change in mean?
- (b) Calculate the CUSUM statistic for the Nile volume data. Using the threshold  $D = \sqrt{2 \log(n)}$ , perform a test to decide if there is a change point. If there is one, where is it?
- (c) Calculate the MOSUM statistic for the Nile volume data using a bandwidth G = 25. Using the threshold D = 3.4, compute the change point estimators as in q1 (e). Does this agree with your answer from part (b)?
- (d) Now use the MOSUM method with bandwidth G = 40. How does your answer differ from part (c)?
- (e) The estimated mean signal  $\hat{f}_t$  can be calculated using sample means of the segments defied by the estimated change points. For the CUSUM method, add the estimated  $\hat{f}_t$  to your plot from part (a).

### Question 3: Using the mosum and changepoint packages

- (a) Load the changepoint and mosum R packages into your R workspace. Using the help files, get aquainted with the cpt.mean and mosum functions.
- (b) Using the default settings, use the cpt.mean function on the Nile data set. Do the results agree with your answer from question 1(a)?
- (c) By altering the input parameters, apply the PELT algorithm using the cpt.mean function on the Nile data set. Is the answer the same as part (b)?
- (d) By investigating the output returned by the cpt.mean function, what was the value of the penalty function used?
- (e) Using the bandwidth G = 25, use the mosum function on the Nile data set. Do the results agree with your answer from question 1(b)?
- (f) By investigating the output returned by the mosum function, what are the p-value(s) of the detected change points?
  - Python users: you can use the ruptures function Pelt and the mosum functions for this question. See https://centre-borelli.github.io/ruptures-docs/ and https://pypi.org/project/mosum/for help.

# Question 4: Global yearly mean sea temperature anomalies

Global mean surface temperature series help enable the monitoring of global warming. The warming can be quantified by, for example, a change from a base period used as reference. The Hadley Centre/Climatic Research Unit (HadCRUT) surface temperature data set provides the annual global temperature anomalies from 1850-2023, where anomalies are calculated relative to the 1961–1990 period. The data are shown below, which can be seen to exhibit a gradual upward trend.



- (a) Load the data into R from the temp\_anomalies.txt file, and plot it. Using e.g. the lm function, fit two linear trends to the data: one from 1850 to 1963, and one from 1964 to 2023, and add these lines to the plot, as well as a vertical line through year 1963.
- (b) Using your own functions, and functions from the changepoint and mosum packages, fit 4 models to the data set;
  - 1. a constant mean model with no change points,
  - 2. a mean change point model,
  - 3. a linear trend model with no change points.
  - 4. a linear trend change point model.

Looking at the data, are there any other models that might be appropriate?

*Hint*: if you are struggling, the EnvCpt R package has everything you need.

Python users: you can use the Pelt function and change the model parameter.

(c) Pick the "best" change point model from your candidates computed in part (b), and justify your choice. Add the estimated mean/trend function onto a plot of the data.

*Hint*: the Akiake information criterion (AIC) is given by AIC =  $2k - 2\log(\hat{L})$ , where k is the number of parameters of the model, and  $\hat{L}$  is the maximised value of the likelihood function for the model.

## Question 5: Multivariate data

(a) Suppose we want to find mean change points in a time series  $\{X_t\}_{t=1}^n$ , where  $X_t = (X_{1t}, \dots, X_{pt})^\mathsf{T}$  is a p-dimensional vector. One approach is to calculate a test statistic for each variable, and then combine

the results across variables to give a single test statistic. For example, for the *i*-th variable of  $\{X_t\}_{t=1}^n$ , denoted  $\{X_{it}\}_{t=1}^n$ , the MOSUM test statistic is given by

$$\mathcal{T}_G(k,i) = \frac{1}{\sqrt{2G}} \left( \sum_{t=k+1}^{k+G} X_{it} - \sum_{t=k-G+1}^{k} X_{it} \right).$$

Then, the MOSUM statistic  $\mathcal{T}_G(k)$  for a change point in  $\{X_t\}_{t=1}^n$  can be computed by aggregating the  $\{\mathcal{T}_G(k,i)\}_{i=1}^p$  using some aggregating function f, so that

$$\mathcal{T}_G(k) = f(\mathcal{T}_G(k,1), \dots, \mathcal{T}_G(k,p)).$$

What possibilities could we use for the aggregating function f?

- (b) Using your aggregating function f, write a function to compute the MOSUM or CUSUM statistic for a change in the mean vector of a multivariate time series.
- (c) Simulate a data set of length n=500 and dimension p=40 as follows. Generate  $\{X_t\}_{t=1}^{250}$  from the standard p-dimensional normal distribution, and generate  $\{X_t\}_{t=251}^{500}$  from the p-dimensional normal distribution with identity covariance matrix, and mean vector given by  $\mu = (1/\sqrt{p}, \dots, 1/\sqrt{p})^{\mathsf{T}}$ . Use the method from part (b) to compute the test statistic for a change in mean. Is the change point easy to see?
- (d) Re-do part (c), but generate  $\{X_t\}_{t=251}^{500}$  from the *p*-dimensional normal distribution with identity covariance matrix, and mean vector given mu with first 3 entries equal to 2, and last p-3 entries equal to 0. Is the change point easy to see?