

Constraining jet quenching models in heavy-ion collisions using Bayesian Inference

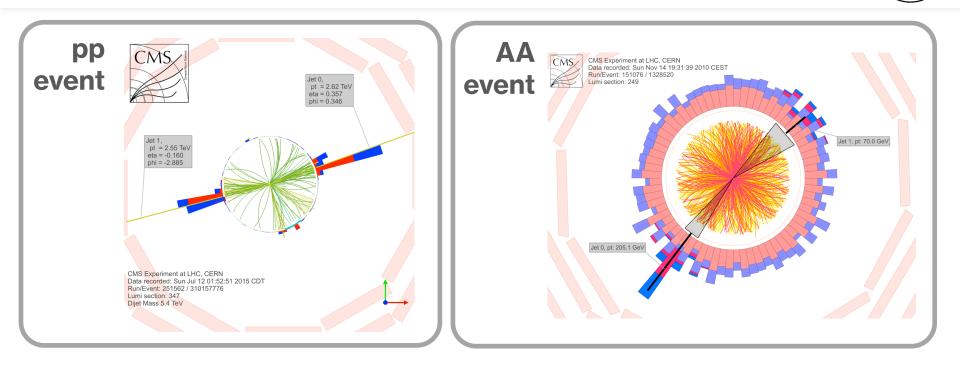
Alexandre Falcão*

Mathias Nilsen Konrad Tywoniuk

* alexandre.falcao@uib.no

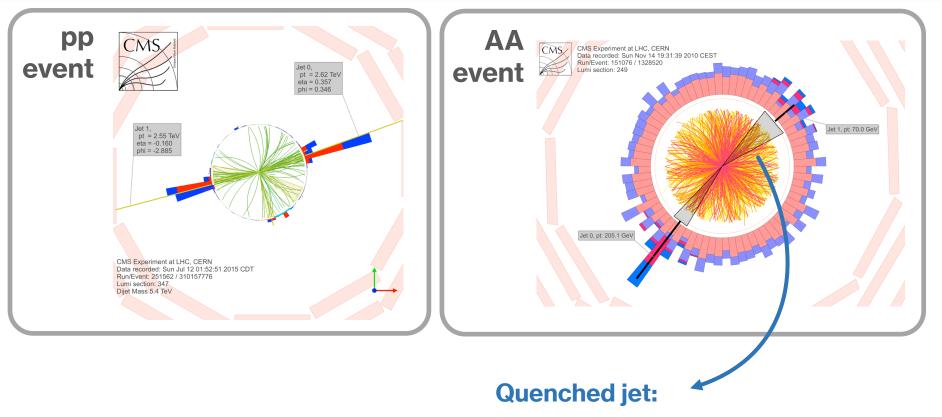
Nov. 3rd, 2022 ML4Jets 2022

Jet quenching in heavy-ion collisions



$$\begin{pmatrix} AA \\ collision \end{pmatrix} \neq A \times \begin{pmatrix} pp \\ collision \end{pmatrix}$$

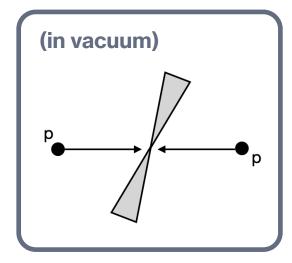
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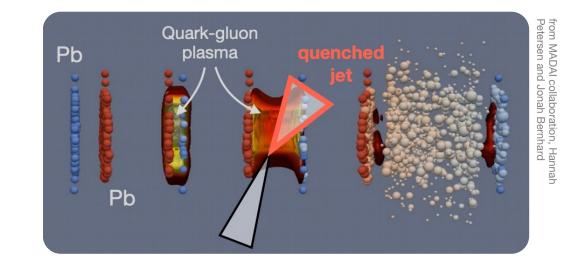


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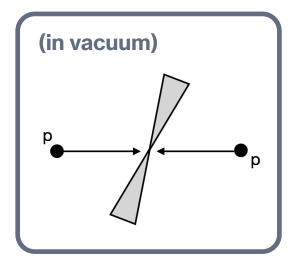
- modification of the transverse energy balance
- modification of jet internal structure
- suppresion of the jet yields

