

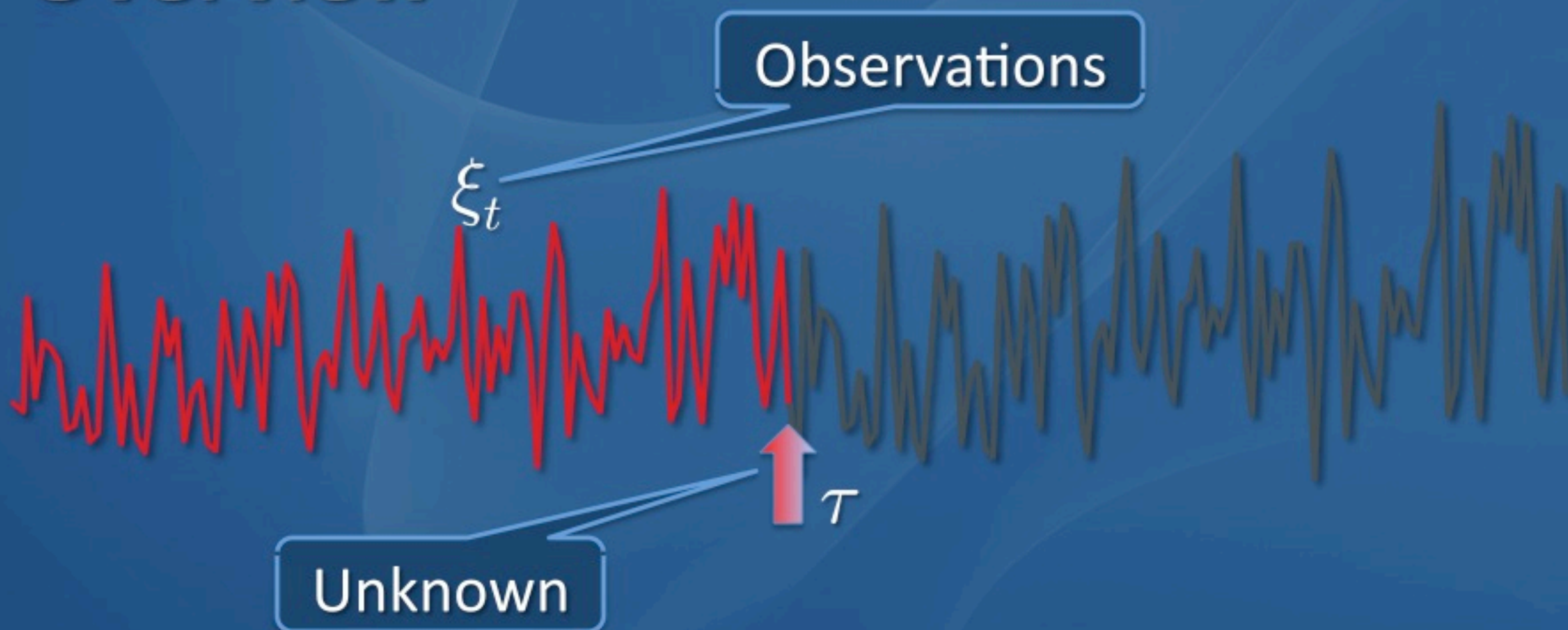
Sequential rate change detection in Poisson processes

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Outline

- Overview of sequential changepoint detection
- CUSUM test and Lorden's criterion
- The homogeneous Poisson disorder problem
 - CUSUM average run length
 - CUSUM optimality
- Non-homogeneous Poisson disorder problem and Epidemic surveillance

Overview



Using $\{\xi_t\}$ **detect τ as soon as possible**

Quality control

Systems monitoring

Remote sensing and GIS

Smart cameras – Human computer interaction

Image processing

⋮

Optical communications

Changepoint models for hazard functions

Occurrence of industrial accidents

Epidemic detection

Monitoring of link failures in computer networks

⋮



P_∞ : nominal measure

P_0 : alternative measure

P_τ : measure induced by the change

$E_\tau [.]$: corresponding expectation

Both P_0 and P_∞ are assumed known

We are interested in **sequential schemes**

With every new data point ξ_t we decide whether

- Stop and raise an alarm
- Continue sampling

For the decision at time t we use all the available information up to time t

$$\mathcal{F}_t = \sigma\{\xi_s : 0 \leq s \leq t\}$$

Sequential test \Leftrightarrow stopping time T adapted to the filtration $\{\mathcal{F}_t\}$

