

# The Window of Change



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## Changepoints and regime shifts ...

Standard framework for **changepoints**: observations  $y_1, \dots, y_n$  follow the model  $f(y, \theta)$ , with parameters

$$\begin{aligned}\theta_i &= \theta_L \text{ if } i \leq a, \\ \theta_i &= \theta_R \text{ if } i \geq a + 1.\end{aligned}$$

There's a large literature on

- (i) testing whether the world has been constant (no changepoint);
- (ii) spotting the changepoint  $a$  (if it's there);
- (iii) constructing confidence statements;
- (iv) assessing how different  $\theta_R$  is from  $\theta_L$ .

Applications abound & multiply – and find uses not merely for 'change of visible level' but for inner-working parameters (has a regression coefficient  $\beta_4$  for education level changed over time, in relation to democracy?).

... but changes often take time

This talk: **Window of Change** ( $a, b$ ):

$$\theta_i = \theta_L \text{ if } i \leq a,$$

$$\theta_i = \theta_R \text{ if } i \geq b,$$

$$\theta_i = \text{in between if } a < i < b.$$

Need to **assume something** for the transition window from **Equilibrium A** to **Equilibrium B**. A natural start is

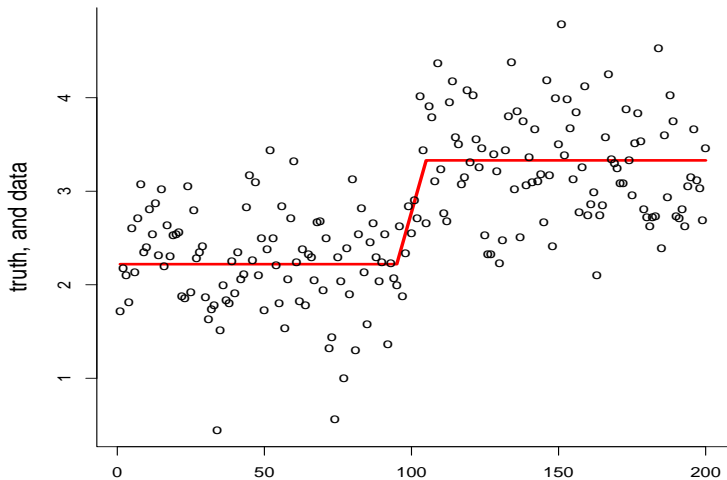
$$\theta_i = \theta_L + \frac{i - a}{b - a}(\theta_R - \theta_L) \quad \text{for } i = a, \dots, b.$$

How to estimate and reach inference for the **Window of Change** ( $a, b$ ), along with  $\theta_L, \theta_R$ ?

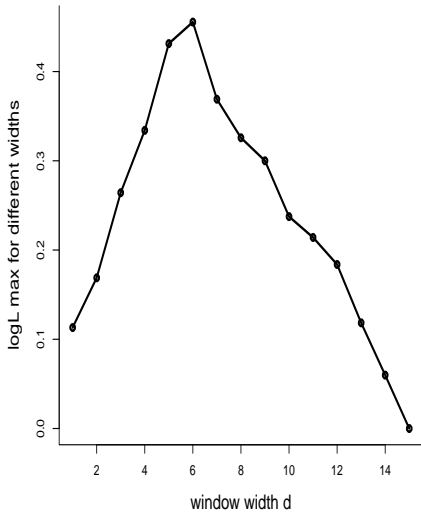
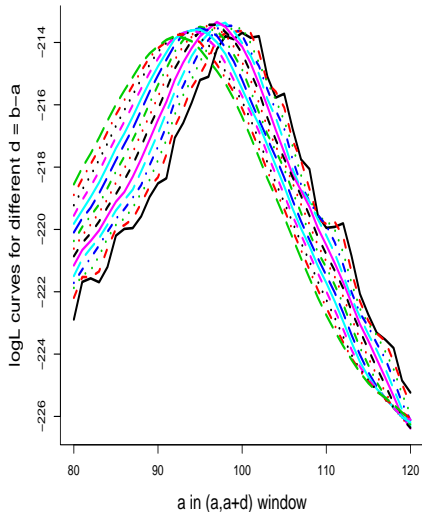
Answers: (i) log-likelihood analyses; (ii) Bayes with MCMC.

