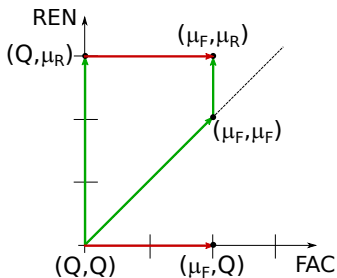


Recurrence relations for scale dependence in QCD



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preprint in preparazione con Matteo Cacciari ed Emanuele Bagnaschi



Structure of QCD prediction

Running coupling and β -function

Structure of QCD prediction

μ_R scale dependence

Recurrence relation for μ_R

An example: running coupling

μ_F scale dependence

μ_F scale and DGLAP equation

Moving in the $\mu_F - \mu_R$ plane

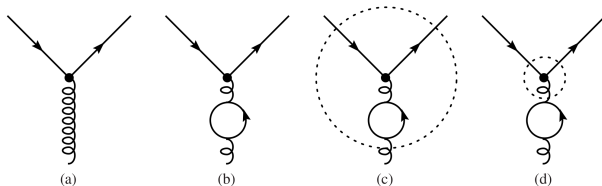
An example: Deep Inelastic Scattering

Conclusions



Renormalization scale in QCD

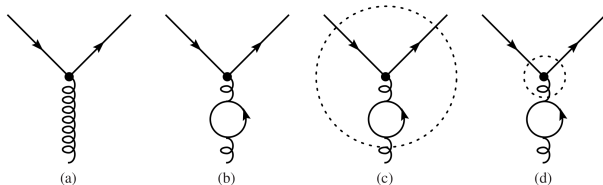
Beyond tree level, we find **divergent integrals** in the UV region



→ a **scale** μ_R appears in the renormalization procedure

Renormalization scale in QCD

Beyond tree level, we find **divergent integrals** in the UV region



→ a **scale** μ_R appears in the renormalization procedure
 Evolution of the coupling α_s with μ_R → **QCD β -function**

$$\frac{d\alpha_s}{d \ln \mu_R^2} = \beta(\alpha_s) = -(\beta_0 \alpha_s^2 + \beta_1 \alpha_s^3 + \dots), \quad \beta_0 = \frac{33 - 2n_f}{12\pi}$$

Residual scale dependence

Expansion of an observable σ in a perturbation series in the coupling α_s evaluated at μ_R :

$$\sigma(Q) = \sum_{n=0}^{\infty} c_n \alpha_s^n$$

σ independent from μ_R , but true only for an *all-order* prediction! \rightarrow fixed-order prediction retains a **scale dependence**

$$\sigma^{(k)}(Q, \mu_R) = \sum_{n=0}^k c_n(Q, \mu_R) \alpha_s^n(\mu_R)$$

The residual scale dependence is proved to be of $\mathcal{O}(\alpha_s^{k+1})$:

$$\sigma^{(k)}(Q, \mu_R) = \sigma^{(k)}(Q, Q) + \mathcal{O}(\alpha_s^{k+1})$$

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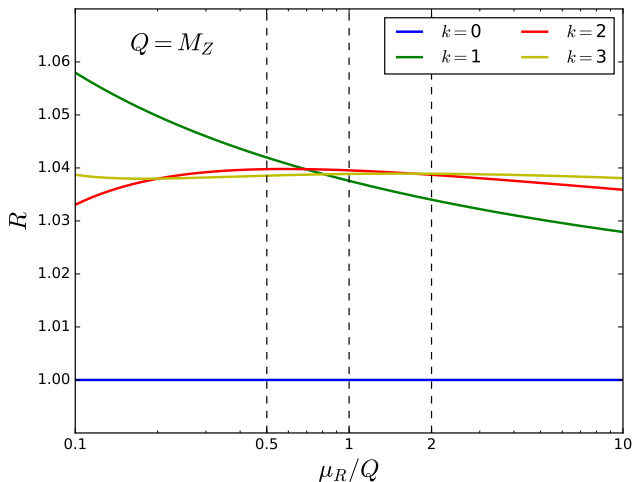
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Example of scale dependence

$$R = \sigma_{e^+e^- \rightarrow \text{had}}(Q) / \sigma_{e^+e^- \rightarrow \mu^+\mu^-}(Q) \propto 1 + c_1 \alpha_s(Q) + c_2 \alpha_s^2(Q) + \dots$$



Renormalization scale dependence

Observable σ expanded in perturbation series:

$$\sigma(Q) = \sum_{n=0}^{\infty} c_n(Q, Q) \alpha_s^n(Q) = \sum_{n=0}^{\infty} c_n(Q, \mu_R) \alpha_s^n(\mu_R)$$

Once given $c_n(Q, Q)$, how can we find $c_n(Q, \mu_R)$? i.e. how can we shift the renormalization scale from Q to a generic μ_R ?

