

# Investigating scattering rates in Thermal Field Theory beyond the leading log

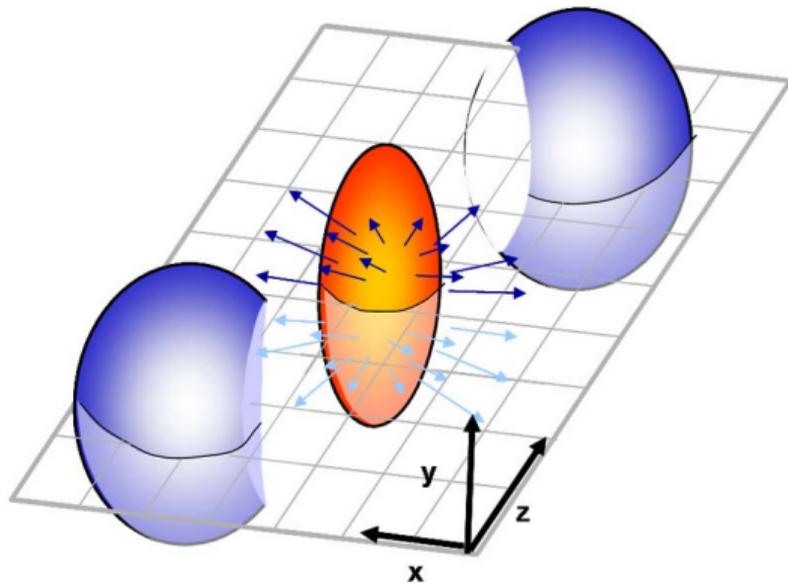
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# Quark-Gluon Plasma (QGP)

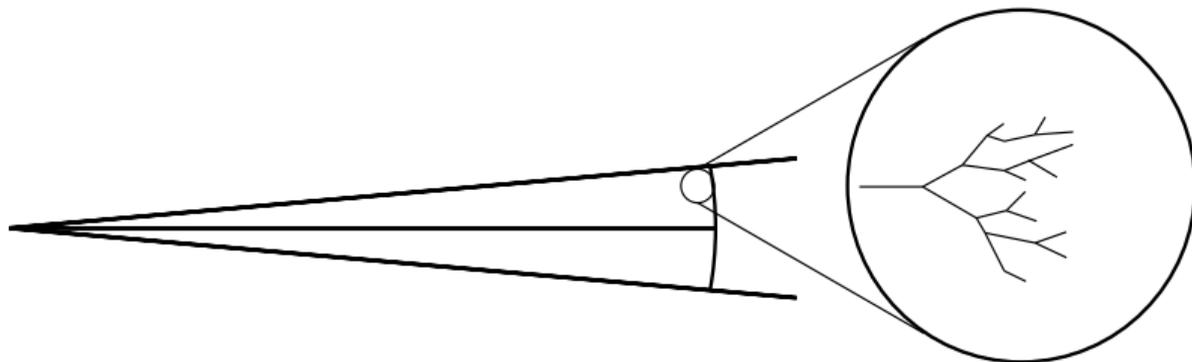
- State of matter, consisting of **deconfined** quarks and gluons
- Existed during the early universe
- Can be experimentally created in Heavy Ion Collisions (at RHIC and LHC)
- **Probes** necessary to gain insights into its behaviour



**Figure:** Collision of two Heavy Ions, creating a QGP

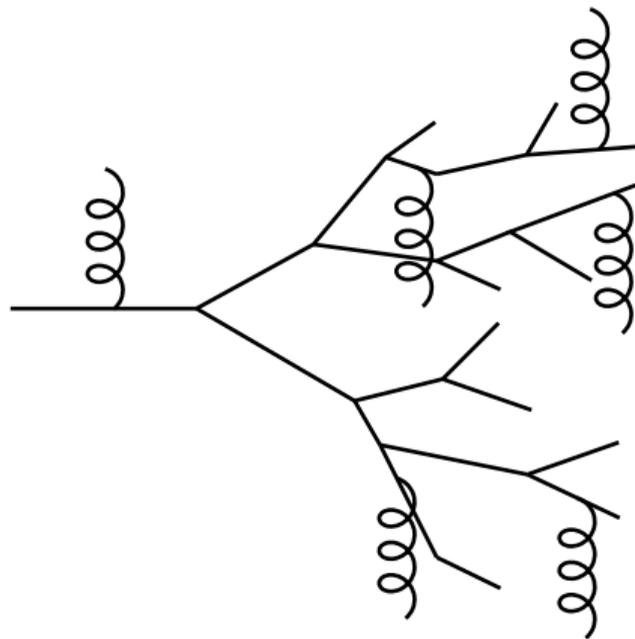
# Jets

- During particle collisions, jets form: a **collimated** spray of **energetic particles**
- These jets evolve through multiple radiations (or splits) into partons of ever increasing lifetime, until partons finally hadronize
- Experimental measurements detect **hadrons** and cluster them to reconstruct the jet's collimated shower
- In collisions not producing a medium (e.g.  $e^+e^-$ ,  $e^-p$ ), the jet evolution is well understood using vacuum splittings



# Jet-medium interactions

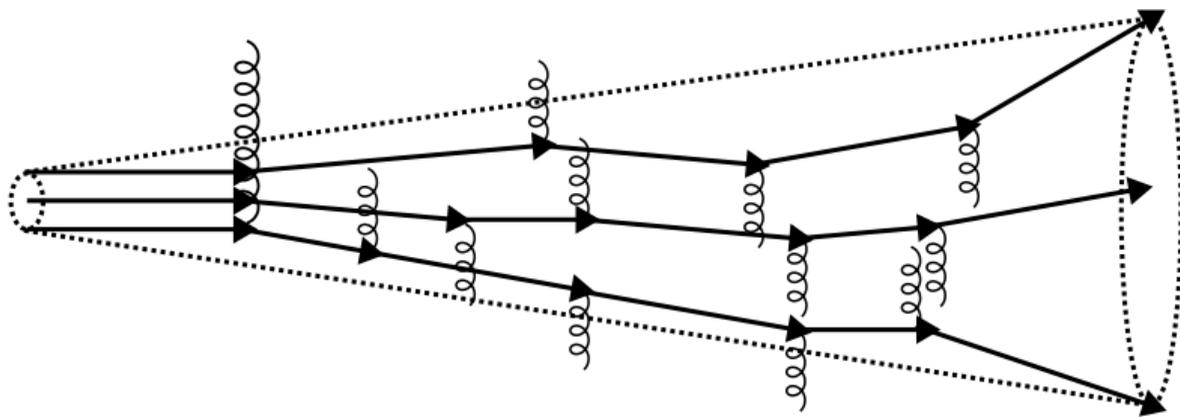
- With a QGP present, **jet-medium interactions** can also decrease the jet parton's virtuality/energy and influence its evolution via scatterings (and medium-modified radiation)
- Leads to **transverse momentum broadening**, a broadening of the jet cone due to “kicks” from the medium