Detection delay

what is τ ?

$$\mathcal{J}(T) = \mathbb{E}_{\tau}[T - \tau | T > \tau]$$

False alarm

Average period between false alarms:

$$\mathbb{E}_{\infty}[T] \geq \gamma$$

False alarm probability:

$$\mathbb{P}(T \leq \tau) \leq \alpha$$

$$\min_T \mathcal{J}(T)$$

Changepoint mechanism

There is a mechanism that **decides when** to impose the change. This decision can be

- **Independent** from the observations $\{\xi_t\}$
- **Depend** on the observations $\{\xi_t\}$

If independent, then τ appears as random variable with prior

$$P(\tau = t) = \pi_t$$

If $\{\pi_t\}$ known then Bayesian formulation

Bayesian Formulation

Zero modified geometric

$$\pi_t = (1 - \varpi)p(1 - p)^{t-1}, t \geq 1; \quad \pi_0 = \varpi.$$

$$\min_T E[T - \tau | T > \tau]; \text{ s.t. } P(T \leq \tau) \leq \alpha$$

$$S_t = (S_{t-1} + 1) \frac{f_0(\xi_t)}{(1 - p)f_\infty(\xi_t)}, \quad S_0 = r$$

$$T_{\text{SR}} = \inf\{t : S_t \geq \nu\}$$

I.i.d.: Shiryaev (1963,1978), Poor (1998). **BM:** Shiryaev (1961). **Poisson:** Peskir and Shiryaev (2002), Dayanik (2005,2006,2008)

The mechanism decides without consulting the observations but $P(\tau = t) = \pi_t$ is **unknown**

$$\begin{aligned}\mathcal{J}_P(T) &= \sup_{\{\pi_t\}} E_\tau [T - \tau | T > \tau] \\ &= \sup_{t \geq 0} E_t [T - t | T > t]\end{aligned}$$

Pollak's (1985)
criterion

$$\begin{aligned}\inf_T \mathcal{J}_P(T) &= \inf_T \sup_{t \geq 0} E_t [T - t | T > t] \\ \text{s.t. } E_\infty[T] &\geq \gamma\end{aligned}$$

Pollak (1985) proposed the following stopping time known as Shiryaev-Roberts-Pollak

$$S_t = (S_{t-1} + 1) \frac{f_0(\xi_t)}{f_\infty(\xi_t)}, \quad S_0 \sim q(S)$$

$$T_{\text{SRP}} = \inf\{t : S_t \geq \nu\}$$

It is an
equalizer

$$\mathbb{E}_t[T_{\text{SRP}} - t | T_{\text{SRP}} > t] = \text{constant}$$

$$\mathcal{J}_P(T_{\text{SRP}}) - \inf_T \mathcal{J}_P(T) = o(1), \quad \text{as } \gamma \rightarrow \infty$$

Strict Optimality ?

In 1997 appears a proof that the SRP test is optimum (Annals of Statistics)

Yajun Mei (2006), shows that the proof is problematic

The conjecture remained unanswered until last year (2010): Tartakovsky and Polunchenko, produced a counterexample. After 25 years we can finally say that

The SRP test is NOT optimum

CUSUM and Lorden's criterion

The changepoint mechanism decides to impose the change by consulting the observations $\{\xi_t\}$ and possibly additional information.