## Story I: A stylised illustration: transition from A to B

200 normal observations; going from  $\theta_L = 2.22$  to  $\theta_R = 3.33$  over time window  $[a, b] = [95, 105]$ . Can we estimate this from data?



Left: log-likelihoods  $\ell(a, a + d)$  in  $a$ , for fixed widths  $d = 1, \dots, 15$ . Right: log-lik maxima  $\ell_{\max}(d)$  over  $d$ .



Then **Bayes MCMC**. With a well-chosen prior  $\pi(a)\pi(d)\pi(\theta_L, \theta_R)$ , in terms of  $d = b - a$ , the posterior is

$$\pi(a, b, \theta_L, \theta_R \mid \text{data}) \propto \pi(a)\pi(d)\pi(\theta_L, \theta_R) \exp\{\ell(a, b, \theta_L, \theta_R)\}$$

in terms of log-likelihood  $\ell(\text{parameter})$ .

I construct a Markov chain of outcomes  $(\theta_L, \theta_R, a, b)$  in my computer. From **old** =  $(\theta_L, \theta_R, a, b)$  I propose **next** =  $(\theta'_L, \theta'_R, a', b')$ , with a gentle symmetric push for  $\theta_L, \theta_R$ , whereas  $a' - a = -1, 0, 1$  and  $b' - b = -1, 0, 1$  with equal probabilities  $(1/3)^2 = 1/9$ . I accept with probability

$$\begin{aligned} \text{pr} &= \min(1, \exp(\delta)), \\ \delta &= \ell(\text{next}) - \ell(\text{old}) + \log \pi(\text{next}) - \log \pi(\text{old}). \end{aligned}$$

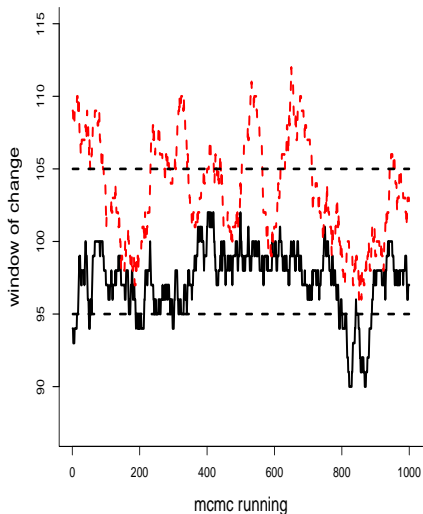
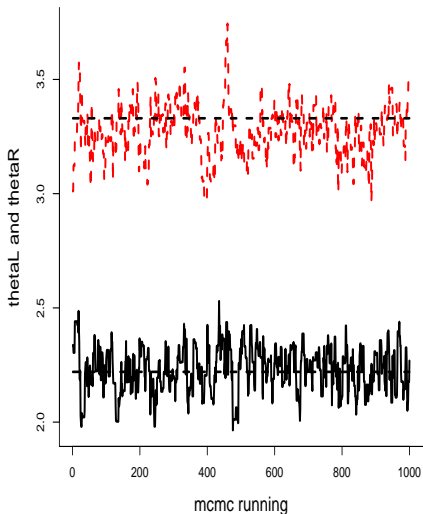
Book-keeping care to ensure **pr** = 0 for non-windows, etc. Then by MCMC theory this produces simulations from the genuine posterior distribution.

It's good clean fun to see **the (a, b) chain** on your computer screen.