**Figure:** Showering of the jet parton with medium interactions

# Transverse momentum broadening

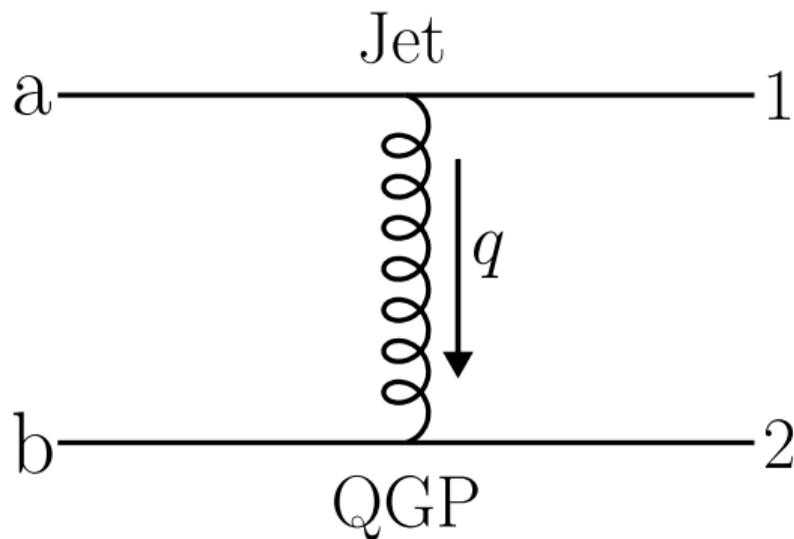
- $\hat{q} := \frac{\langle q_{\perp}^2 \rangle}{L}$ , transport-like coefficient characterizing jet broadening
- Analogous to Brownian motion, where the **spatial** diffusion coefficient is proportional to  $\langle x^2 \rangle$
- $\hat{q}$  is a measure of **transverse momentum** diffusion
- Medium-specific quantity (like other QGP transport coefficients: viscosity, conductivity)



## 2→2-scattering

- To calculate  $\hat{q}$ , **microscopic physics** of the scattering of jet-medium particles is needed
- The fundamental quantity to calculate is the **2 → 2 scattering rate density**  $dR$ , which gives a number of scatterings per unit time and volume:  

$$dR = \frac{d^4 N}{d^4 x}$$

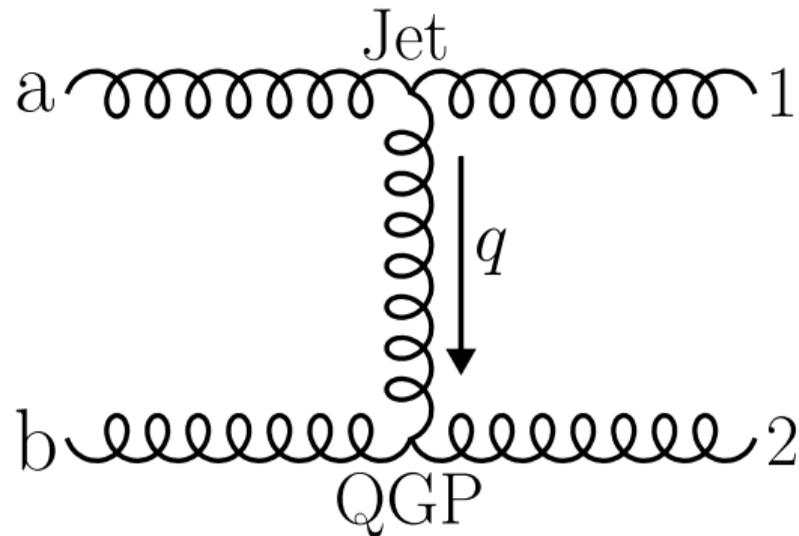


**Figure:** 2→2-scattering of a jet with a medium particle, mediated by a propagator of momentum  $q$

# Gluon-gluon scattering

- To calculate  $\hat{q}$ , **microscopic physics** of the scattering of jet-medium particles is needed
- The fundamental quantity to calculate is the  $2 \rightarrow 2$  **scattering rate density**  $dR$ , which gives a number of scatterings per unit time and volume:  

$$dR = \frac{d^4 N}{d^4 x}$$
- This work considers gg-scattering, for now at tree-level, but other processes and diagrams work similar



**Figure:** Gluon-gluon scattering of a jet with a medium particle, mediated by a propagator of momentum  $q$

# The Scattering Rate in the medium

$$dR = \prod_{i=a,b,1,2} \left[ \frac{d^3 p_i}{(2\pi)^3 2E_i} F_i(p_i) \right] d^4 q \overline{|\mathcal{M}|^2} (2\pi)^4 \delta^{(4)}(p_a - p_1 - q) \delta^{(4)}(p_b - p_2 + q)$$

- $\overline{|\mathcal{M}|^2}$  is the matrix element, encapsulating the probability amplitude of the scattering process
- For **thermal equilibrium**, the initial medium particle obeys the Fermi-Dirac (for quarks), or Bose-Einstein (for gluons) statistics, while the final medium particle is either Fermi-suppressed or Bose-enhanced, giving

$$F_{\text{initial}}(p) = f_{f/b}(p) = \frac{e^{-\beta(u \cdot p)}}{1 \pm e^{-\beta(u \cdot p)}}$$

$$F_{\text{final}}(p) = 1 \mp f_{f/b}(p) = \frac{1}{1 \pm e^{-\beta(u \cdot p)}}$$

