

# The biochemical resolving power of fluorescence lifetime imaging

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In this Notebook, we retrace the mathematical formalism presented in Kollner and Wolfrum ([https://doi.org/10.1016/0009-2614\(92\)87068-Z](https://doi.org/10.1016/0009-2614(92)87068-Z))

We will generate Fig. 1 of their paper to provide an example of how to evaluate Fisher information and to test our analytical techniques used for the case of Gaussian IRFs in a different file.

First, we initialize the Notebook and some parameters as defined in Kollner and Wolfrum.

```
In[*]:= (* Define subscripted symbols *)
<< Notation`
Symbolize[σ_]

(* Define positivity of all variables, and set substitution rules *)
opts = {B > 0, α > 0, m > 1, i > 1, τ > 0, A > 0, T > 0, C > 0, t ≥ 0, p > 0};
subs = {τ → 1/α, A → B α};
subsinv = {α → 1/τ, B → A/α};
```

We then make a convolution integral (f1) of a Dirac excitation pulse with an exponential decay starting from the origin. The number of photons emitted from t=0, to a generic time is the integral  $\Lambda[t]$ .

```
In[*]:= f1 = Convolve[DiracDelta[x], UnitStep[x] A Exp[-x / τ], x, y]
Λ[t_] = Integrate[f1 /. subs, {y, 0, t}, Assumptions → opts]
```

```
Out[*]:= A e- $\frac{y}{\tau}$  UnitStep[y]
```

```
Out[*]:= B - B e-t α
```

The number of photons detected within a time bin is  $G[i] = \Lambda[iT] - \Lambda[(i-1)T]$ , where  $i$  is the index bin and  $T$  is the bin width. With  $G[i]$  we indicate the expectation of  $\Lambda[iT] - \Lambda[(i-1)T]$ , which is otherwise explicitly indicated with the analytical form of the integral  $\Lambda[t]$  to permit derivation of  $\Lambda[t]$  but not of the expectations.

```
In[*]:= LogL[B_, α_] = Sum[Log[Λ[i T] - Λ[(i - 1) T]] G[i], {i, 1, m}] - Sum[(Λ[i T] - Λ[(i - 1) T]), {i, 1, m}]
Out[*]:= -B + B e-m T α + Sum[G[i] Log[B e-(-1+i) T α - B e-i T α], {i, 1, m}]
```

The Fisher information matrix can be defined like

$$I = \begin{pmatrix} a & c \\ c & b \end{pmatrix}$$

where a, b, and c are:

```
In[ ]:= a =
  Simplify[-D[LogL[B, α], {α, 2}] /. {G[i] → (Λ[i T] - Λ[(i - 1) T])}, Assumptions → opts]
b = Simplify[-D[LogL[B, α], {B, 2}] /. {G[i] → (Λ[i T] - Λ[(i - 1) T])},
  Assumptions → opts]
c = Simplify[-D[LogL[B, α], {B, 1}, {α, 1}] /. {G[i] → (Λ[i T] - Λ[(i - 1) T])},
  Assumptions → opts]
Out[ ]:= - 
$$\frac{B e^{-m T \alpha} \left( -e^{(1+m) T \alpha} + m^2 + e^{2 T \alpha} m^2 + e^{T \alpha} (1 - 2 m^2) \right) T^2}{(-1 + e^{T \alpha})^2}$$

Out[ ]:= 
$$\frac{1 - e^{-m T \alpha}}{B}$$

Out[ ]:= 
$$e^{-m T \alpha} m T$$

```

Note that we have substituted  $G[i] \rightarrow (\Lambda[i T] - \Lambda[(i-1) T])$  only after differentiation as these now are the expectations. Next, the standard deviation on the parameter  $\alpha$  can be evaluated from the top-left element of  $I^T$ , and  $\sigma_\tau$  can be evaluated by error propagation from  $\sigma_\alpha$ .

```
In[ ]:= σα = 
$$\sqrt{\frac{b}{a b - c^2}}$$
 /. subsinv
στ = Simplify[
$$\sqrt{\left( (D[\tau /. \text{subs}[[1]], \{\alpha, 1\}])^2 /. \text{subsinv} \right) \sigma_\alpha, \text{Assumptions} \rightarrow \text{opts}}$$
]
Out[ ]:= 
$$\sqrt{\frac{1 - e^{-\frac{m T}{\tau}}}{A \left( -e^{-\frac{2 m T}{\tau}} m^2 T^2 - \frac{e^{-\frac{m T}{\tau}} \left( 1 - e^{-\frac{m T}{\tau}} \right) \left( -e^{-\frac{(1+m) T}{\tau}} + m^2 + e^{\frac{2 T}{\tau}} m^2 + e^{T/\tau} (1 - 2 m^2) \right) T^2}{(-1 + e^{T/\tau})^2} \right) T}}$$

Out[ ]:= 
$$\sqrt{\frac{e^{\frac{m T}{\tau}} (-1 + e^{T/\tau})^2 (-1 + e^{\frac{m T}{\tau}}) T^3}{A \left( e^{T/\tau} + e^{\frac{T+2 m T}{\tau}} - e^{\frac{m T}{\tau}} m^2 - e^{\frac{(2+m) T}{\tau}} m^2 + 2 e^{\frac{(1+m) T}{\tau}} (-1 + m^2) \right) T^2}}$$