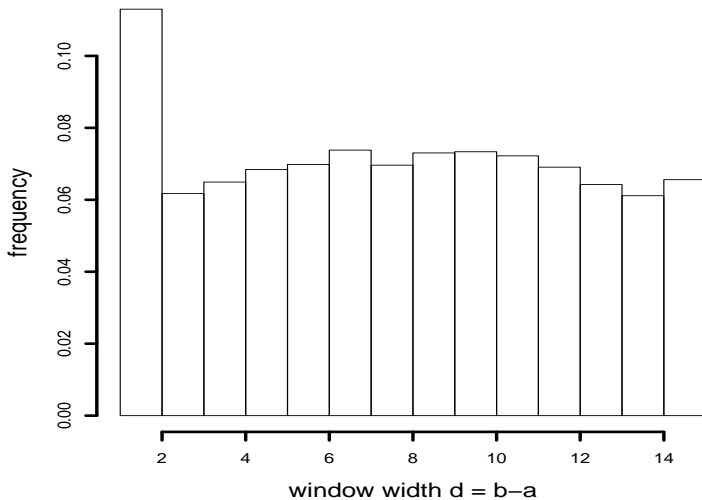
Bayes MCMC, with flat prior for  $d = b - a$  on  $1, \dots, 15$ .

It works – but easier for  $\theta_L, \theta_R$  than for  $[a, b]$ . Difficult to get window right, even with good data. Prior for  $d$  matters. Here

$d_{\text{true}} = 10$ .



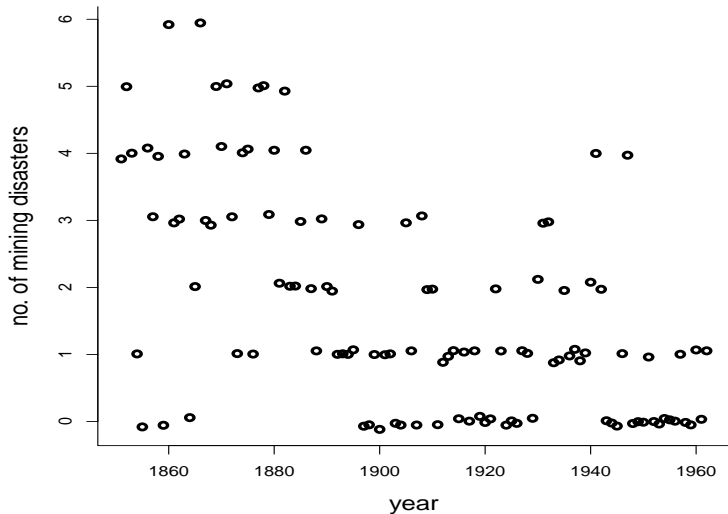
Flat prior for window width  $d$  on  $1, \dots, 15$ ; long MCMC to read off  $\pi(d \mid \text{data})$ . Here  $d_{\text{true}} = 10$ . Prior not easily 'washed out by data'.





## Story II: British mining disasters

No. of disasters, from 1851 to 1962 (from Jarrett, 1979). The Poisson level has diminished from  $\theta_L \approx 3.0$  to  $\theta_R \approx 1.0$  ... but about when, and how quickly?



## Poisson model with window of change

I take  $y_i \sim \text{Pois}(\theta_i)$  with

$$\theta_i = \theta_L \text{ if } i \leq a,$$

$$\theta_i = \theta_R \text{ if } i \geq b,$$

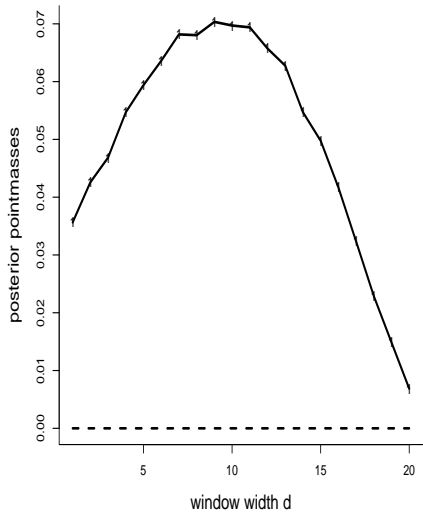
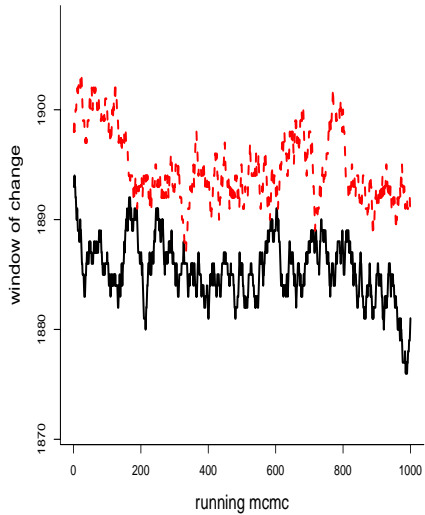
$$\theta_i = \text{linear in between} \text{ if } a < i < b.$$

Several papers in the literature have aimed for **simple changepoint**, i.e.  $b = a + 1$ .