- The unknowns are  $\beta = 1/T$ , where  $T$  is the temperature (natural units were used, so  $c = \hbar = k_B = 1$ ), and fluid flow  $u^\mu$

# The Scattering Rate and $\hat{q}$ in the medium

$$dR = \prod_{i=a,b,1,2} \left[ \frac{d^3 p_i}{(2\pi)^3 2E_i} F_i(p_i) \right] d^4 q \overline{|\mathcal{M}|^2} (2\pi)^4 \delta^{(4)}(p_a - p_1 - q) \delta^{(4)}(p_b - p_2 + q)$$

- $\overline{|\mathcal{M}|^2}$  is the matrix element, encapsulating the probability amplitude of the scattering process
- The  $F_i$  are **particle distributions** in thermal equilibrium
- $\hat{q}$  can be calculated from microscopic physics similar to the rate:

$$\hat{q} = \int dq_{\perp} q_{\perp}^2 \frac{d^4 R}{d^3 p_a dq_{\perp}}$$

- $q_{\perp} = |\vec{q}| \sin(\theta)$ , with  $\theta$  being the angle relative to  $\vec{p}_a$
- **Leading log approximation** widely used in QGP jet physics

# The leading log

- Assumes a **small energy-momentum transfer** of the propagator:  
 $p_a \approx p_1 \Rightarrow q \sim q_\perp \ll p_a, p_1$
- Incoming jet scatters mostly in the forward direction, deposits only a small fraction of energy-momentum into the medium
- All integrals solved analytically with this approximation give the following behaviour for gluon-propagators:

$$\frac{d^3 R}{d^3 p_a} \sim \frac{3\alpha_S^2 T^3}{\pi^2} \left[ \frac{1}{m_D^2} - \frac{1}{E_a^2} + \mathcal{O}\left(e^{-\beta E_a}\right) \right]$$

$$\hat{q} \sim \frac{3\alpha_S^2 T^3}{\pi^2} \left[ \ln\left(\frac{E_a}{m_D}\right) + \mathcal{O}\left(e^{-\beta E_a}\right) \right]$$

## Beyond the leading log

- **Approximations** done to reach the leading log result **not necessary**
- Integral up to the last two can be done analytically
- Functional dependence can be gauged from there by expanding distribution function:

$$\frac{d^3 R}{d^3 p_a} \sim \frac{9\alpha_S^2 T}{2\pi^4} \left[ C_1(m_D) - \frac{\pi^2 T}{6E_a} \operatorname{arcsinh} \left( \frac{E_a}{m_D} \right) - \frac{C_2(m_D) T}{E_a} + \mathcal{O}(E_a^{-2}) \right]$$

$$\hat{q} \sim \frac{9\alpha_S^2 T^3}{2\pi^4} \left[ 4\zeta(3) \operatorname{arcsinh} \left( \frac{E_a}{m_D} \right) + C_3(m_D) - \frac{C_4(m_D) T}{E_a} \operatorname{arcsinh} \left( \frac{E_a}{m_D} \right) + \mathcal{O}(E_a^{-1}) \right]$$

- **Differences** for both quantities in their **functional dependence**
- $\hat{q}$  fares better at leading log order, but the leading order terms of the rate are missed

# Graphical results - Rate

- Leading log approximation plateaus much earlier than the full rate
- Functional differences best seen by rescaling the leading log to match in the UV and plot the relative difference of full rate and approximation

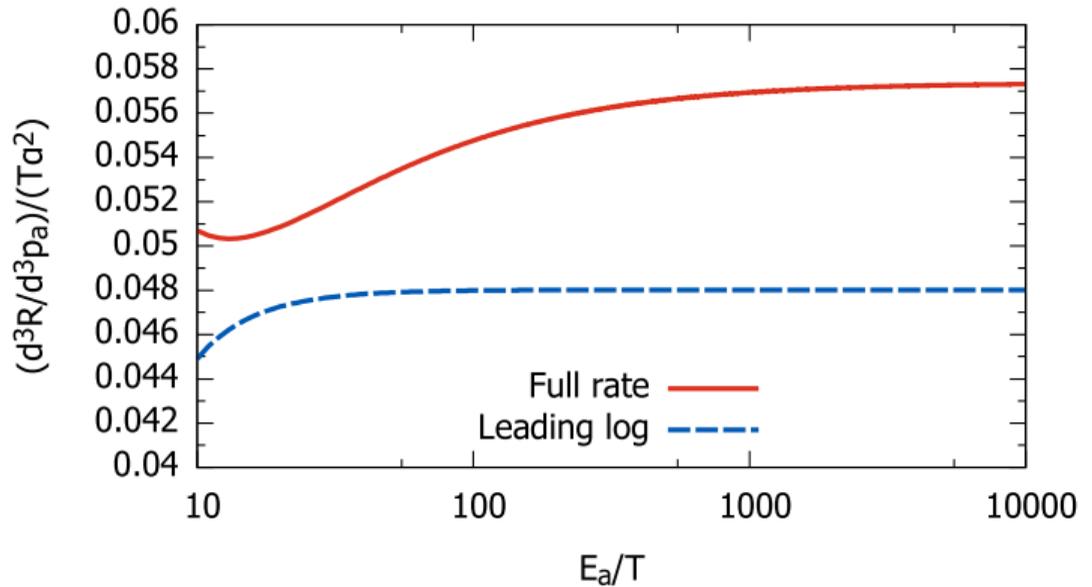
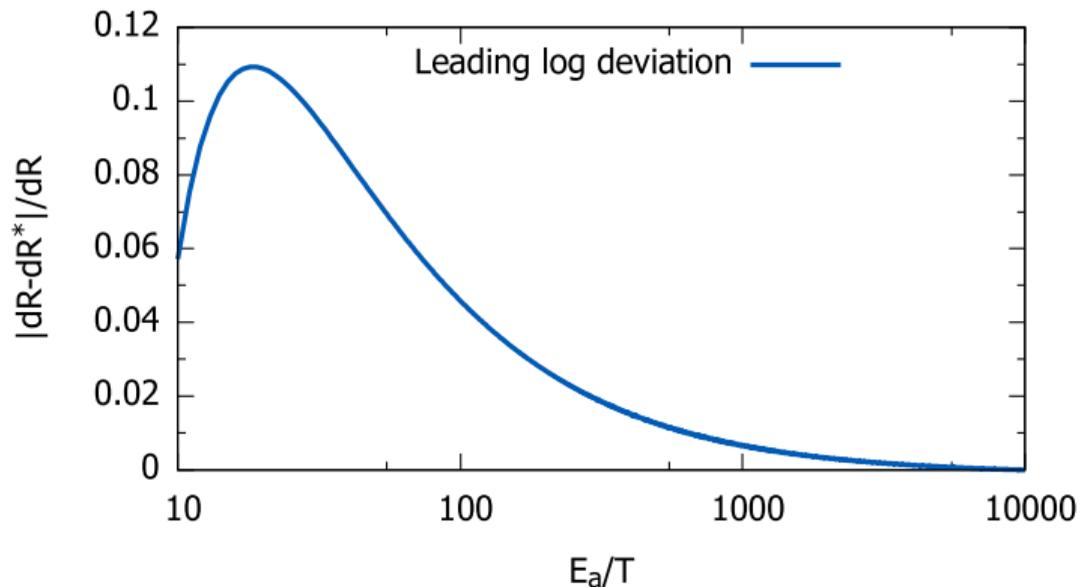


