# Weakly-supervised inferences of molecular dynamics for fluorescence imaging in physiological environments



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## A. Context

### Measurements from fluorescence microscopy are noisy & cluttered [1]



### Mapping measurements to trajectories is an inverse data association problem [2]





Microscope



 $\mathbf{X}_t = \{\mathbf{x}_t^i\}_{i \in \Omega_t}$ 

$$\mathbf{Z}_t = \{\mathbf{z}_t^j\}_{i \in D_t}$$

Data association is a combinatorially hard problem [3]





Can LLMs be used for long-range sequence-dependencies & combinatorial complexity?

**B.** Method



**Experimental setup** 





Attention-based method (transformer [4])

Simulation of a simple system of 2 particles in 2 dimensions without any false positives & increasing measurement noise following standard Brownian motion

### C. Main Results: Transformers are robust for long sequences but MHT remains optimal for short sequences







### Hint to hybridise?



Emergence of 2 regimes: small sequences where MHT performs better and big sequences where attention performs better

While attention prolongs the noise-cliff, it breaks the same way as MHT does

Hypothesis: MHT would be optimal (and hence, perform best) if it could easily access the entire lookback window.

For small sequences, this is still feasible. What happens then?

When MHT is indeed optimal, i.e., for small sequences, attention is unable to match its accuracy

References:	Ackn
[1] Vonesch et al, 2006	The pro
[2] P. Emami, P. M. Pardalos, 2020	2030, t
[3] Reid <i>,</i> 1979	manage
[4] Vaswani et al, 2017	Agency
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