

# A precise $\alpha_s$ determination from the R-improved QCD Static Energy

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$\alpha_s$  2025 Workshop

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Full version of this work available in [arXiv:2510.24846](https://arxiv.org/abs/2510.24846)



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# Introduction

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- Our strategy  $\rightarrow$  comparing the QCD Static Energy obtained in lattice simulations with highly accurate perturbative results.
- $V_{\text{QCD}} \propto \alpha_s \rightarrow$  very sensitive.
- We improve this method building on previous analyses [A. Bazavov, N. Brambilla, X. Garcia i Tormo et al, 2012, A. Bazavov, N. Brambilla, X. Garcia i Tormo et al, 2014] in several ways:
  - Leading renormalon subtraction  $\rightarrow$  short-distance scheme (MSR).
  - Resummation of associated large logs with R-evolution.
  - Profiles functions for the renormalization scales.
- We can fit (for the first time) lattice data up to  $r \sim 0.5$  fm.

# Static Energy and Static Potential

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- The Static Energy is defined as the potential energy between an infinitely massive quark anti-quark pair at a distance  $r$ , corrected by ultra-soft effects. In pNRQCD

$$E_s(r) = V_s(r, \mu) + \delta_{\text{us}}(r, \mu).$$

- The Static Potential is the basic object to understand the behavior of non-relativistic QCD:

$$V_s(r, \mu) = V_s^{\text{soft}}(r, \mu) + V_s^{\text{us}}(r, \mu),$$

$$V_s^{\text{soft}}(r) = -C_F \frac{\alpha_s(\mu)}{r} \sum_{i=0}^3 \left[ \frac{\alpha_s(\mu)}{4\pi} \right]^i \sum_{j=0}^i a_{ij} \log^j(r\mu e^{\gamma_E}).$$

- Coefficients  $a_{i0}$  are known to three loops.  $a_{ij \geq 0}$  obtained with RGE.  $\alpha_s(\mu) = \alpha_s^{(n_\ell=3)}(\mu)$ .
- At  $\mathcal{O}(\alpha_s^4)$  ultra-soft contributions show up for the first time

$$V_s^{\text{us}}(r, \mu) = -\frac{C_A^3 C_F}{12\pi} \frac{\alpha_s^4(\mu)}{r} \log(\mu r e^{\gamma_E}).$$

# Ultra-soft term and resummation

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- $\mu_{us}$  appears both in the ultrasoft static potential  $V_s^{us}$  and in the matrix element  $\delta_{us}$

$$\delta_{us}(\mu_s, \mu_{us}) = -\frac{C_A^3 C_F}{12\pi} \frac{\alpha_s^3(\mu_s) \alpha_s(\mu_{us})}{r} \log \left[ \frac{C_A \alpha_s(\mu_s) e^{-5/6}}{\mu_{us} r} \right]$$

$$V_s^{us}(r, \mu_s, \mu_{us}) = -\frac{C_A^3 C_F}{12\pi} \frac{\alpha_s^3(\mu_s) \alpha_s(\mu_{us})}{r} \log(\mu_{us} r e^{\gamma_E})$$

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- One can see the necessity of resummation with these two (incompatible) choices that minimize logs.
  - $\mu_{us} \sim \alpha_s/r$  [use this one: no large logs in  $\delta_{us}$ ]
  - $\mu_{us} \sim 1/r$  [discarded: sum up logs in  $V_s^{us}$ ]

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- We perform resummation using pNRQCD RGE

$$\mu \frac{dV_s(r, \mu_s, \mu)}{d\mu} = -\frac{2C_F C_A^3}{24r} \frac{\alpha_s(\mu)}{\pi} \left[ 1 + B \frac{\alpha_s(\mu)}{\pi} \right] \alpha_s^3(\mu_s) \left\{ 1 + 3 \frac{\alpha_s(\mu_s)}{4\pi} \left[ a_{1,0} + 2\beta_0 \log(r\mu_s e^{\gamma_E}) \right] \right\}.$$

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- Integrating from  $\mu = \mu_s \sim 1/r$  to  $\mu = \mu_{us} \sim \alpha_s/r$

$$V_s(r, \mu_s, \mu_{us}) = V_s(r, \mu_s) + U_{us}(r, \mu_s, \mu_{us}),$$
$$U_{us}(r, \mu_s, \mu_{us}) = \frac{C_A^3 C_F}{6\beta_0 r} \alpha_s^3(\mu_s) \left\{ \left( 1 + 3 \frac{\alpha_s(\mu_s)}{4\pi} \left[ a_{1,0} + 2\beta_0 \log(r\mu_s e^{\gamma_E}) \right] \right) \log \left[ \frac{\alpha_s(\mu_{us})}{\alpha_s(\mu_s)} \right] + \left( B - \frac{\beta_1}{4\beta_0} \right) \left[ \frac{\alpha_s(\mu_{us})}{\pi} - \frac{\alpha_s(\mu_s)}{\pi} \right] \right\}$$

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- Replacing in the static energy  $\rightarrow$  no large logs for  $\mu_s \sim 1/r$  and  $\mu_{us} \sim \alpha_s/r$

$$E_s(r) = V_s^{\text{soft}}(r, \mu_s) + \frac{C_A^3 C_F}{12} \frac{\alpha_s^3(\mu_s)}{r} \left\{ \frac{2}{\beta_0} \left( B - \frac{\beta_1}{4\beta_0} \right) \left[ \frac{\alpha_s(\mu_{us})}{\pi} - \frac{\alpha_s(\mu_s)}{\pi} \right] \right. \\ \left. + \frac{2}{\beta_0} \left( 1 + 3 \frac{\alpha_s(\mu_s)}{4\pi} \left[ a_{1,0} + 2\beta_0 \log(r\mu_s e^{\gamma_E}) \right] \right) \log \left[ \frac{\alpha_s(\mu_{us})}{\alpha_s(\mu_s)} \right] \right. \\ \left. - \frac{\alpha_s(\mu_{us})}{\pi} \log \left[ \frac{C_A \alpha_s(\mu_s) e^{-5/6}}{r\mu_{us}} \right] - \frac{\alpha_s(\mu_s)}{\pi} \log(r\mu_s e^{\gamma_E}) \right\}.$$

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- Static energy suffers from an  $r$ -independent  $\mathcal{O}(\Lambda_{\text{QCD}})$  renormalon ambiguity.

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$$E_{Q\bar{Q}}(r) = 2m_Q^{\text{pole}} + E_s(r) \quad \text{renormalon-free}$$

- To cancel the renormalon we need a short-distance mass scheme  $\rightarrow$  MSR mass [A. H. Hoang, A. Jain, C. Lepenik, V. Mateu, I. Scimemi and I. W. Stewart, 2018]

$$\begin{aligned} \delta m_Q^{\text{MSR}}(\mu, R) &\equiv m_Q^{\text{pole}} - m_Q^{\text{MSR}}(R) = R \sum_{n=1}^{\infty} \delta_n^R \left[ \frac{\alpha_s(R)}{4\pi} \right]^n \\ &= R \sum_{n=1}^{\infty} \left[ \frac{\alpha_s(\mu)}{4\pi} \right]^n \sum_{j=0}^{n-1} \delta_{nj}^R \log^j \left( \frac{\mu}{R} \right). \end{aligned}$$

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- **New type of large logs**  $\rightarrow$  resummation using R-evolution [A. H. Hoang, A. Jain, I. Scimemi and I. W. Stewart, 2008].