Figure: Comparing the full rate with the leading log

# Graphical results - Rate

- Relative difference of full rate  $dR$  and rescaled leading log  $dR^*$
- $dR^* = dR_{LL} * \frac{dR(10^4)}{dR_{LL}(10^4)}$
- Discrepancy in functional shape easily seen, diverges towards the infrared
- Up to 11% difference, doesn't fall below 1% until late into the UV



**Figure:** Difference of the functional shape of full rate and leading log

# Graphical results - $\hat{q}$

- Leading log works better for  $\hat{q}$  than the rate
- Subtle difference with argument and overall scale of logarithm (shown on log-scale as intercept and slope)
- Again best to rescale leading log and compare the two

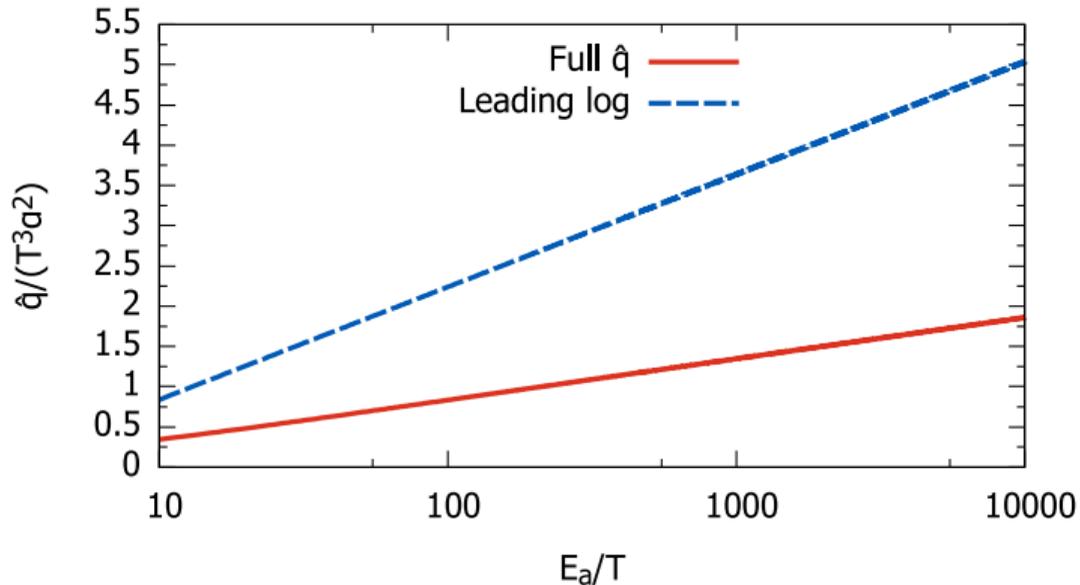
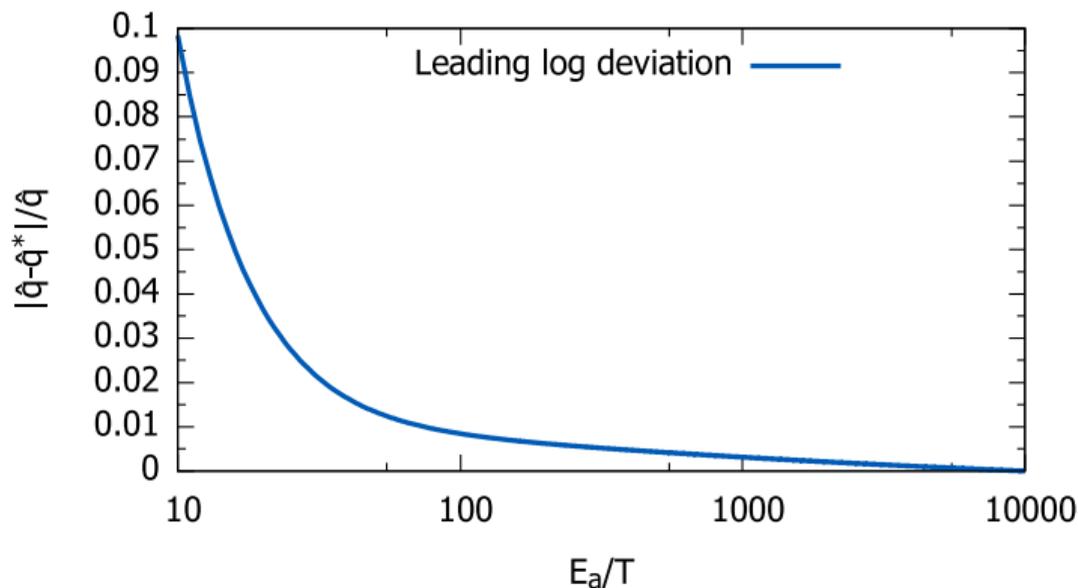