In this case τ becomes a stopping time adapted to a larger filtration than $\{\mathcal{F}_t\}$.

$$\mathcal{J}_L(T) = \sup_{\tau} \mathbb{E}_{\tau}[T - \tau | T > \tau]$$

Lorden's (1971)
criterion

$$= \sup_{t \geq 0} \text{essup} \mathbb{E}_t[T - t | T > t, \mathcal{F}_t]$$

$$X_t = AX_{t-1} + BW_t$$

State space
model

$$\xi_t = CX_t + Dw_t$$

Measurements

Possible cause for change is for example “large” oscillations

$$\tau = \inf\{t : \|X_t\| \geq c\}$$

$$A \rightarrow A'$$

A bomb set to explode at a specific time

$$\inf_T \mathcal{J}_L(T) = \inf_T \sup_{t \geq 0} \text{ess sup } \mathbf{E}_t[T - t | T > t, \mathcal{F}_t]$$

$$\mathbf{E}_\infty[T] \geq \gamma$$

The optimum scheme: CUSUM

$$u_t = \log \left(\frac{d\mathbf{P}_0}{d\mathbf{P}_\infty}(\mathcal{F}_t) \right)$$

I.i.d.: Moustakides (1986)
Ritov (1990), Poor (1998)

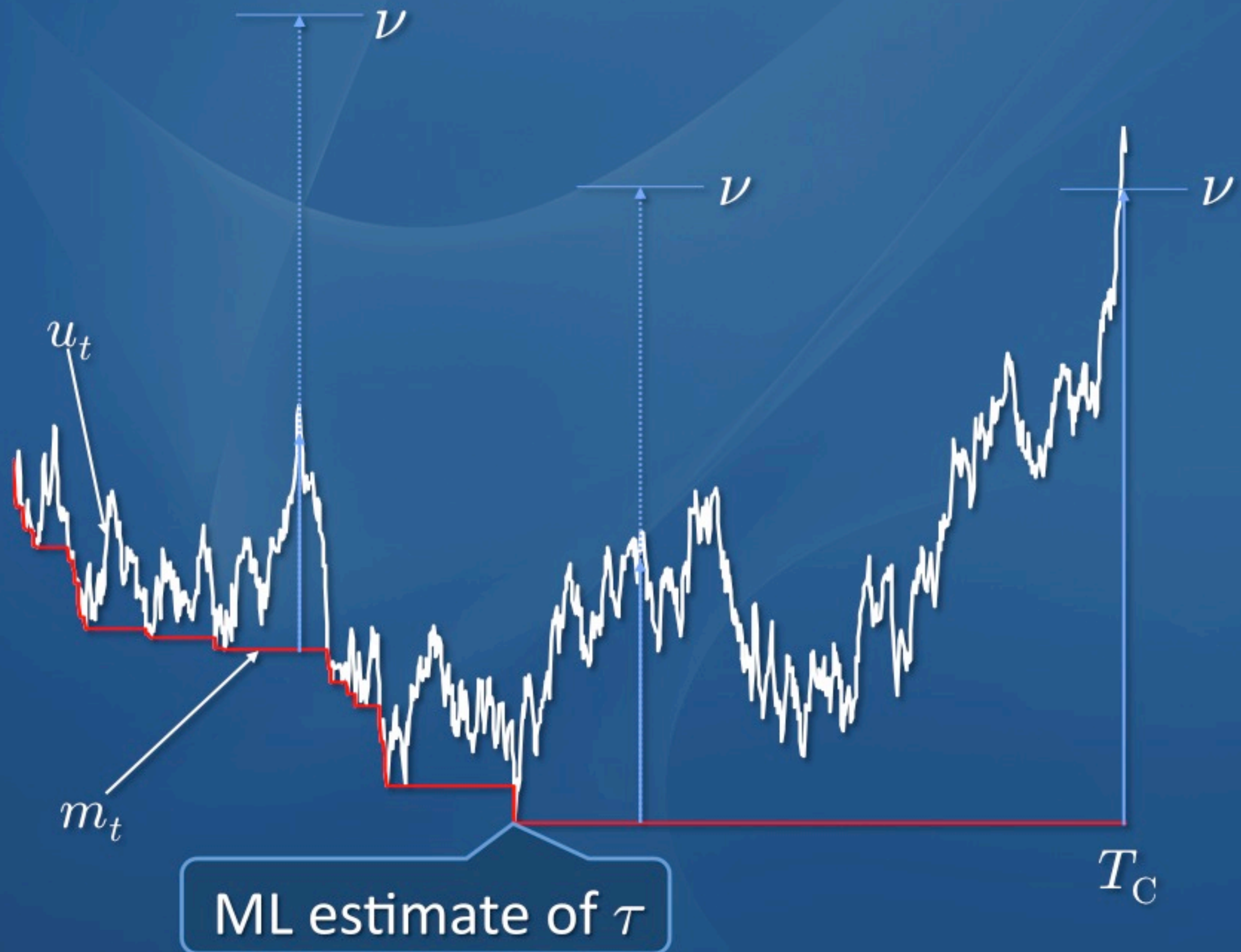
$$m_t = \inf_{0 \leq s \leq t} u_s$$

BM: Shiryaev (1996),
Beibel (1996)