Their story: change from **1891** to **1892** !, from  $\theta_L \doteq 3.25$  to  $\theta_R \doteq 0.88$ .

My story: gradual change from **1889** to **1898** !

I'm using Bayes with prior  $1/d$  on  $1, \dots, 20$ , and build my MCMC running in  $(\theta_L, \theta_R, a, b)$ , reading off  $\pi(d \mid \text{data})$ , position of window  $[a, a + d]$ , etc. – Reasonably similar frequentist results, with log-likelihoods, but **prior on  $d$  matters**.



# Story III: When (and how quickly) did Author B take over for Author A?

Tirant lo Blanch is the world's first novel, written in Catalan, c. 1460–1464, and published in València in 1490. Somewhere in the sequence of 487 chapters, Martí Joan de Galba took over for Joanot Martorell. But where, precisely? And did the change take place instantly (from chapter 371 to chapter 372, claim Cunen, Hermansen, Hjort, JSPI 2018), or did it take a few chapters?

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## What can we look for?

What can we **measure and monitor**, chapter for chapter?

From **Chapter XII**:

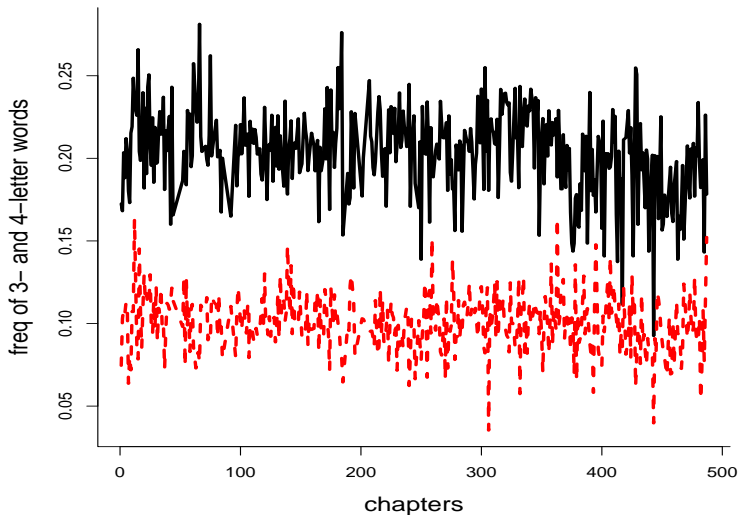
*E lo rey dix que era molt content. E en la scura nit lo virtuós hermità mudà's les vestidures que tenia aparellades de moro, e per la porta falsa del castell isqué molt secretament, que per negú no fonch vist ne conegut, e posà's dins lo camp dels moros.*

From **Chapter CCCCLXXII**:

*¡Despullau a mi daurades robes y dels palaus leven les riques porpres! ¡Cobriu-me prest de hun aspre scilici, visten-se tots de fort y negra màrrega, sonen ensemps les campanes sens orde, dolga's tothom de tanta pèrdua, per a rahonar la qual ma lengua és feta scaça!*

See **Céline Cunen's talk**: We go for **word lengths** and their proportions, chapter by chapter.

We compute and examine frequencies  $\hat{p}_1, \dots, \hat{p}_{10}$  through chapters 1, ..., 487. Here 3-letter and 4-letter words. Where is the change?