Figure: Comparing the full  $\hat{q}$  with the leading log

# Graphical results - $\hat{q}$

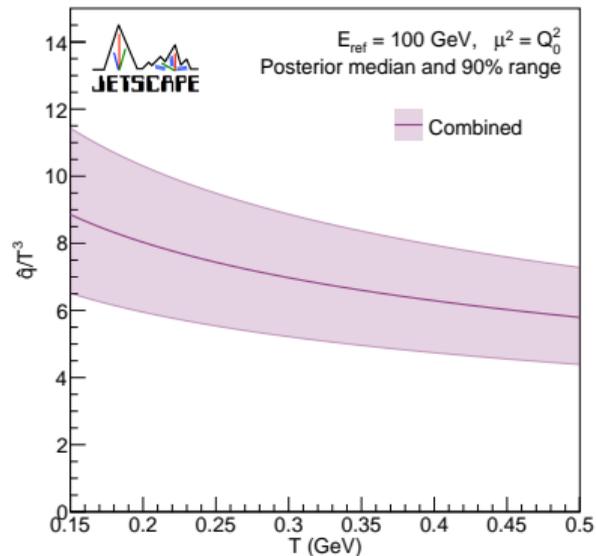
- Relative difference of full  $\hat{q}$  and rescaled leading log  $\hat{q}^*$
- $\hat{q}^* = \hat{q}_{LL} * \frac{\hat{q}(10^4)}{\hat{q}_{LL}(10^4)}$
- Discrepancy again diverges towards the infrared, but falls off faster than for the rate (reaches 1% earlier)
- Remaining difference due to constant shift/argument inside the log



**Figure:** Difference of the functional shape of full  $\hat{q}$  and leading log

# Bayesian analysis

- **Bayesian analyses** model in-medium jet propagation, take model parameter to simulate data
- Bayes theorem then gives **probability distribution** to put constraints on model parameter (i.e.  $\hat{q}$ )
- Forcing the leading log onto the data leads to **biased** constraints
- Not accounting for this **theoretical uncertainty**, the error bands become unreliable



**Figure:** Bayesian constraints on  $\hat{q}$  from JETSCAPE [Phys. Rev. C 111, 054913]

# Summary and outlook

## Results:

- **Leading log** for gg-scattering found to **not accurately** describe the scattering rate
- $\hat{q}$  fares better, but subtle differences found
- Current **Bayesian analyses** of  $\hat{q}$  do not include these terms as **theoretical uncertainties**, yielding an imprecise uncertainty on  $\hat{q}$

## Further improvements:

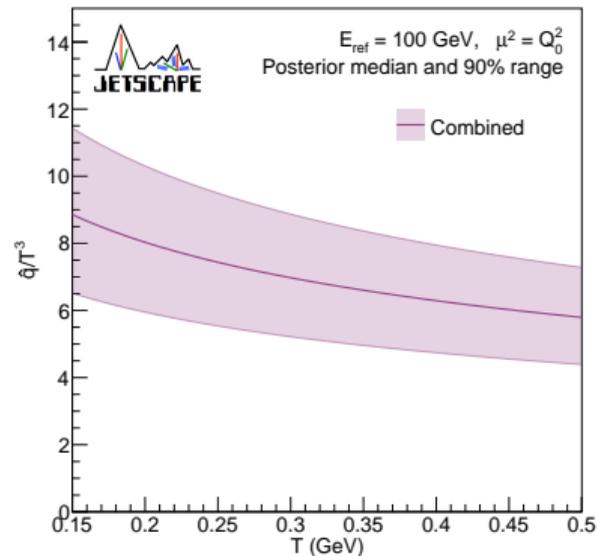
- **Other processes:** since leading log calculations depend mostly on the propagator's form, for quark propagator diagrams results are expected to change
- Instead of the naive cutoff, actual **resummations** have to be taken into account in the IR region (Hard Thermal Loops)
- **Mass and viscous corrections**
- **NLO** contributions

# Section 1

Backup Slides

# Bayesian analysis of nuclear medium properties

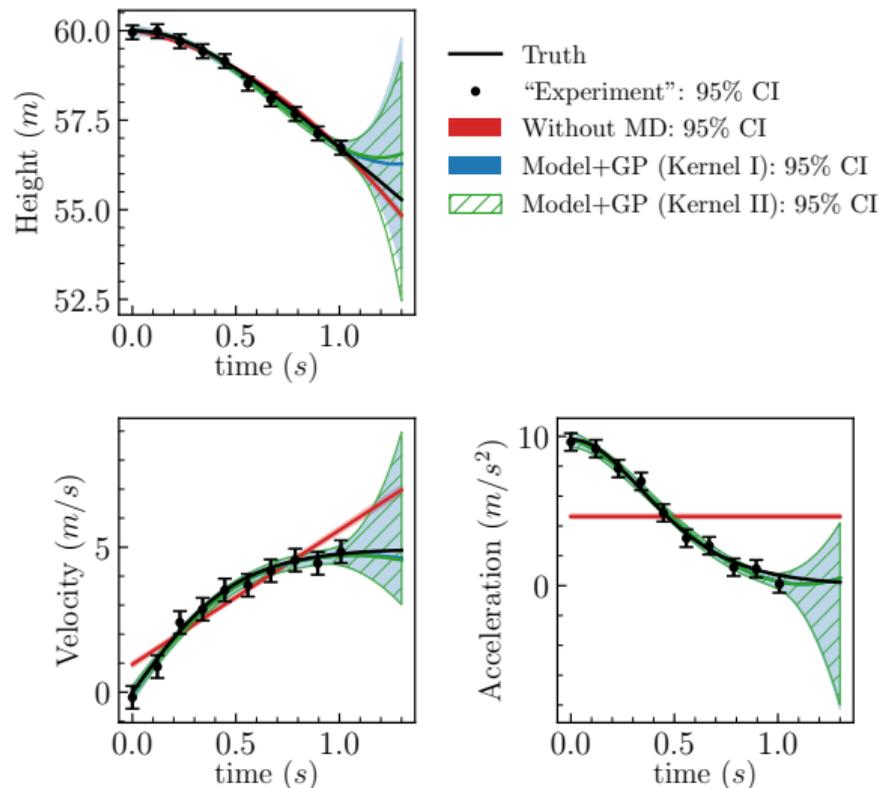
- Large scale Bayesian analyses model jet propagation after the initial collision
- $\hat{q}$  and the rate  $\frac{d^3R}{d^3p_a}$  are used as input to the model to simulate the nature of jet-medium interactions
- Bayesian analysis extract the posterior probability distribution giving constraints on transport coefficients
- Uncertainties in the theory have started to be accounted for (Bayesian model averaging, theory uncertainty via Gaussian process), but leave room for theoretical improvement