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- Imposing:

$$\frac{d}{d \ln \mu_R^2} \sigma^{(k)}(Q) = \frac{d}{d \ln \mu_R^2} \sum_{n=0}^k c_n(Q, \mu_R) \alpha_s^n(\mu_R) = 0,$$

and neglecting higher order terms which appear in the calculation;

Renormalization scale dependence

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Once given $c_n(Q, Q)$, how can we find $c_n(Q, \mu_R)$? i.e. how can we shift the renormalization scale from Q to a generic μ_R ?

- Solving the β -function up to some order and expanding the result around the coupling evaluated at μ_R^2 :

$$\alpha_s(Q^2) = \alpha_s(\mu_R^2) + \beta_0 L \alpha_s^2(\mu_R^2) + \{\beta_1 L + \beta_0^2 L^2\} \alpha_s^3(\mu_R^2) + \dots$$

where $L = \ln(\mu_R^2/Q^2)$, and then substituting this expression in $\sigma(Q)$.

Renormalization scale dependence

Observable σ expanded in perturbation series:

$$\sigma(Q) = \sum_{n=0}^{\infty} c_n(Q, Q) \alpha_s^n(Q) = \sum_{n=0}^{\infty} c_n(Q, \mu_R) \alpha_s^n(\mu_R)$$

Once given $c_n(Q, Q)$, how can we find $c_n(Q, \mu_R)$? i.e. how can we shift the renormalization scale from Q to a generic μ_R ?

- Working with an *all-order* expression and imposing:

$$\frac{d}{d \ln \mu_R^2} \sigma(Q) = 0,$$

using identities well known in statistical mechanics:

$$\sum_{a=\lambda}^{\infty} \alpha^a \sum_{b=0}^{\infty} f_{a,b} \alpha^{b+1} = \sum_{a=\lambda+1}^{\infty} \alpha^a \sum_{b=\lambda}^{a-1} f_{b,a-1-b}$$

→ **recurrence relation!**

Recurrence relation for μ_R scale dependence

[arXiv:1105.5152]

The $c_n(Q, \mu_R)$ coefficient is given by a polynomial of degree less than or equal to $n - 1$ in the $\ln(\mu_R^2/Q^2)$ variable:

$$c_n(Q, \mu_R) = \sum_{l=0}^{n-1} c_{n,l} \left(\ln \frac{\mu_R^2}{Q^2} \right)^l \quad (1)$$

with $c_{n,l}$ given by the following recurrence relation:

$$c_{n,l} = \begin{cases} c_{n,0} = c_n(Q, Q) \\ c_{n,l} = \frac{1}{l} \sum_{m=l-1}^{n-1} m \beta_{n-1-m} c_{m,l-1} \end{cases} \quad (2)$$

An example: running coupling

$$\alpha_s(\mu) = \sum_{n=0}^{\infty} c_n(\mu, \mu) \alpha_s(\mu) = \sum_{n=0}^{\infty} c_n(\mu, \mu_0) \alpha_s(\mu_0)$$

with $c_n(\mu, \mu) = \delta_{n1}$ as initial condition. First terms, with $L = \ln(\mu_0^2/\mu^2)$:

$$\begin{aligned} \alpha_s(\mu^2) &= \alpha_s(\mu_0^2) + \beta_0 L \alpha_s^2(\mu_0^2) + \{\beta_1 L + \beta_0^2 L^2\} \alpha_s^3(\mu_0^2) \\ &+ \left\{ \beta_2 L + \frac{5}{2} \beta_0 \beta_1 L^2 + \beta_0^3 L^3 \right\} \alpha_s^4(\mu_0^2) \\ &+ \left\{ \beta_3 L + \frac{3}{2} (\beta_1^2 + 2\beta_0 \beta_2) L^2 + \frac{13}{3} \beta_0^2 \beta_1 L^3 + \beta_0^4 L^4 \right\} \alpha_s^4(\mu_0^2) \\ &+ \mathcal{O}(\alpha_s^5) \end{aligned}$$