$$y_t = u_t - m_t \geq 0$$

Ito: Moustakides (2004)

$$T_C = \inf\{t : y_t \geq \nu\}$$



Homogeneous Poisson disorder

Let $\{\mathcal{N}_t\}$ denote a homogeneous Poisson process with rate λ satisfying

$$\lambda = \begin{cases} \lambda_\infty & \text{for } t \leq \tau \\ \lambda_0 & \text{for } t > \tau. \end{cases}$$

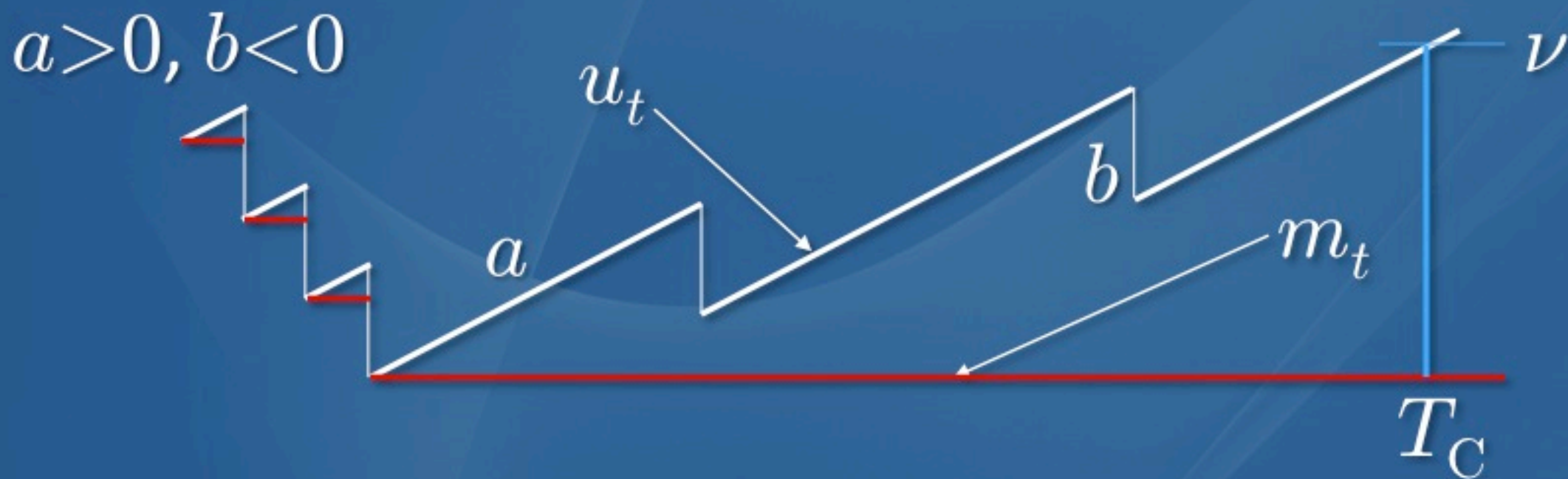
$$u_t = (\lambda_\infty - \lambda_0)t + \log \frac{\lambda_0}{\lambda_\infty} \mathcal{N}_t$$

$$u_t = a t + b \mathcal{N}_t$$

a, b opposite signs

Show optimality of CUSUM

First step compute ARL: $E_0[T_C]$ and $E_\infty[T_C]$



We are looking for $f(y) = E[T_C | y_0 = y]$

$$E[f(y_{T_C})] - f(y_0) =$$

$$E \left[\int_0^{T_C} \underbrace{\{a f'(y_{t-}) + \lambda [f(y_{t-} + b) - f(y_{t-})]\}}_{=-1} dt \right]$$