**Figure:** Bayesian constraints on  $\hat{q}$  from JETSCAPE [Phys. Rev. C 111, 054913]

# Theoretical uncertainty through Gaussian process - a simple example

- Simple example of a Bayesian analysis, including theory uncertainty for a free-falling ball with air drag, from [arXiv:2504.13144]
- For a model not incorporating this friction term, predictions of observables can be quite off
- Can be remedied by including a Model Discrepancy Gaussian Process (GP) into the Bayesian analysis

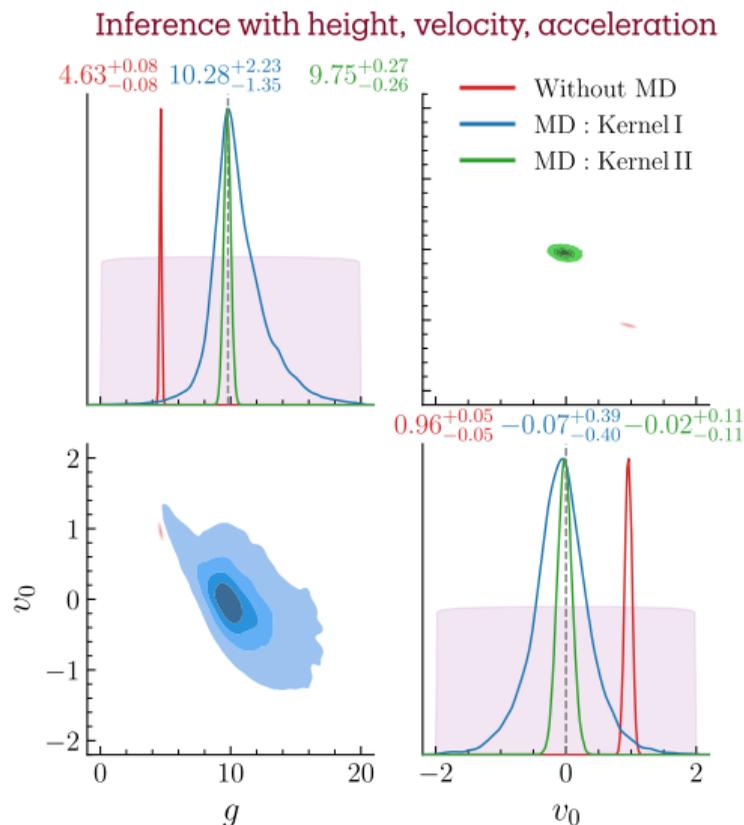


# Theoretical uncertainty through Gaussian process - a simple example

- Observables as function of input space are modeled as  

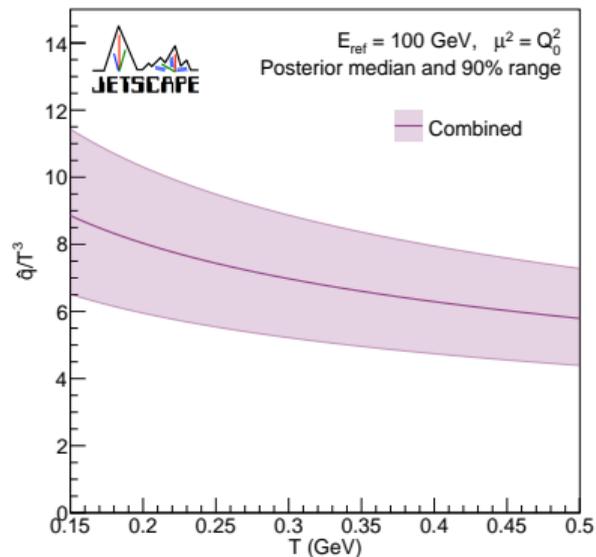
$$y(x_i) = \zeta(x_i) + \epsilon_i = \eta(x_i, \vec{\theta}) + \delta(x_i) + \epsilon_i$$
 with the true value  $\zeta$ , experimental uncertainty  $\epsilon$ , model  $\eta$  and model discrepancy  $\delta$ , and model parameters  $\vec{\theta}$
- $\delta$  is given as a Gaussian process whose covariance kernel depends on the assumed model discrepancy
- Simple kernel in this example:  

$$\sim (x_i x_j)^r \exp\left(-\frac{|x_i - x_j|^2}{l^2}\right)$$



# Bayesian re-analysis

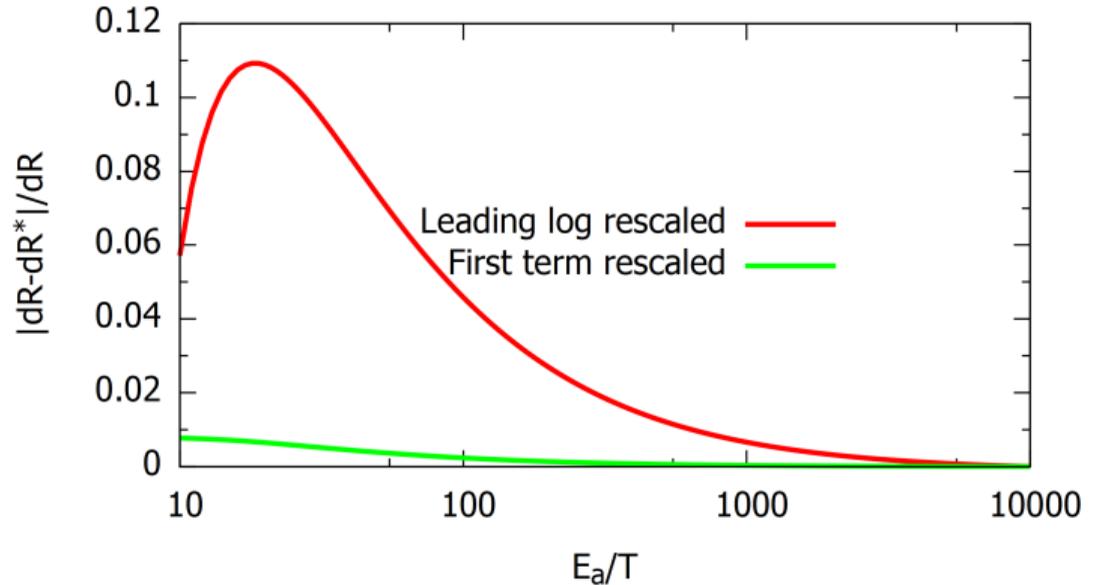
- Need to account for theory uncertainty in the rate and in  $\hat{q}$  owing to truncations in the systematic expansion (e.g. leading-log vs leading order, NLO vs LO and so on)
- This is best done in the ultraviolet regime of QCD, where perturbation theory works best
- Once theory corrections are obtained, they can serve as theoretical systematic uncertainties to previous Bayesian analysis that didn't include these theory corrections