## Previous works

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- To see the advantages of R-evolution we go back to the first work that used the static energy to obtain  $\alpha_s$  [A. Bazavov, N. Brambilla, X. Garcia i Tormo et al, 2012]

$$E_s(r) = V_s(r, \mu_s, \mu_{us}) + \delta_{US}(r, \mu_s, \mu_{us}) + RS(\rho),$$

- $V_s(r, \mu_s, \mu_{us}) \rightarrow \ln(r\mu_{us})$
- $RS(\rho) \rightarrow \ln(\rho/\mu_{us})$
- There is no right choice for  $\mu_{us} \rightarrow$  fits only possible for small values of  $r$ .
- The force was used in [TUMQCD collaboration, 2019]

$$F_s(r) = \frac{dE_s(r)}{dr}$$

- Force avoids explicit subtraction.
- Using fully canonical scales  $\sim 1/r \rightarrow$  limits fit-range to  $r \sim 0.076$  fm.

# R-improvement, renormalon subtractions

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- The next goal is to obtain a renormalon-free potential.

$$E_{Q\bar{Q}}(r) = 2m_Q^{\text{MSR}}(R_0) + 2\delta m_Q^{\text{MSR}}(R_0, \mu) + E_s(r) \equiv 2m_Q^{\text{MSR}}(R_0) + V_s^{\text{MSR}}(r, \mu, R_0).$$

- We sum up large logs of  $\mu/R_0$ .

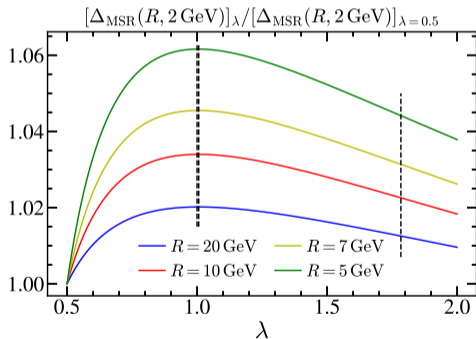
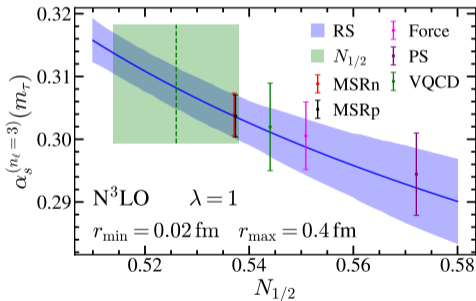
$$\begin{aligned}\delta m_Q^{\text{MSR}}(R_0) &= \delta m_Q^{\text{MSR}}(R_0) + \delta m_Q^{\text{MSR}}(R) - \delta m_Q^{\text{MSR}}(R) \\ &= \delta m_Q^{\text{MSR}}(R) + m_Q^{\text{MSR}}(R) - m_Q^{\text{MSR}}(R_0) \\ &= \delta m_Q^{\text{MSR}}(R) + \Delta^{\text{MSR}}(R, R_0).\end{aligned}$$

- $\Delta^{\text{MSR}}(R, R_0)$ : solution to MSR mass R-RGE  $\rightarrow$  sums up logs of  $R/R_0$ .
- We have  $\delta m_Q^{\text{MSR}}(R) \sim \log(\mu/R)$ . By choosing  $\mu \sim R \rightarrow$  no large logs.
- We define the R-improved static potential:

$$V_s^{\text{MSR}}(r, \mu, R_0) = V_s(r, \mu) + 2\delta^{\text{MSR}}(R, \mu) + 2\Delta^{\text{MSR}}(R, R_0).$$

# Dependence on $\lambda$

- We don't know the R-evolution kernel up to infinite order  $\rightarrow$  can estimate the truncation error with a dimensionless parameter  $\lambda$ , its variation account for higher-order remnants (similar to renormalization scale variation).
- $\alpha_s$  is sensitive to  $N_{1/2}$  (also depends on  $\lambda$ ), we have studied the R-evolution dependence with  $\lambda$ .



- The canonical value of  $\lambda = 1$  is clearly biased.
- We pick  $\lambda = 1.784$  and vary it from 1.5 to 2.1.

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- Solution: use profile functions that ensure series convergence:

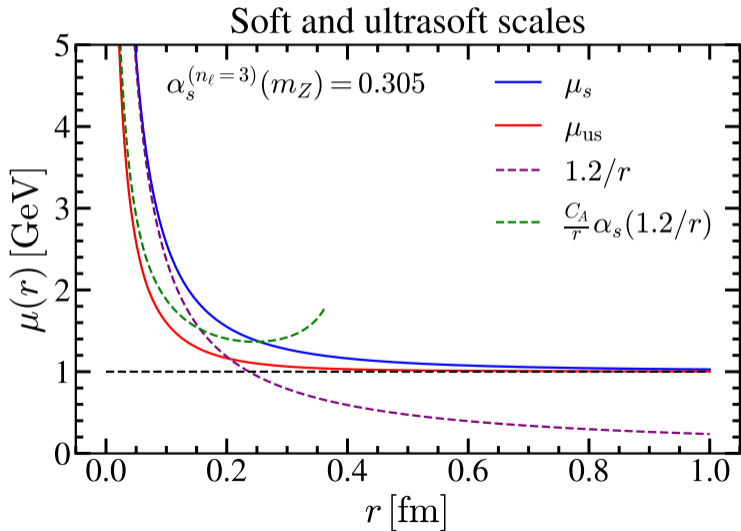
$$\mu_s(r, \xi, \mu_\infty, \Delta, b) = \sqrt{\left(\frac{\xi}{r}\right)^2 + \frac{b}{r} + (\mu_\infty - \Delta)^2} - \Delta = \begin{cases} \frac{\xi}{r} & \text{for } r \rightarrow 0 \\ \mu_\infty & \text{for } r \rightarrow \infty \end{cases}$$

$$R(r, \beta, R_\infty, \Delta, b) = \mu_s(r, \beta, R_\infty, \Delta, b)$$

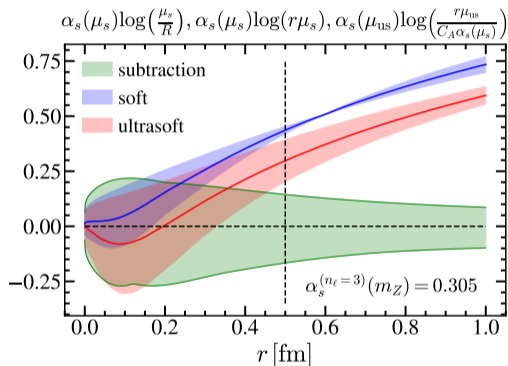
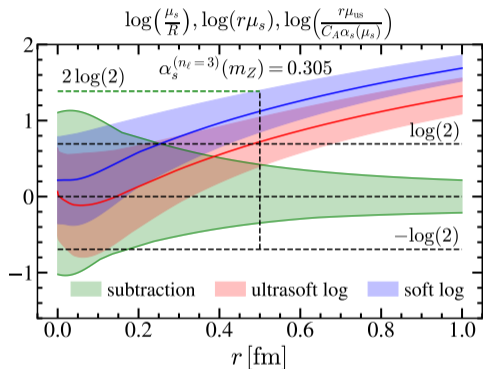
$$\mu_{\text{us}}(r, \xi, \kappa, \mu_\infty, \Delta, b) = C_A \left\{ \mu_s(r, \kappa\xi, \mu_\infty, \Delta, b) \alpha_s[\mu_s(r, \kappa\xi, \mu_\infty, \Delta, b)] - \mu_\infty \alpha_s(\mu_\infty) \right\} + \mu_\infty$$

with  $\xi = \mathcal{O}(1)$ ,  $\mu_\infty \sim 1 \text{ GeV}$  and  $\mu_s \gg \mu_{\text{us}}$  for small distances. This makes sure the series is stable and convergent over the entire spectrum.