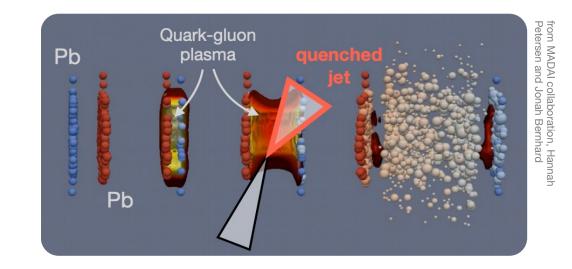








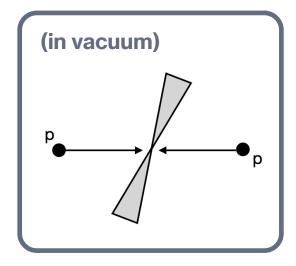




Jets in medium:

- Quark gluon plasma (QGP) is created in the heavy-ion collision
- Jet created by hard process within QGP probes the medium
- Medium properties can be retrieved by studying jet quenching

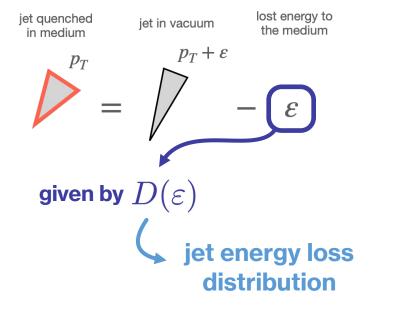




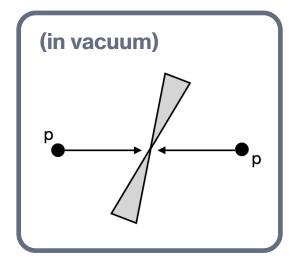
Pb Pb Pb Pb Pb

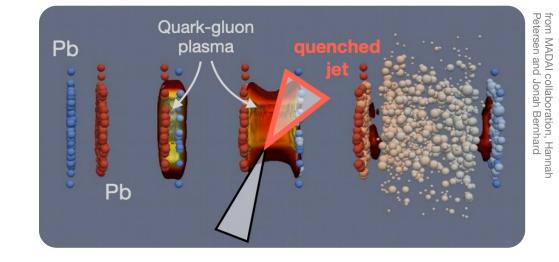
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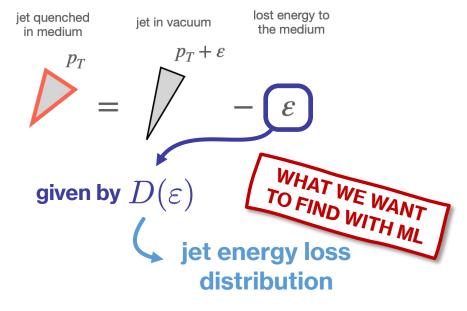






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One parton through the medium

The energy loss distribution via medium induced gluon emissions of a hard parton can be computed from the theory side [Arleo 2002, Baier 2001]

$$D(\epsilon) = \sum_{n=0}^{\infty} \frac{1}{n!} \left[\prod_{i=1}^{n} \int d\omega_i \frac{dI(\omega_i)}{d\omega} \right] \delta \left(\epsilon - \sum_{i=1}^{n} \omega_i \right) \cdot \exp \left[-\int_0^{+\infty} d\omega \frac{dI(\omega)}{d\omega} \right]$$

$$\begin{pmatrix} L & n & \text{number of radiated gluons} \\ \omega_i & \text{energy of emitted gluon } i \\ \frac{dI}{d\omega} & \frac{dI}{gluon spectrum} \end{pmatrix}$$

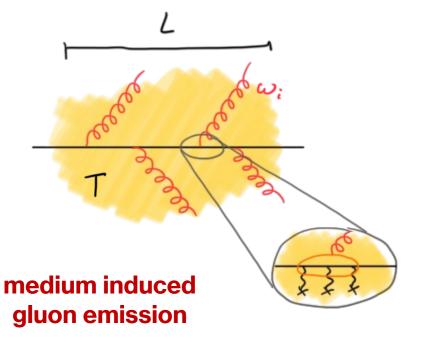
medium induced gluon emission



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 $\frac{\mathrm{d}I}{\mathrm{d}\omega} \quad \begin{array}{l} \text{medium-induced} \\ \text{gluon spectrum} \end{array}$