**Figure:** Bayesian constraints on  $\hat{q}$  from JETSCAPE [Phys. Rev. C 111, 054913]

# Graphical results - Rate

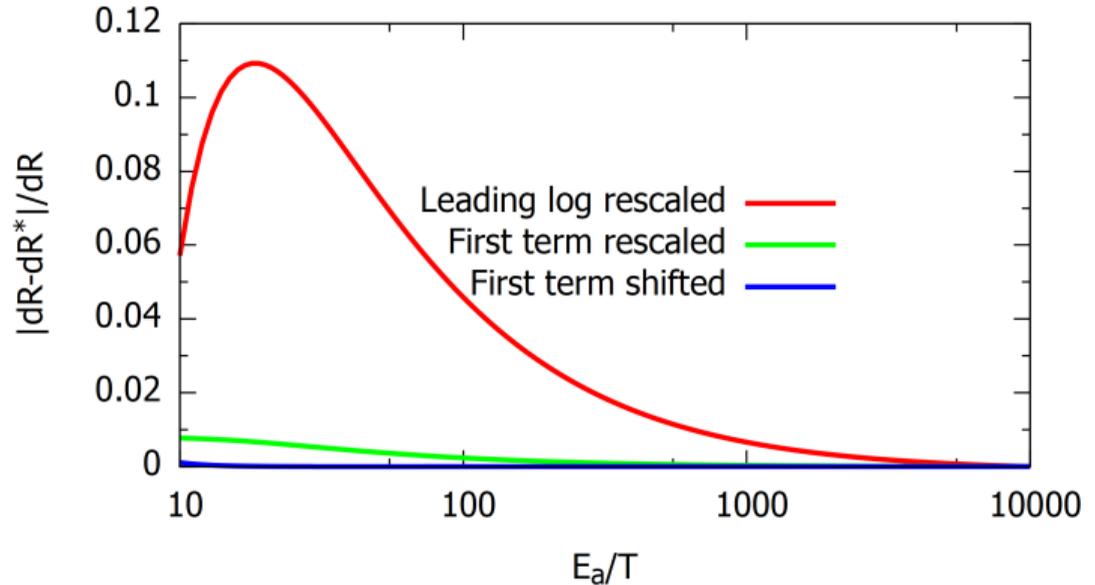
- Option 1: Rescale approximation via  $dR^* = dR_{\text{app}} \cdot \frac{A}{B}$  for  $A := dR(10^4)$ ,  $B := dR_{\text{app}}(10^4)$



**Figure:** Comparing the difference of the full rate and two approximations

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- Option 2: Shift approximation via  $dR^* = dR_{\text{app}} + (A - B)$



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- Option 2: Shift approximation via  $dR^* = dR_{\text{app}} + (A - B)$
- Rescaling works better for leading log
- Shifting more in line with full expansion, as next terms will mostly be constants

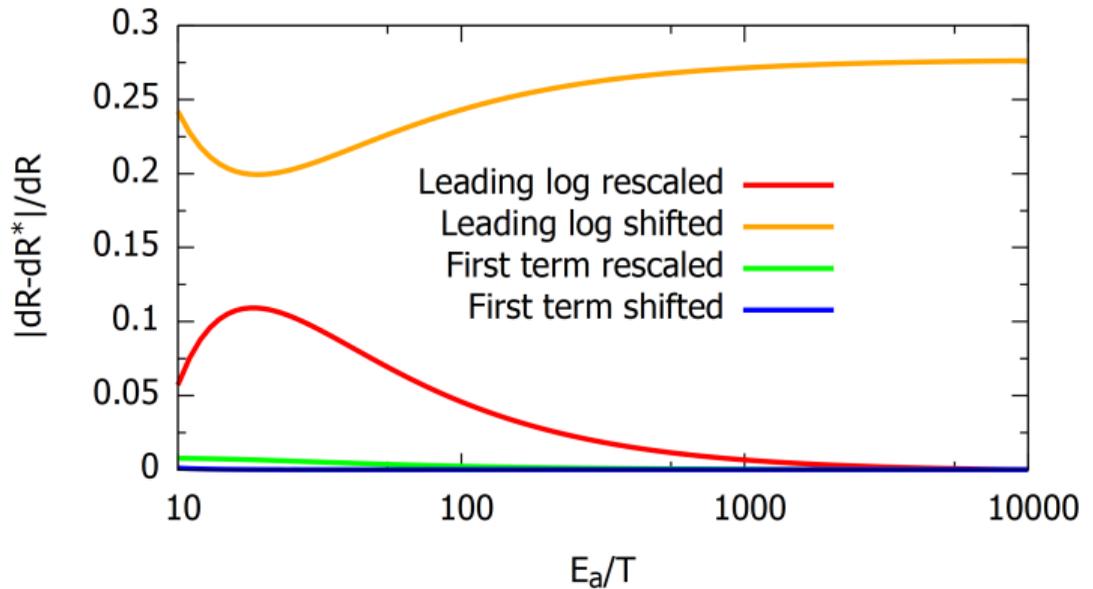
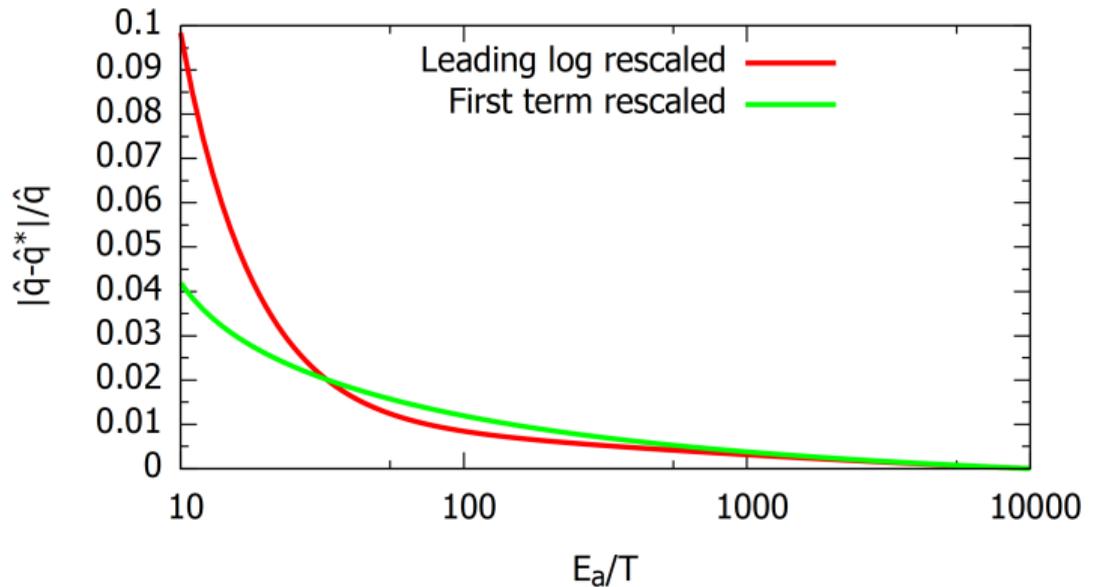