# Profiles

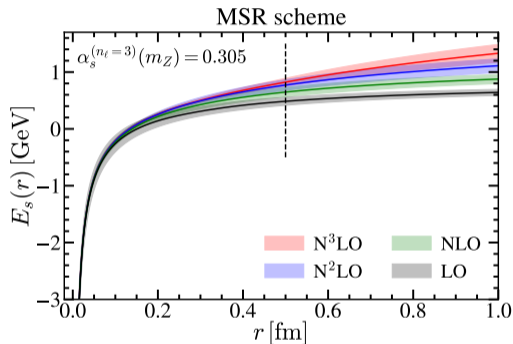
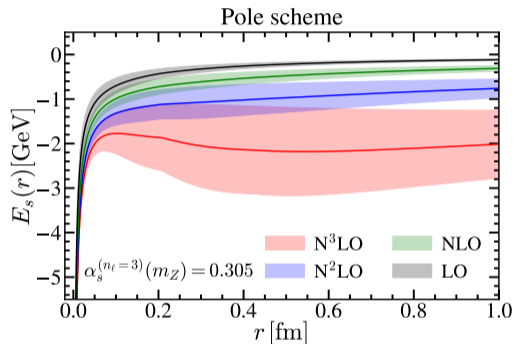


# Profiles



- Ultrasoft logarithm smaller than the soft one and  $\mathcal{O}(1)$  in the fit range.
- The product of  $\alpha_s(\mu) \times \log(\mu)$  is smaller than 1.

# Perturbative analysis



At  $r = 0.3$  fm we have the following uncertainties for LO, NLO, N<sup>2</sup>LO and N<sup>3</sup>LO:

- Pole scheme  $\rightarrow$  [0.16, 0.31, 0.71, 1.86] GeV.
- MSR scheme  $\rightarrow$  [0.17, 0.15, 0.13, 0.07] GeV.

# Fits

- Lattice QCD data from HotQCD. (2 + 1) flavor simulations. We have 9 lattices sets adding up to 2512 data points
- To change from lattice to physical units we use  $r_1 = 0.3093(20)$  fm, average of [A. Bazavov et al, 2014] and [R. Larsen, S. Mukherjee, P. Petreczky et al, arXiv 2502.08061].
- Each dataset  $n$  has a different origin of the static energy  $\rightarrow$  offset  $A_n$ , included in the  $\chi^2$  function:

$$\chi^2(\alpha_s, \{A_k\}) = \sum_{k=1}^{N_s} \chi_k^2(\alpha_s, A_k) \quad \chi_k^2(\alpha_s, A_k) = \sum_{i=1}^{n_{\text{data}}^k} \frac{[V_{ik}(\alpha_s) + A_k - V_{ik}^{\text{exp}}]^2}{[\sigma_k^2]_i}$$

- We can marginalize analytically first with respect to the offsets.

$$\frac{\partial \chi^2}{\partial A_k} = 0 \implies \tilde{A}_k(\alpha_s) = \frac{\sum_{i=1}^{n_{\text{data}}^k} [\sigma_k^{-1}]_i^2 [V_{ik}^{\text{exp}} - V_{ik}(\alpha_s)]}{\sum_{i=1}^{n_{\text{data}}^k} [\sigma_k^{-1}]_i^2},$$
$$\implies \tilde{\chi}_k^2(\alpha_s) = \chi_k^2(\alpha_s, \tilde{A}_k(\alpha_s)) - \frac{\left\{ \sum_{i=1}^{n_{\text{data}}^k} [\sigma_k^{-1}]_i [V_{ik}(\alpha_s) - V_{ik}^{\text{exp}}] \right\}^2}{\sum_{i=1}^{n_{\text{data}}^k} [\sigma_k^{-1}]_i}.$$

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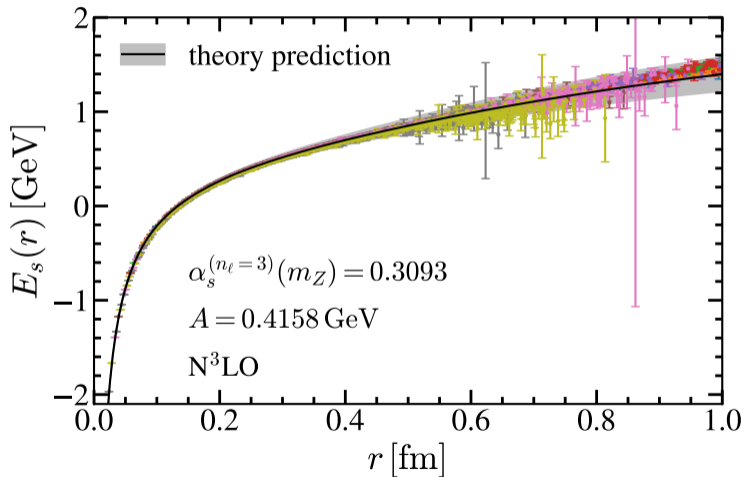
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- Minimizing we obtain  $\alpha_s^{\text{BF}}$  that verifies  $\tilde{\chi}^2(\alpha_s^{\text{BF}}) = \chi^2(\alpha_s^{\text{BF}}, A^{\text{BF}}) = \chi_{\min}^2$ .
- The fit uncertainty is defined as  $\tilde{\chi}^2(\alpha_s^{\text{BF}} \pm \Delta_{\text{fit}}\alpha_s) = \chi_{\min}^2 + 1$ .
- Theory uncertainties are estimated through a random scan.
- We perform the fit for 500 random profiles  $\rightarrow \alpha_s^{\text{BF}}$  and  $\Delta_{\text{fit}}\alpha_s$  for all of them.
- The theory (or perturbative) uncertainty is computed as

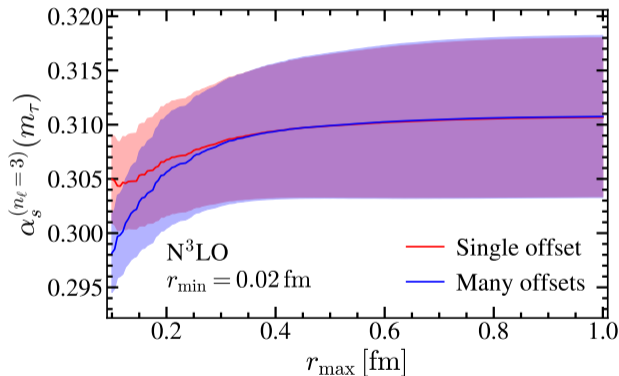
$$\Delta_{\text{pert}}\alpha_s = (\alpha_{\max} - \alpha_{\min})/2$$

- The final result is the mean of the 500  $\alpha_s^{\text{BF}}$  and  $\Delta_{\text{fit}}\alpha_s$ .
- We obtain  $\mathcal{O}(10)\chi^2$  values  $\rightarrow$  inflate lattice error by  $\sqrt{\chi_{\min}^2/\text{d.o.f.}}$ .
- Fit for  $r \in [r_{\min}, r_{\max}]$ .  $r_{\text{data}} \geq 0.023$  fm.
- $r_{\min} \in [0.02, 0.045]$  fm and  $r_{\max} \in [0.1, 1]$  fm with 0.05 fm spacing  $\rightarrow$  1086 different datasets.