Depends on:

- medium length: L
- transport coefficient: $\hat{q}(T)$
- parton colour: C_R

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$$\begin{array}{c}
 q & \overbrace{f} & f & g & f & g \\
 \hline
 q & \overbrace{f} & f & f & g & f & f & f \\
 \langle \varepsilon \rangle_g &= \frac{C_A}{C_F} \langle \varepsilon \rangle_q
\end{array}$$

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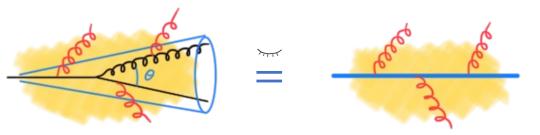


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A jet through the medium

When a "vacuum" splitting happens:

• If splitting anlge is smaller than medium resolution angle



splitting is not resolved

(medium does not see the splitting)

$$D_{\rm jet}(\varepsilon) = D_q(\varepsilon)$$

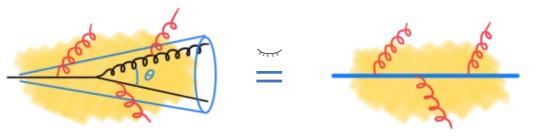


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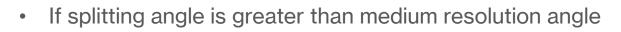
 $D_{\rm jet}(\varepsilon) = D_q(\varepsilon) \otimes D_{\rm MR}$ mediur response

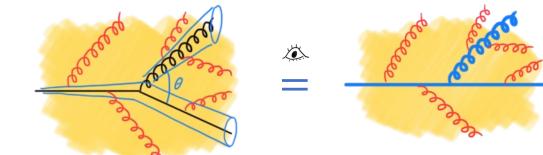


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medium
response

splitting is resolved $D_{\text{jet}}(\varepsilon) = D_q(\varepsilon_q) \otimes D_g(\varepsilon_g) \otimes D_{\text{MR}}$ with $\varepsilon = \varepsilon_q + \varepsilon_q$

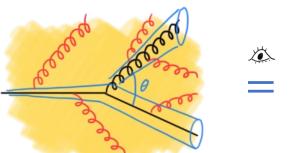


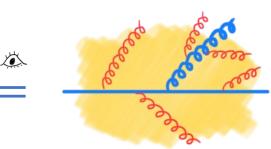
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 $egin{aligned} D_{ ext{jet}}(arepsilon) &= D_q(arepsilon_q) \otimes D_g(arepsilon_g) \otimes D_{ ext{MR}} \ \end{aligned}$ with $egin{aligned} arepsilon &= arepsilon_q + arepsilon_g \ \end{aligned}$

$D(\varepsilon)$ is sensitive to the jet substructure and medium response

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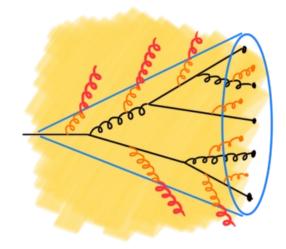
Factorisation



What to keep in D(e) to achieve universality?

• has been done [arXiv:1808.05310]:

 $D(\varepsilon|C_R, \hat{q}(T), L, p_T, R) = D(\varepsilon)$



Factorisation



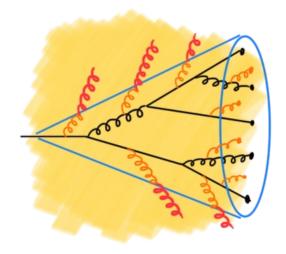
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$$D(\varepsilon|C_R, \hat{q}(T), L, p_T, R) = D(\varepsilon|i), \quad i = q, g$$