```

We can now define the  $F^2$ -value. Note that with C we refer to the total number of photon counts actually measured. C is equal to the integral  $\Lambda[m T]$ , fact that can be used to replace the amplitude A.

FullSimplify can be used to return a compact equation as per Kollner and Wolfrum. However, we also define F2b. F2b is the very same quantity but its extended analytical form avoids undetermined operations of the type 0/0 or 1/0.

```

In[ ]:= F2[τ_, m_, T_] = FullSimplify[
  C  $\frac{\sigma_\tau^2}{\tau^2}$  /. {Solve[(Δ[m T] /. subsinv) == C, A] [[1]] [[1]]}, Assumptions → opts]
F2b[p_, m_] = ExpandAll[
  C  $\frac{\sigma_\tau^2}{\tau^2}$  /. {Solve[(Δ[m T] /. subsinv) == C, A] [[1]] [[1]]} /. {τ -> m T p-1};
  (* Expand All is used to avoid numerical issues with Mathematica *)

```

$$\text{Out[ ]} = \frac{\left(-1 + e^{T/\tau}\right)^2 \left(-1 + e^{\frac{m T}{\tau}}\right)^2 \tau^2}{\left(e^{T/\tau} + e^{\frac{T+2 m T}{\tau}} - e^{\frac{m T}{\tau}} m^2 - e^{\frac{(2+m) T}{\tau}} m^2 + 2 e^{\frac{(1+m) T}{\tau}} (-1 + m^2)\right) T^2}$$

```

In[ ]:=

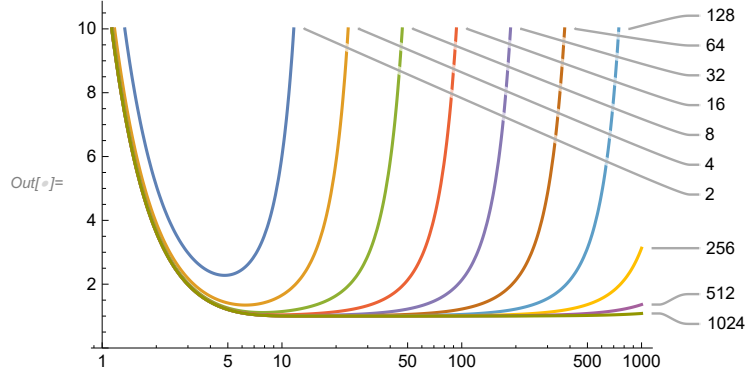
```

Next we generate Fig. 1 from Kollner and Wolfrum.

```

In[ ]:= mlist = 2^Table[m, {m, 1, 10}];
LogLinearPlot[Evaluate@F2b[p, mlist],
  {p, 1., 1000.}, PlotRange → {0, 10}, PlotLabels → mlist]

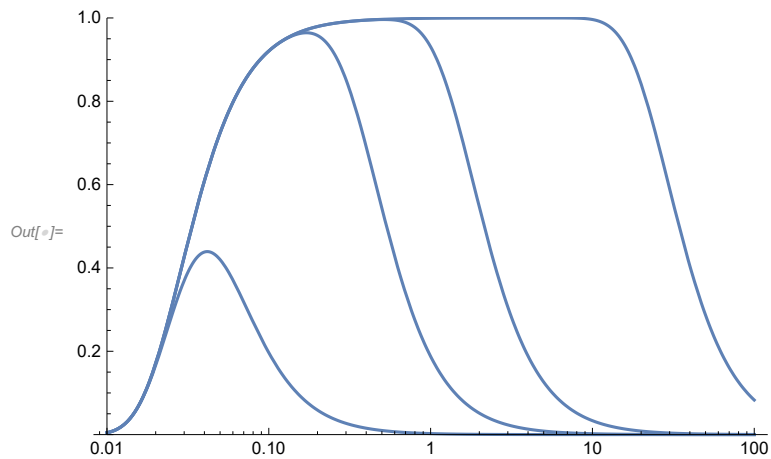
```



```

In[ ]:= (* This is the photon-efficiency *)
LogLinearPlot[1/F2[τ, {2, 16, 64, 1024}], .1], {τ, 0.01, 100}, PlotRange → {0, 1}]

```



```

In[ ]:=

```