# Static Energy

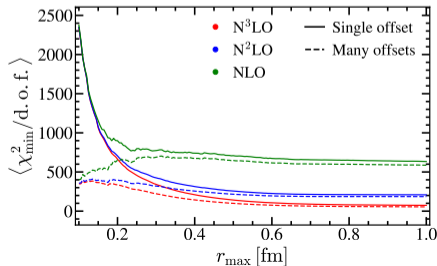
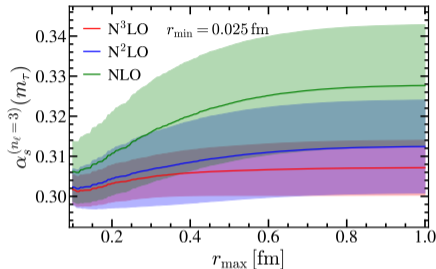
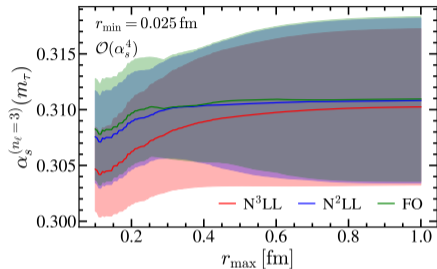


# Offset approach



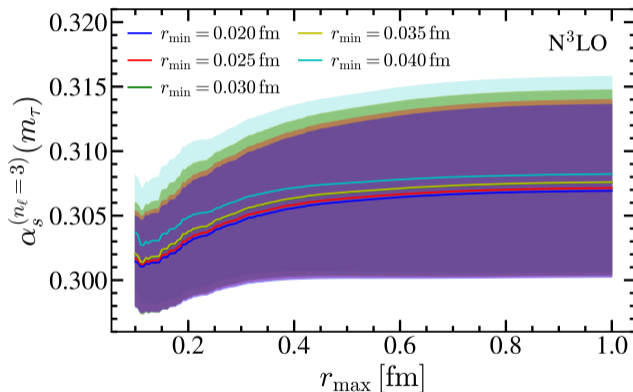
- We consider two offsets approaches:
  - Each ensemble has a different offset.
  - Common offset for all the ensembles.
- Both agree for  $r_{\max} > 0.3$  fm.
- Overparametrization of  $\chi^2$  at small distances for the many offset case.
- We consider  $r_{\max} \geq 0.35$  fm.

# Order-by-order agreement



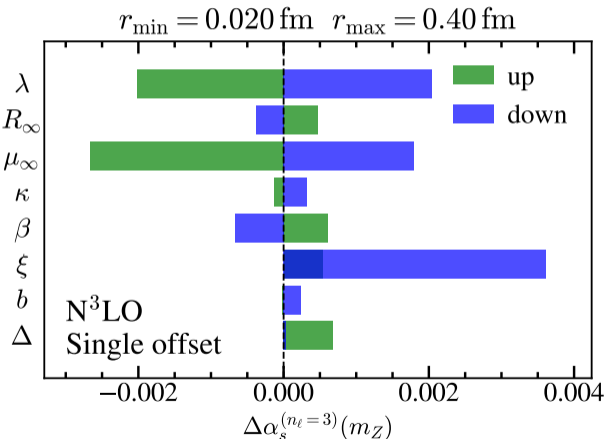
- Uncertainty bands nested for  $r_{\max} < 0.5$  fm.
- We choose  $r_{\max} \leq 0.45$  fm

# Study of the dataset



- We consider the range  $0.02 \leq r_{\min} \leq 0.04$  fm .

# Variation of profiles parameters one at a time



| Parameter    | Default              | Range                        |
|--------------|----------------------|------------------------------|
| $\xi$        | $1.2 \times \hbar c$ | $[0.7, 2.2] \times \hbar c$  |
| $\beta$      | $1.2 \times \hbar c$ | $[0.7, 2.2] \times \hbar c$  |
| $b$          | 0                    | $[-0.3, 0.3] \times \hbar c$ |
| $\Delta$     | 0                    | $[-0.6, 0.6] \text{ GeV}$    |
| $\mu_\infty$ | 1 GeV                | $[0.9, 1.1] \text{ GeV}$     |
| $R_\infty$   | 1 GeV                | $[0.9, 1.1] \text{ GeV}$     |
| $\kappa$     | 1                    | $[0.8, 1.2]$                 |
| $\lambda$    | 1.8                  | $[1.5, 2.1]$                 |

# Final results

- Requirements:  $r_{\min} \in [0.02, 0.04]$  fm and  $r_{\max} \in [0.35, 0.45]$  fm  $\rightarrow$  105-element dataset.
- We perform a fit in all of them:

$$\alpha_s^{(n_f=3)}(m_\tau) = 0.3093 \pm 0.00001_{\text{lattice}} \pm 0.0061_{\text{th}} \pm 0.0011_{\text{set}} \pm 0.0011_{r_1}$$

$\downarrow$

$$\alpha_s^{(n_f=3)}(m_\tau) = 0.3093 \pm 0.0063$$

$$\alpha_s^{(n_f=5)}(m_Z) = 0.1170 \pm 0.0008_{\text{th}} \pm 0.0001_{\text{set}} \pm 0.0001_{r_1} \pm 0.0003_{\mu_c} \pm 0.0002_{\mu_b}$$

$\downarrow$

$$\alpha_s^{(n_f=5)}(m_Z) = 0.1170 \pm 0.0009$$

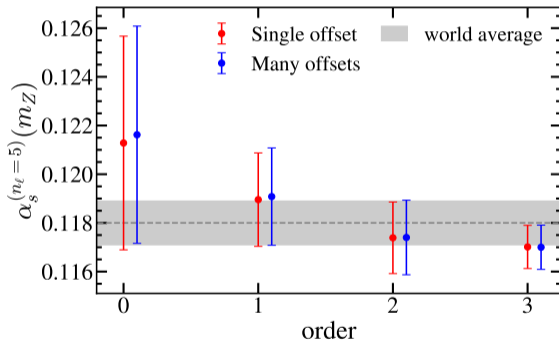
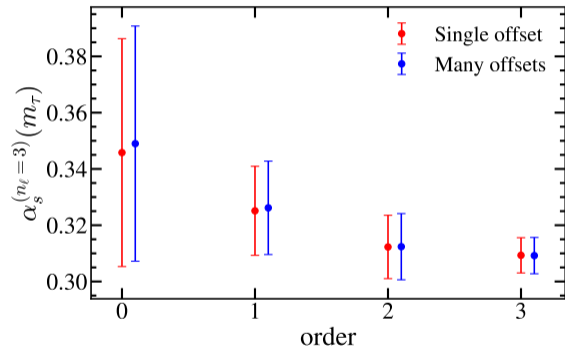
Competitive with the w.a.  $\alpha_s^{(5)}(m_Z) = 0.1180 \pm 0.0009$  and compatible at 0.5- $\sigma$  level.