Factorisation



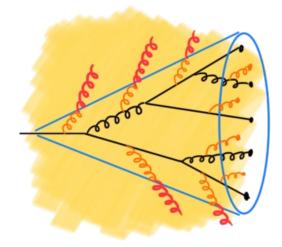
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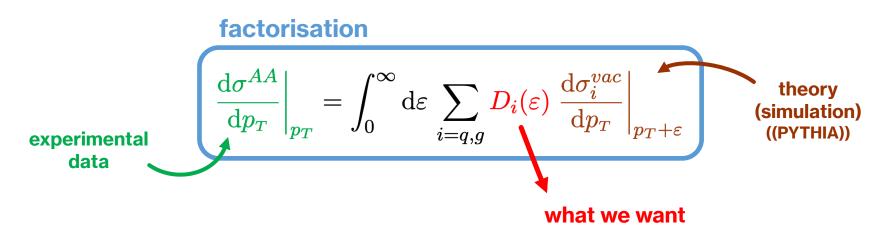
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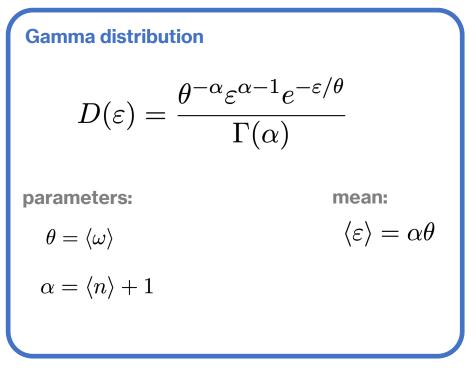
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Modelling the jet energy loss

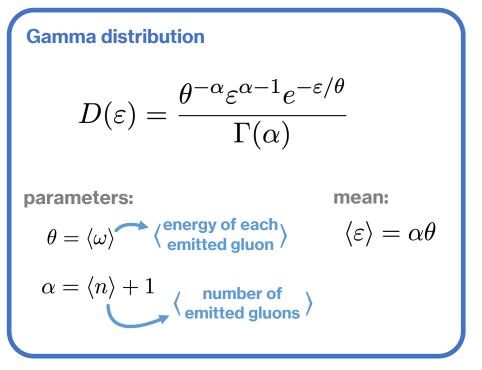
Proposed distribution:





Modelling the jet energy loss

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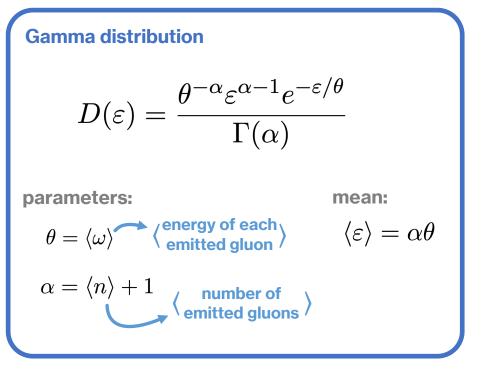
Pros:

- interepretable parameters
- convolution of gamma dist. is still a gamma dist.



Modelling the jet energy loss

Proposed distribution:



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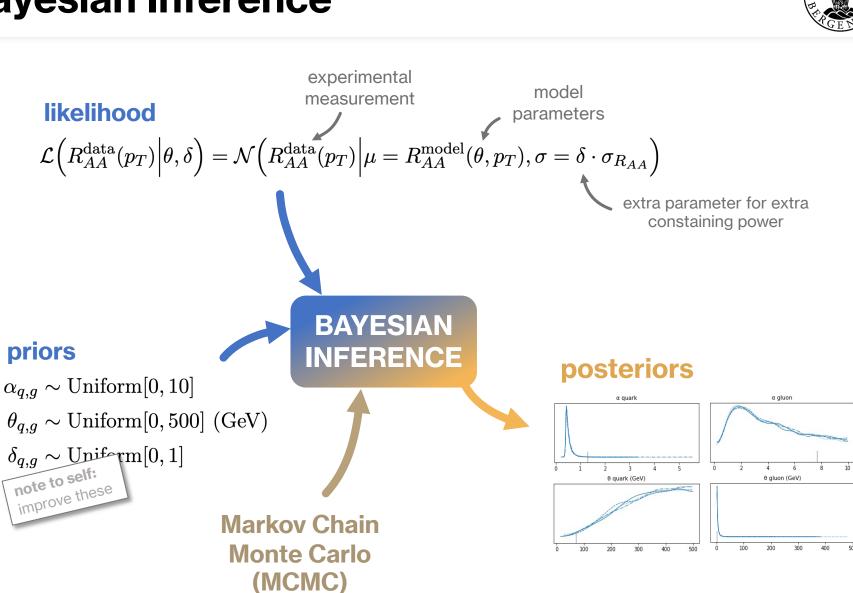
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Cons:

