

# Does latency prevent the Ogata recursion trick?

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## 1 Setup

### 1.1 Two Kernels

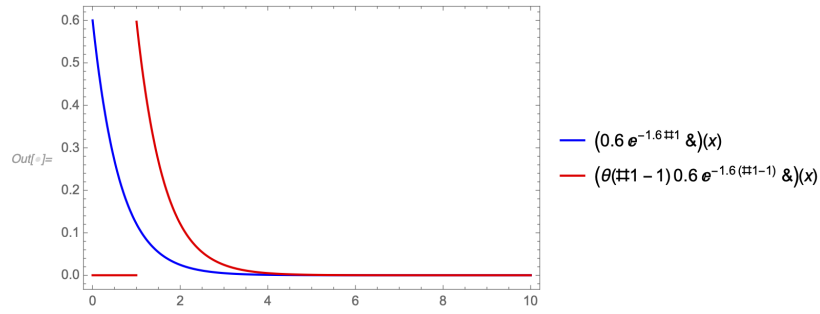


Figure 1: Kernels

### 1.2 Ogata recursion trick

The double summation:

$$\sum_{i=1}^n \log \left( \lambda_0 + \alpha \cdot \sum_{j=1}^{i-1} \exp(-\beta(t_i - t_j)) \right) \quad (1)$$

Can be converted into (Ogata 1978):

$$\sum_{i=1}^n \log(\lambda_0 + \alpha \cdot A_i) \quad (2)$$

With:

$$A_i = \exp(-\beta(t_i - t_{i-1})) \cdot A_{i-1} \quad (3)$$

$$A_1 = 0 \quad (4)$$

But it seems to me that we can only do that for the exponential kernel without latency; if we had 4 timestamps in sequence with a constant difference of  $0.3 \cdot \text{latency}$ , the last timestamp could only be influenced by the first, and I don't see how to do it with the recursion.

## 2 Take the long way home

I programmed the cluster simulation algorithm and the Log-Likelihood estimation (without the Ogata trick and considering latency) in Mathematica.

I checked the average number of points with a large latency against Tick and the average with zero latency against the expected theoretical value and against Tick.

For  $\{\lambda_0, \alpha, \beta\} = \{1.2, 0.6, 0.8\}$ ,  $T = 200$  and the exponential kernel with latency 1:  $\alpha \cdot \exp(-\beta \cdot (t-1)) \cdot \mathbb{1}_{t \geq 1}$  I have two results for the estimation (without latency and with latency):

```
Timing[NMaximize[{loglk[λ0, α, β, testel, 0], 0.5 < λ0 < 2.0 && 0.2 < α < β < 1.2}, {λ0, α, β}]]
{364.383, {555.102, {λ0 → 1.20848, α → 0.27753, β → 0.366348}}}
```

```
Timing[NMaximize[{loglk[λ0, α, β, testel, 1], 0.5 < λ0 < 2.0 && 0.2 < α < β < 1.2}, {λ0, α, β}]]
{355.78, {563.68, {λ0 → 1.33498, α → 0.532636, β → 0.730975}}}
```

Figure 2: Mathematica estimation

## 3 Question

I'm going to program the Log-Likelihood estimation (without the Ogata trick and considering latency) in Python now. Is it needed or can I do something else? Is the Ogata trick impossible with latency?