Comparing with previous analyses:

$$\alpha_s^{(n_f=5)}(m_Z) = 0.11660_{-0.00056}^{+0.00110} \rightarrow \alpha_s^{(n_f=5)}(m_Z) = 0.1162(8) \text{ [TUMQCD collaboration, 2019]}$$

$$\alpha_s^{(n_f=5)}(m_Z) = 0.1181 \pm 0.0009 \rightarrow \alpha_s^{(n_f=5)}(m_Z) = 0.1177(9) \text{ [C. Ayala, X. Lobregat, A. Pineda, 2020]}$$

# Final results



# Conclusions

---

- Used the static energy to determine  $\alpha_s$ , building (and improving) on previous analyses.
- Performed ultra-soft large-log resummation up to N<sup>3</sup>LL.
- Employed the MSR mass and R-evolution to improve the static potential.
- Designed profile functions to increase the validity of the potential up to  $r \sim 0.5$  fm.
- Studied carefully the dependence on the dataset.
- Carried out fits to lattice data to obtain a very competitive result for  $\alpha_s$ .

$$\alpha_s^{(n_f=5)}(m_Z) = 0.3093 \pm 0.0063 \quad \alpha_s^{(n_f=5)}(m_Z) = 0.1170 \pm 0.0009$$

BACKUP

# Force-type subtractions

---

- Integrating the Force is equal to perform R-evolution.
- The renormalon doesn't depend on  $r \rightarrow$  we can subtract the potential at  $r_0$ .

$$\begin{aligned} E_s^F(r, r_0) &\equiv E_s(r) - E_s(r_0) = E_s(r) - E_s(r_1) + [E_s(r_1) - E_s(r_0)] \\ &= E_s(r) - E_s(r_1) + \int_{r_0}^{r_1} dr' F_s(r') \equiv E_s(r) - E_s(r_1) + \Delta_F(r_0, r_1) \end{aligned}$$

- [A. Bazavov, N. Brambilla, X. Garcia i Tormo et al, 2014] chooses  $r_1 = r \rightarrow$  only  $\Delta_F$  left.
- Connecting with R-evolution, the subtraction term is (choosing  $R = 1/r_1$  and  $\mu = R$ ):

$$E_s(r_1) \equiv \delta_{\text{soft}}^F(R) = \frac{1}{2} V_s^{\text{soft}}\left(\frac{1}{R}\right) = -2\pi C_F R \sum_{i=1} \left[ \frac{\alpha_s(R)}{4\pi} \right]^i \sum_{j=0}^{i-1} a_{i-1,j} \gamma_E^j \equiv R \sum_{i=1} \left[ \frac{\alpha_s(R)}{4\pi} \right]^i \delta_i^F.$$

- We can express  $\Delta_F$  as an R-evolution integral:

$$\begin{aligned} \gamma_{\text{soft}}^F(R) &= -\frac{1}{2} \left[ r^2 F_s^{\text{soft}}(r) \right]_{r=1/R}, \\ \Delta_F^{\text{soft}}(r_0, r_1) &= \int_{r_0}^{r_1} \frac{dr'}{(r')^2} \left[ (r')^2 F_s^{\text{soft}}(r') \right] = \frac{1}{2} \int_{1/r_0}^{1/r_1} dR' \gamma_{\text{soft}}^F(R'). \end{aligned}$$

- It inherits the infrared sensitivity of the Static Potential.

# Potential-type subtractions

---

- If we choose  $\mu = R$  and  $R = \frac{e^{-\gamma E}}{r}$

$$\delta_{\text{soft}}^V(R) \equiv V_s^{\text{soft}}\left(\frac{e^{-\gamma E}}{R}\right) = -2\pi e^{\gamma E} C_F R \sum_{i=1} \left[\frac{\alpha_s(R)}{4\pi}\right]^i a_{i-1,0} \equiv R \sum_{i=1} \left[\frac{\alpha_s(R)}{4\pi}\right]^i \delta_i^V.$$

- We can obtain their anomalous dimensions

$$\gamma_{\text{soft}}^V(R) = -\frac{1}{2} \left[ r^2 V_s^{\text{soft}}(r) \right]_{r=1/R}$$
$$\Delta_V^{\text{soft}}(r_0, r_1) = \int_{r_0}^{r_1} \frac{dr'}{(r')^2} \left[ (r')^2 V_s^{\text{soft}}(r') \right] = \frac{1}{2} \int_{1/r_0}^{1/r_1} dR' \gamma_{\text{soft}}^V(R')$$

# Subtraction schemes: PS scheme

---

- Defined from its relation to the pole mass:  $m_p - m^{\text{PS}}(R) \equiv \delta_{\text{PS}}(R)$

$$\delta_{\text{soft}}^{\text{PS}}(R) = -\frac{1}{2} \int_{|\vec{q}| < R} \frac{d^3 \vec{q}}{(2\pi)^3} \tilde{V}_s^{\text{soft}}(q, R) \equiv C_F R \frac{\alpha_s(R)}{\pi} \sum_{i=0} \left[ \frac{\alpha_s(R)}{4\pi} \right]^i c_i,$$

$$c_i = \sum_{j=0}^i a_{i,j} h_j, \quad \text{with} \quad h_j = j! \sum_{\ell=0}^j \sum_{k=0}^{\text{floor}[\frac{\ell}{2}]} \kappa_{\ell-2k} \left(\frac{\pi}{2}\right)^{2k} \frac{(-1)^k}{(2k)!}.$$

# Subtraction schemes: ultrasoft terms

---

- Since all these three schemes come from the potential, they have ultrasoft terms in their R-evolution.

$$\gamma^F(R) = \gamma_{\text{soft}}^F(R) - \frac{C_A^3 C_F}{24\pi} [\alpha_s(R)]^4 \log \left[ C_A \alpha_s(R) e^{\gamma_E - 5/6} \right],$$

$$\gamma^V(R) = \gamma_{\text{soft}}^V(R) - \frac{C_A^3 C_F}{24\pi} e^{\gamma_E} [\alpha_s(R)]^4 \log \left[ C_A \alpha_s(R) e^{\gamma_E - 5/6} \right],$$

$$\gamma^{\text{PS}}(R) = \gamma_{\text{soft}}^{\text{PS}}(R) - \frac{C_F C_A^3}{12\pi^2} [\alpha_s(R)]^4 \log \left[ C_A \alpha_s(R) e^{\gamma_E - 5/6} \right].$$

- They inherit the infrared problem of the static potential.

# RS scheme

---

- It is defined from the pole mass by subtracting its leading asymptotic behavior [Antonio Pineda, 2001].