$$af'(y) + \lambda[f(y+b) - f(y)] = -1$$

$$f(\nu) = 0; \quad f(y) = f(0) \quad y \leq 0$$

$$f(y) = \frac{1}{\lambda} \left\{ \sum_{n=0}^{\lceil \frac{\nu}{|b|} \rceil} \phi_n(\nu - n|b|) - \sum_{n=0}^{\lceil \frac{y}{|b|} \rceil} \phi_n(y - n|b|) \right\}$$

$$\phi_n(y) = e^{\frac{\lambda y}{a}} \left(\sum_{k=0}^n \frac{\left(\frac{-\lambda y}{a} \right)^k}{k!} \right) - 1;$$

Optimality of CUSUM

$$\mathcal{J}_L(T) = \sup_{t \geq 0} \text{essup} \mathbf{E}_t[T - t | T > t, \mathcal{F}_t]$$
$$\mathbf{E}_\infty[T] \geq \gamma$$

$$\mathbf{E}_\infty[T] \geq \mathbf{E}_\infty[T_C] = \gamma \quad \Rightarrow \quad \mathcal{J}_L(T) \geq \mathcal{J}_L(T_C)$$

$$h(y) = \mathbf{E}_\infty[T_C | y_0 = y]$$

$$g(y) = \mathbf{E}_0[T_C | y_0 = y]$$

Equalizer

$$\text{essup} \mathbf{E}_t[T_C - t | T_C > t, \mathcal{F}_t] = g(0)$$

$$\mathcal{J}_L(T_C) = g(0) \quad \mathbf{E}_\infty[T_C] = h(0)$$

$$E_{\infty}[T] \geq h(0) \Rightarrow \mathcal{J}_L(T) \geq g(0)$$

Lemma: $\mathcal{J}_L(T) \geq \frac{E_{\infty}[\int_0^T e^{y_t} dt]}{E_{\infty}[e^{y_T}]}$

$$E_{\infty}[T] \geq h(0) \Rightarrow \frac{E_{\infty}[\int_0^T e^{y_t} dt]}{E_{\infty}[e^{y_T}]} \geq g(0)$$

$$E_{\infty}[\int_0^T e^{y_t} dt] = g(0)E_{\infty}[e^{y_T}] + \underbrace{E_{\infty}[T] - \bar{h}(0)}_{\geq 0} + \underbrace{E_{\infty}[h(y_T) - e^{y_T} g(y_T)]}_{\geq 0}$$

Lemma: $h(y) - e^y g(y) \geq 0, \quad \forall y \geq 0$

Nonhomogeneous Poisson

Let $\{\mathcal{N}_t\}$ denote a nonhomogeneous Poisson process with rate λ_t that satisfies

$$\lambda_t = \begin{cases} \omega_t & \text{for } t \leq \tau \\ \rho \omega_t & \text{for } t > \tau, \end{cases}$$

ω_t is adapted to our observations (that can include more information than $\{\mathcal{N}_t\}$) and ρ is a known constant.

CUSUM test:

$$u_t = (1 - \rho) \int_0^t \omega_s ds + (\log \rho) \mathcal{N}_t$$

$$m_t = \inf_{0 \leq s \leq t} u_s$$

$$y_t = u_t - m_t$$

$$T_C = \inf\{t : y_t \geq \nu\}$$

Is it optimum? **YES !** but in a Lorden-like sense

$$\mathcal{J}_L(T) = \sup_{t \geq 0} \text{essup} \mathbf{E}_t[\mathcal{N}_T - \mathcal{N}_t | T > t, \mathcal{F}_t]$$

$$\mathbf{E}_\infty[\mathcal{N}_T] \geq \gamma$$

Epidemic Detection

Detect the onset of an epidemic: when the **incidence rate** of a particular disease increases significantly above some standard level

ω_t : describes the nominal incidence rate, which can be a function of several observable quantities as population, pollution measurements etc.

$\rho > 1$: smallest increase in incidence rate that should be considered ... alarming.

Detect rapidly **in number of incidences**