• more assumptions ⇒ more bias

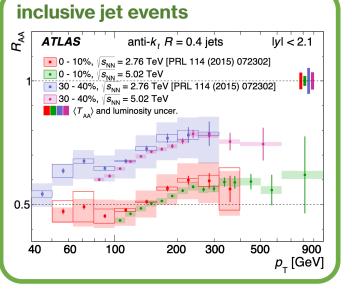


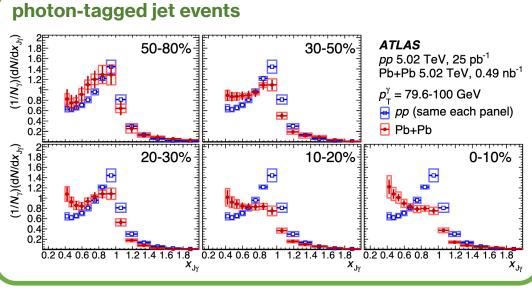
Bayesian Inference



Available experimental data





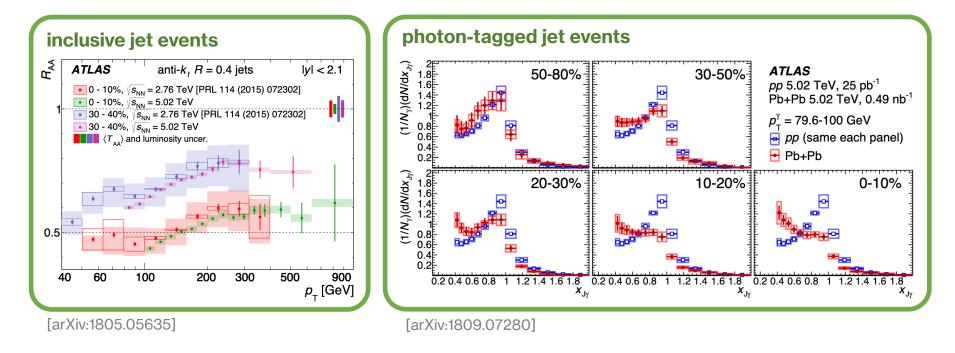


[arXiv:1805.05635]

[arXiv:1809.07280]

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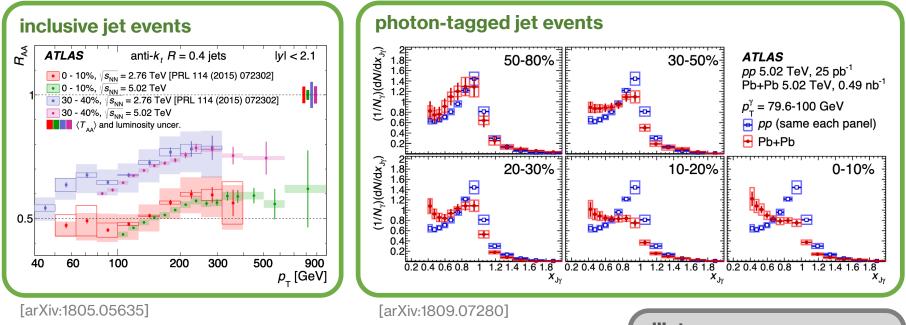




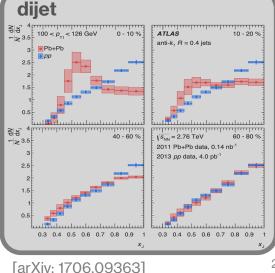
- + different center of mass energies
- + different centralities
- + rapidity cuts