Figure: Comparing the difference of the full rate and two approximations

# Graphical results - $\hat{q}$

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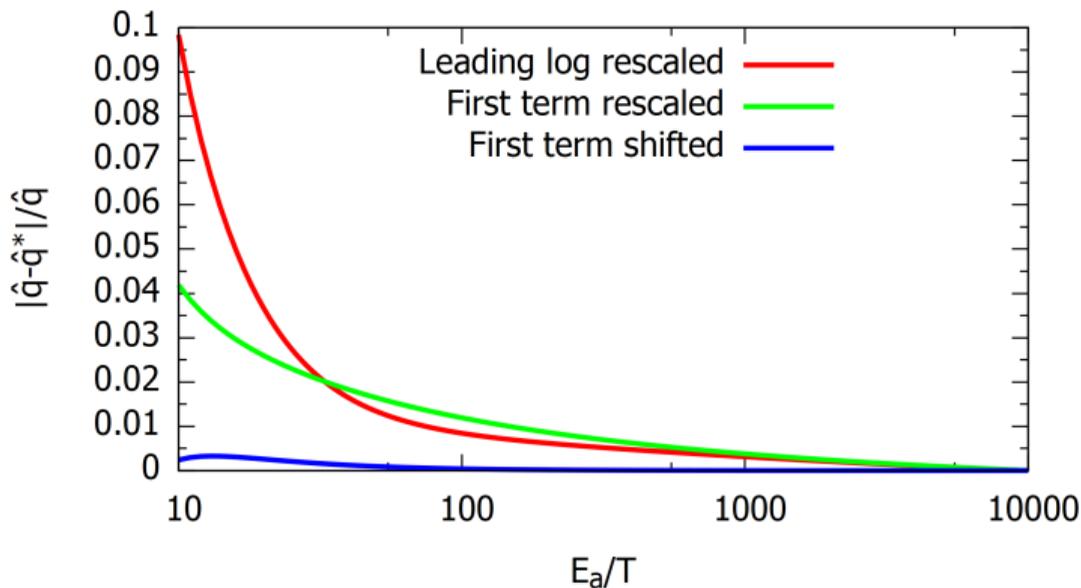
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- Option 2: Shift approximation via

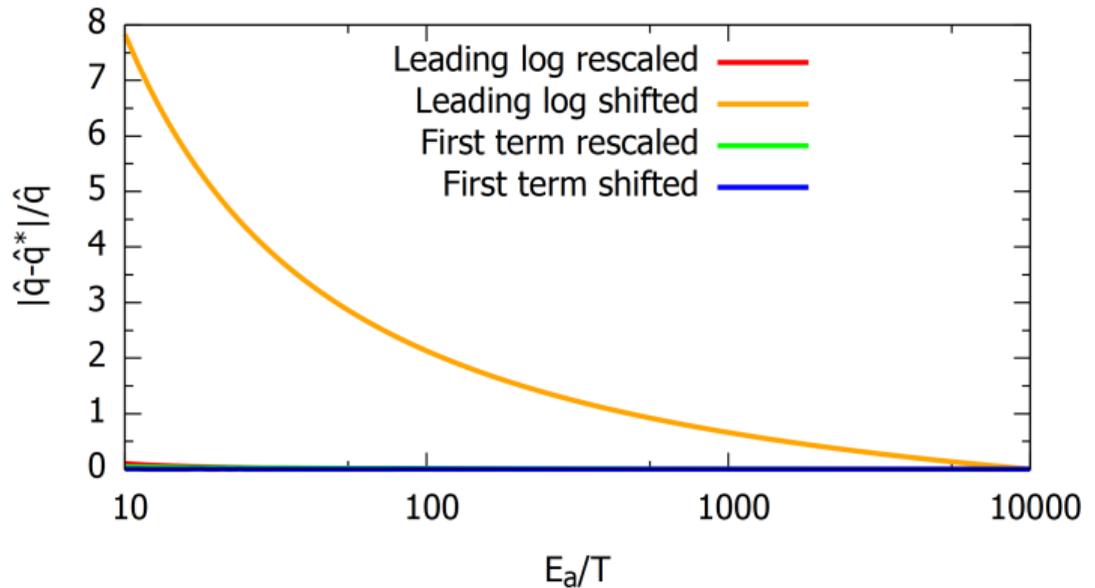
$$\hat{q}^* = \hat{q}_{\text{app}} + (A - B)$$



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- Option 2: Shift approximation via  $\hat{q}^* = \hat{q}_{\text{app}} + (A - B)$
- Subtle difference in argument and overall factor of leading log and log from the first term of full expansion



**Figure:** Comparing the difference of the full  $\hat{q}$  and two approximations