## The multinomial-Dirichlet window of change model for word lengths

In chapter  $i$  there are  $m_i$  words, sorted into  $y_{i,1}, \dots, y_{i,10}$  of lengths  $1, \dots, 10$ . My model takes

$$\begin{aligned} f_i &= \int \frac{m_i!}{y_{i,1}! \cdots y_{i,10}!} p_1^{y_{i,1}} \cdots p_{10}^{y_{i,10}} \text{Dir}(dp) \\ &= \frac{m_i!}{y_{i,1}! \cdots y_{i,10}!} \frac{\Gamma(k_i p_{i,0,1} + y_{i,1}) \cdots \Gamma(k_i p_{i,0,10} + y_{i,10})}{\Gamma(k_i p_{i,0,1}) \cdots \Gamma(k_i p_{i,0,10})} \frac{\Gamma(k_i)}{\Gamma(k_i + m_i)} \end{aligned}$$

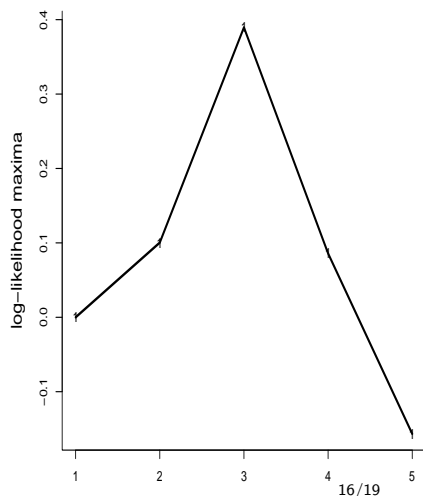
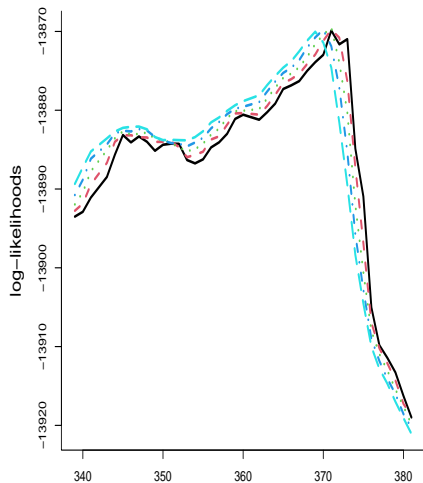
for chapter  $i$ , with the **Window of Change** setup:

$(k_L, p_L)$  up to  $a$ ;  $(k_R, p_R)$  after  $a + d$ ; linear interpolation inside  $(a, a + d)$ .

For given  $(a, a + d)$ , need to optimise over  $1 + 9 + 1 + 9 = 20$  parameters, and over the full dataset.

## Did it take ... 3 chapters to settle in?

So change sets in at c. [Chapter 371](#) of the 487. I'm checking chapter windows  $[a, a + 1], \dots, [a, a + 5]$ : but no,  $\hat{d} = 3$  is not significant.





## Remarks

- ♠ Estimation and inference for  $(a, b)$  window inherently more difficult than for a single changepoint. Methods are harder to construct; harder to analyse well; and precision is lower.
- ♠ There's room for [good Bayesian methods](#) (with MCMC), but prior for  $d = b - a$  is crucial, and matters more than for other components. Use context & knowledge. Also easy to read off [how much  \$\theta\$  has changed](#).
- ♠ In [Statistical Sightings of Better Angels](#), Céline and I did changepoint analysis for CoW battle deaths for 95 interstate wars, 1824 to 2004, and found  $\hat{a} = 1953$  (with a confidence curve). Wish to attempt Window of Change methods there too. I then need some more careful book-keeping code for dealing with non-equal time differences. See also Dennis Christensen's talk.

- ♠ Straight Bayesian flat priors for  $a$  and for  $d$  on  $1, \dots, d_{\max}$  make sense ... but have difficulties. Posterior will be tempted to push the windows to one of the two sides:

$$\pi(a, d \mid \text{data}) \approx \pi(a, d) \exp\{\ell_{\text{prof,max}}(a, d)\} \frac{1}{\sqrt{c(n-c)}},$$

with  $c$  the midpoint of  $(a, a + d)$ . So there are certain mathematical differences between non-Bayes and Bayes here; 'bigger than we are used to'.

- ♠ For what happens **inside the window of change** I've posited simple linear change from  $\theta_L$  to  $\theta_R$ . I think (a) this is ok, (b) other attempts at more sophistication might not change results much, as long as the window of change is not a very long one.

## (Some) references

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