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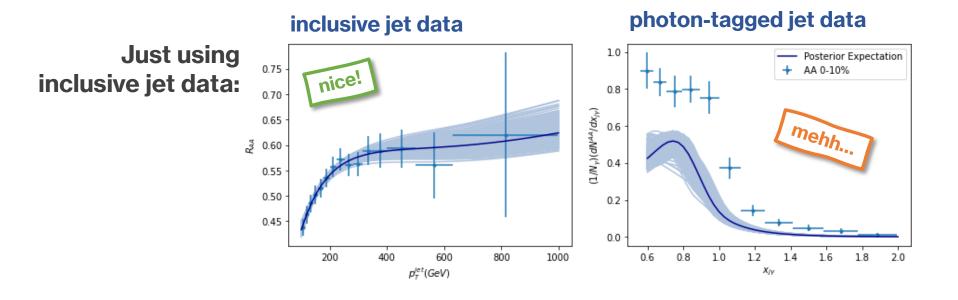




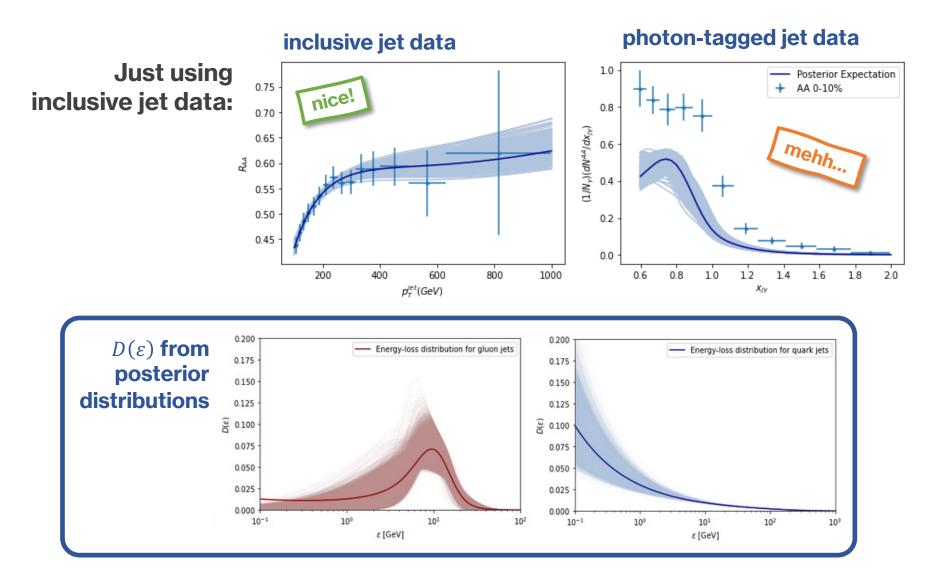
- + different center of mass energies
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- + rapidity cuts
- + dijets