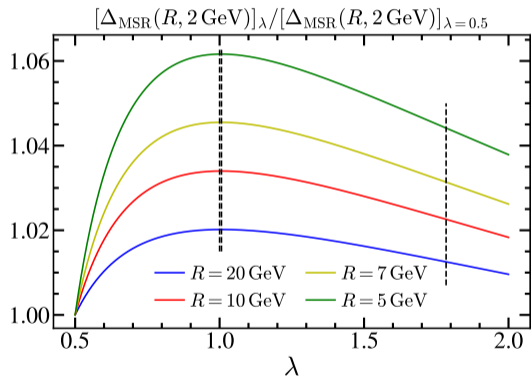
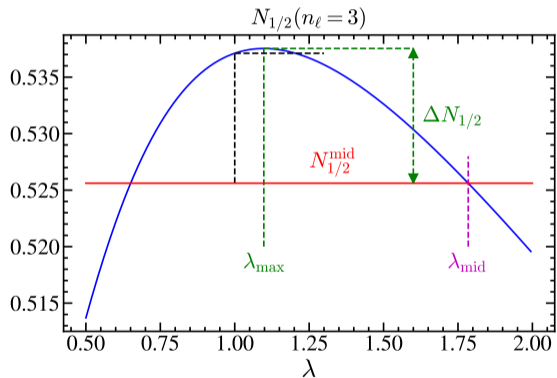
$$m_Q^{\text{pole}} - m_Q^{\text{RS}}(R) = \frac{2\pi}{\beta_0} RN_{1/2} \sum_{n=1}^{\infty} \left[ \frac{\beta_0 \alpha_s(R)}{2\pi} \right]^n \sum_{\ell=0}^{\infty} g_\ell \left(1 + \hat{b}_1\right)_{n-1-\ell},$$

- $N_{1/2}$  is the normalization of the leading renormalon

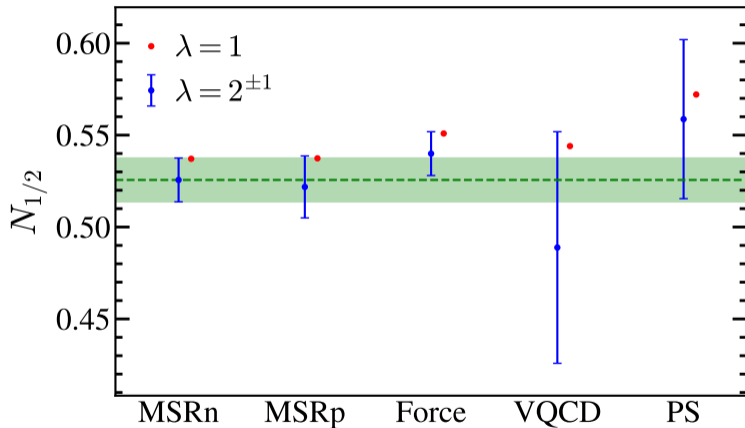
$$N_{1/2}^{(n)} = \frac{\beta_0}{2\pi} \sum_{k=0}^n \frac{S_k}{\left(1 + \hat{b}_1\right)_k} \quad S_k = \sum_{k=0}^j \tilde{\gamma}_k^R \sum_{i=0}^{j-k} (-1)^i \tilde{b}_i^N \tilde{g}_{j-i-k}^N,$$

- $N_{1/2}$  depends on  $\lambda$ , it reshuffles higher perturbative orders. We vary it to estimate  $N_{1/2}$  (similar to scale variation).
- $\lambda$  is also used to estimate R-evolution uncertainty.

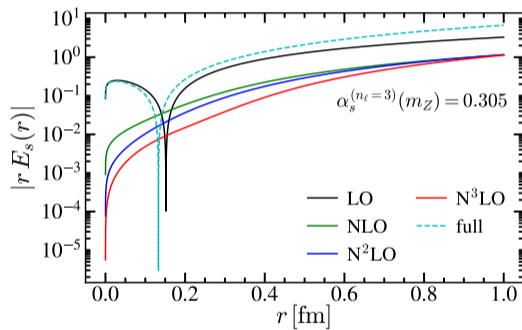
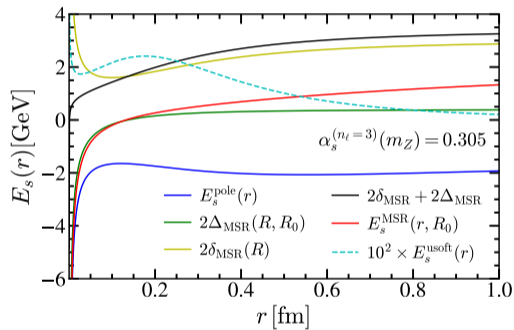
# $N_{1/2}$



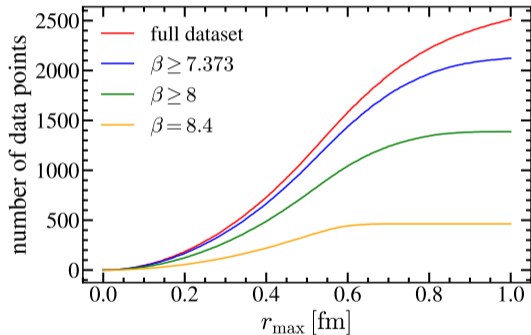
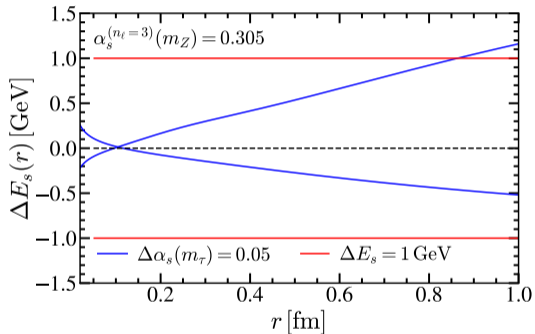
# Renormalon subtraction



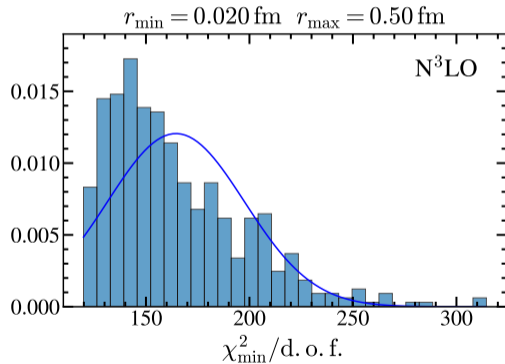
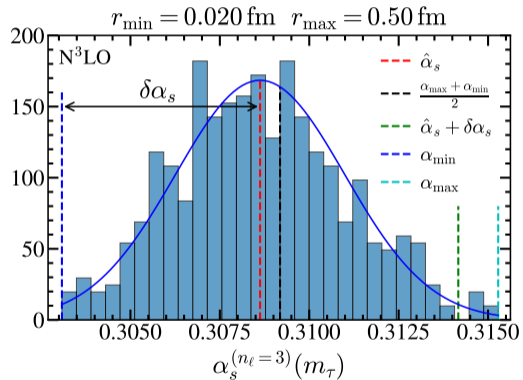
# Perturbative analysis



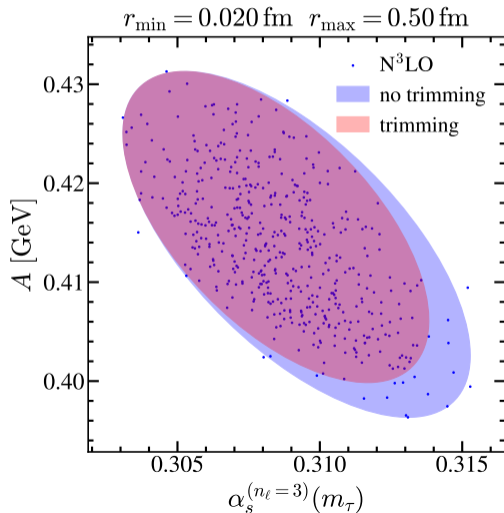
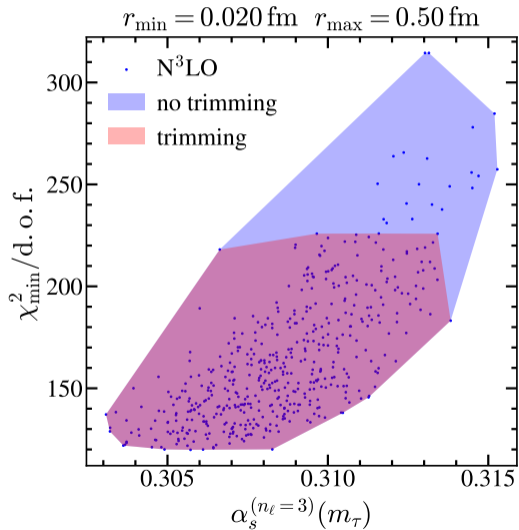
# Auxiliar figures



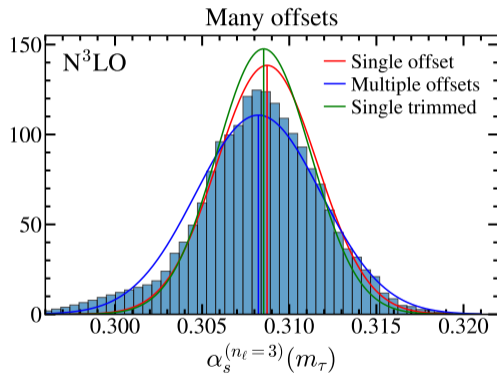
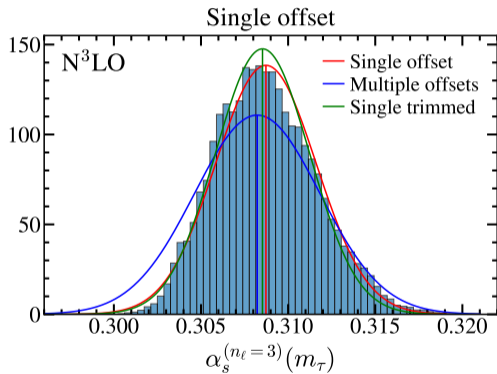
# Random scan histograms



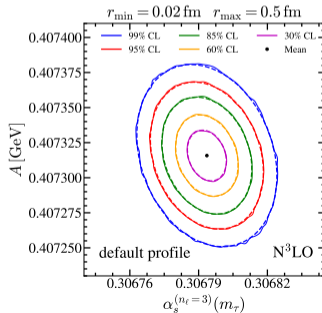
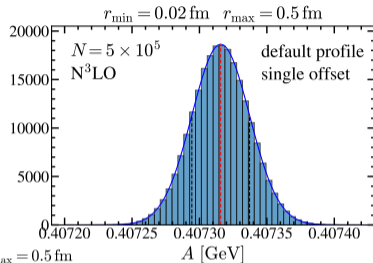
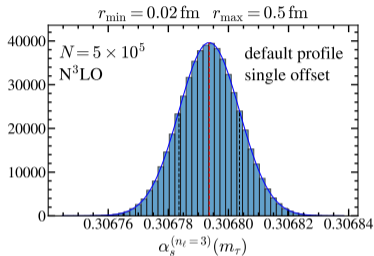
# $\chi^2$ trimming



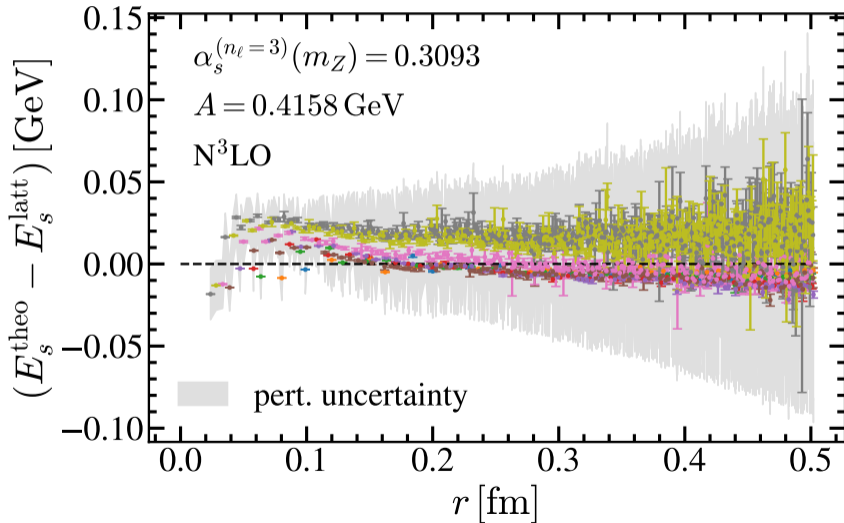
# Fits histograms



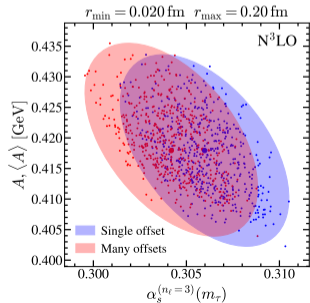
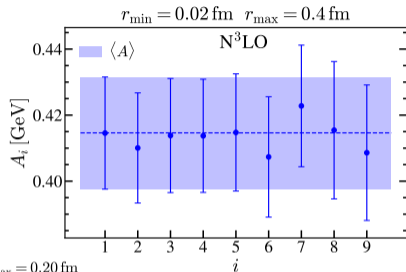
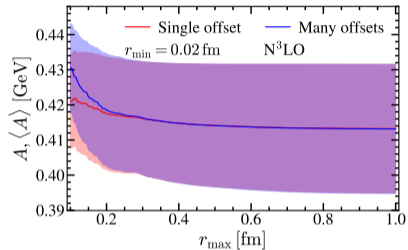
# Replica method



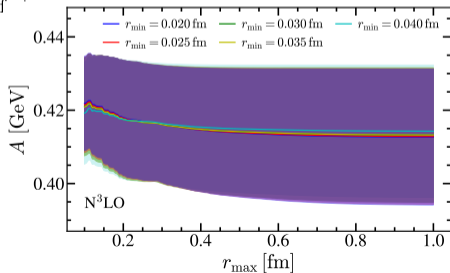
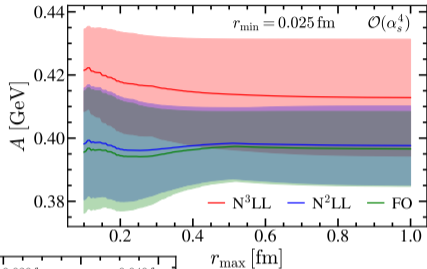
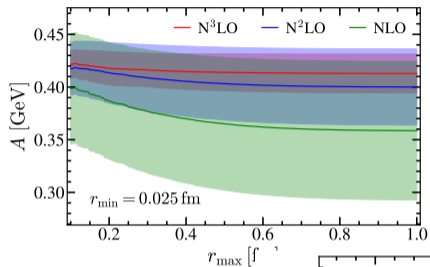
# Lattice data