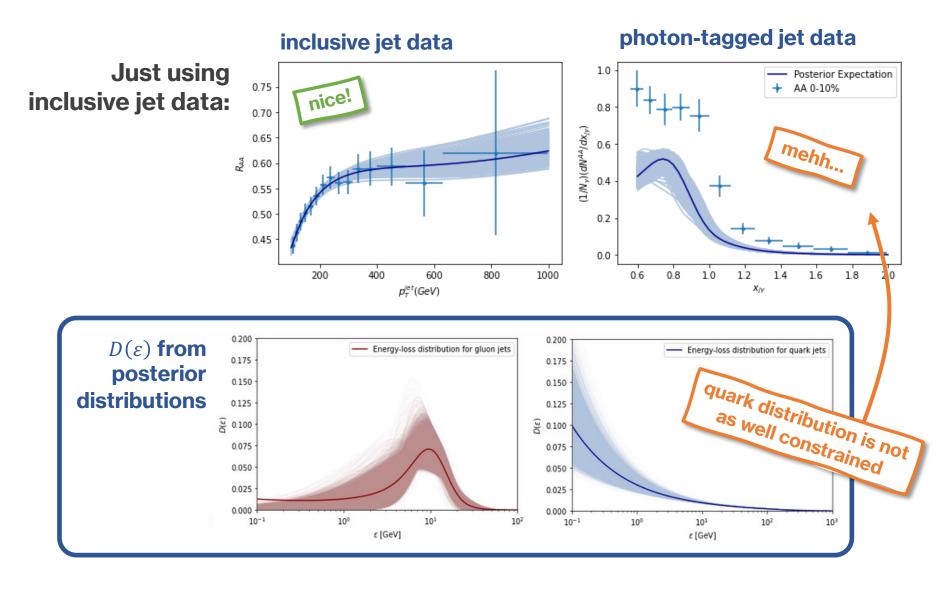






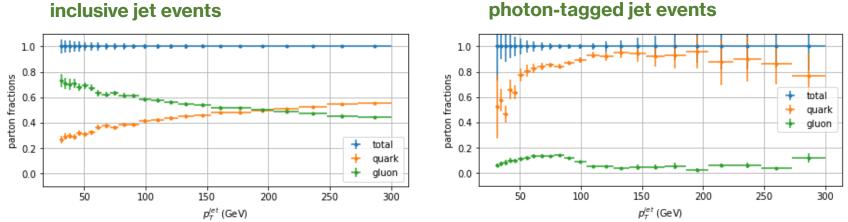








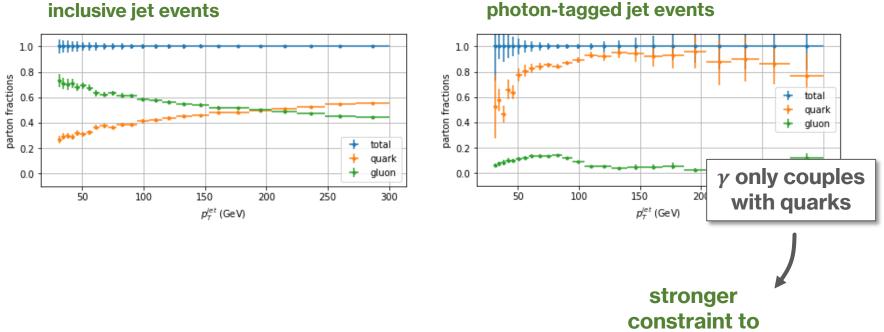
Adding photon-tagged data



photon-tagged jet events



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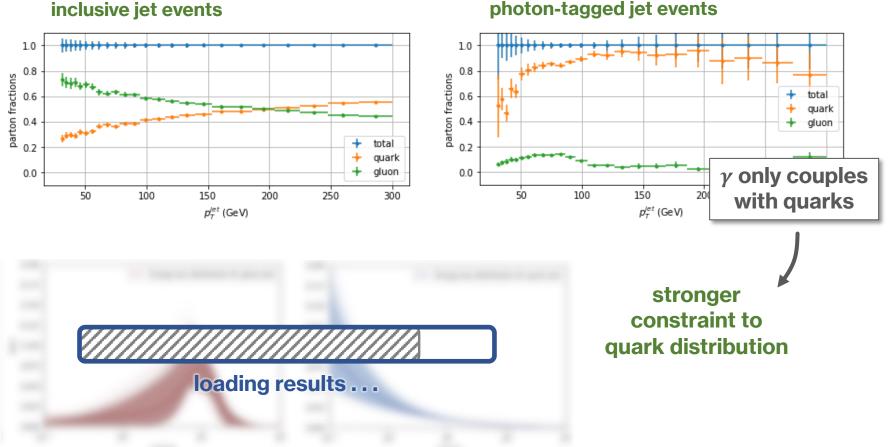


photon-tagged jet events

quark distribution



Adding photon-tagged data



photon-tagged jet events

Outlook and next steps



- The quark gluon plasma (QGP) is formed in heavy-ion collisions;
- Jets produced via hard scatterings probe the QGP;
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Next steps:

- Add different measurements to better learn and validate the model;
- Test the model generalization by using the extracted energy loss distributions to predict other kind of jet observables;
- Address the uncertainties from the simulation data and parameterisation choice
- Move deeper in ML to find a less biased approach.