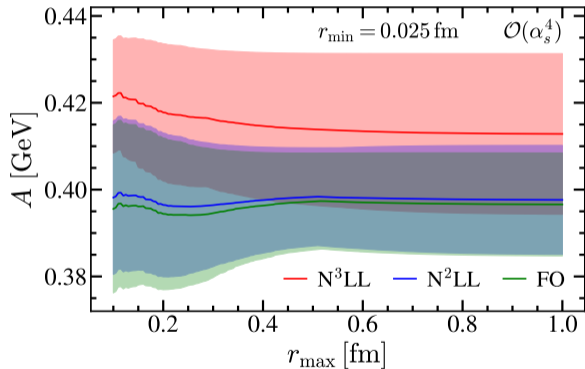
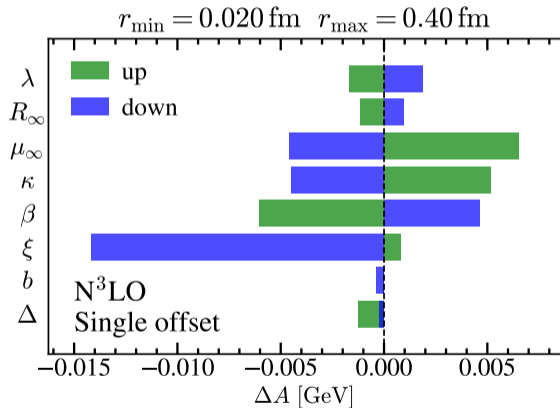
# Offset analysis



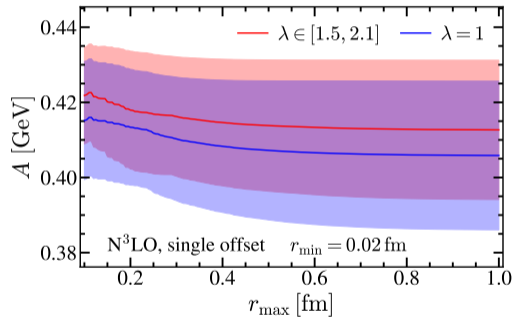
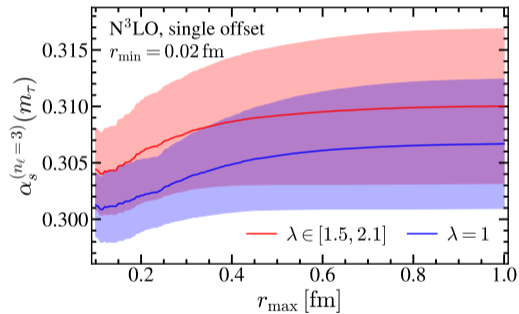
# Offset analysis



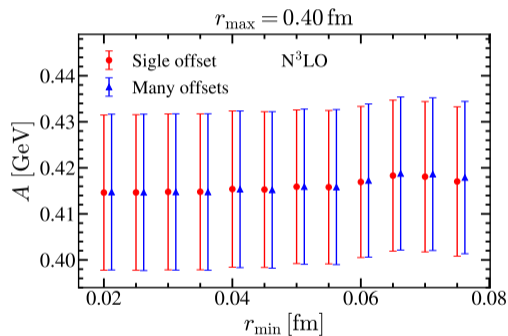
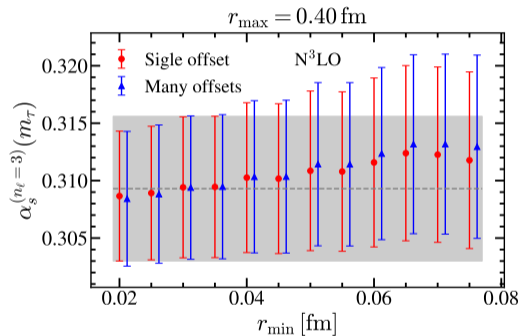
# Offset analysis



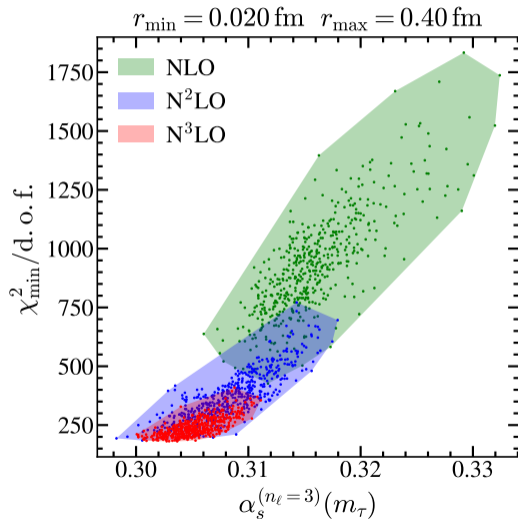
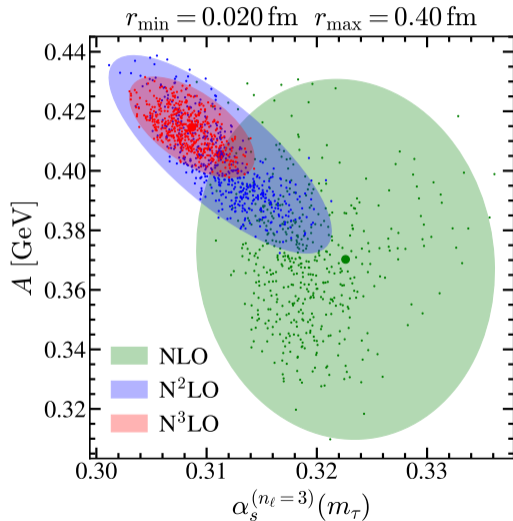
# $\lambda$ variation



# Further dependence check on $r_{min}$



# Scattered fit points



## Additional fits

---

|                                 | full                | $\beta \geq 7.737$  | $\beta \geq 8$      | $\beta = 8.4$       |
|---------------------------------|---------------------|---------------------|---------------------|---------------------|
| $\alpha_s^{(n_\ell=3)}(m_\tau)$ | $0.3093 \pm 0.0063$ | $0.3089 \pm 0.0061$ | $0.3089 \pm 0.0070$ | $0.3080 \pm 0.0072$ |
| $\alpha_s^{(n_\ell=5)}(m_Z)$    | $0.1170 \pm 0.0009$ | $0.1170 \pm 0.0009$ | $0.1170 \pm 0.0010$ | $0.1169 \pm 0.0010$ |

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|                                 | $\mu_c = \bar{m}_c$ | $\lambda = 1$       | FO                  | many offsets        |
|---------------------------------|---------------------|---------------------|---------------------|---------------------|
| $\alpha_s^{(n_\ell=3)}(m_\tau)$ | $0.3093 \pm 0.0063$ | $0.3055 \pm 0.0049$ | $0.3108 \pm 0.0057$ | $0.3092 \pm 0.0064$ |
| $\alpha_s^{(n_\ell=5)}(m_Z)$    | $0.1174 \pm 0.0009$ | $0.1165 \pm 0.0007$ | $0.1172 \pm 0.0008$ | $0.1170 \pm 0.0009$ |