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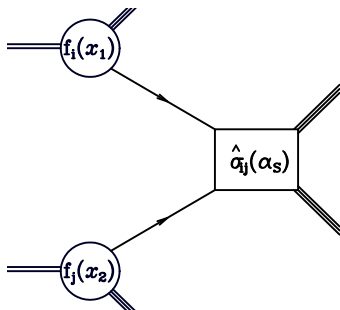
with $c_n(\mu, \mu) = \delta_{n1}$ as initial condition. First terms, with $L = \ln(\mu_0^2/\mu^2)$: **check: if $\mu < \mu_0$, then $\alpha_s(\mu^2) > \alpha_s(\mu_0^2)$**

$$\begin{aligned} \alpha_s(\mu^2) &= \alpha_s(\mu_0^2) + \beta_0 L \alpha_s^2(\mu_0^2) + \{\beta_1 L + \beta_0^2 L^2\} \alpha_s^3(\mu_0^2) \\ &+ \left\{ \beta_2 L + \frac{5}{2} \beta_0 \beta_1 L^2 + \beta_0^3 L^3 \right\} \alpha_s^4(\mu_0^2) \\ &+ \left\{ \beta_3 L + \frac{3}{2} (\beta_1^2 + 2\beta_0 \beta_2) L^2 + \frac{13}{3} \beta_0^2 \beta_1 L^3 + \beta_0^4 L^4 \right\} \alpha_s^4(\mu_0^2) \\ &+ \mathcal{O}(\alpha_s^5) \end{aligned}$$

$$\sigma(Q) = \sum_{ij} \underbrace{\hat{\sigma}_{ij}(Q)}_{\text{Partonic cross section}} \otimes$$

Parton Distribution Functions (PDFs)

$$\underbrace{f_i \otimes f_j}$$



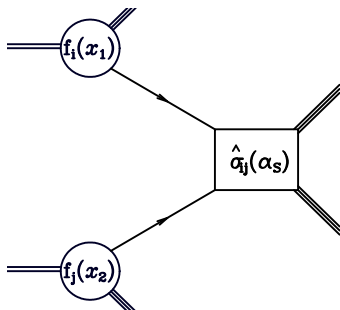
“Multiplicative” convolution:

$$(a \otimes b)(x) = \int_x^1 \frac{dy}{y} a(y) b\left(\frac{x}{y}\right)$$

$$a \otimes b = b \otimes a$$

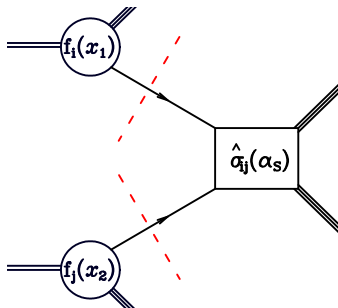
$$(a \otimes b) \otimes c = a \otimes (b \otimes c)$$

$$\sigma(Q) = \sum_{ij} \underbrace{\hat{\sigma}_{ij}(Q)}_{\text{Partonic cross section}} \otimes \underbrace{f_i \otimes f_j}_{\text{Parton Distribution Functions (PDFs)}}$$



Divergence due to **collinear** emission of gluon

$$\sigma(Q) = \sum_{ij} \underbrace{\hat{\sigma}_{ij}(Q, \mu_F)}_{\text{Partonic cross section}} \otimes \underbrace{f_i(\mu_F) \otimes f_j(\mu_F)}_{\text{Parton Distribution Functions (PDFs)}}$$



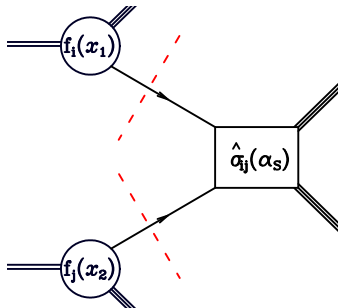
Divergence due to collinear emission of gluon

↓
absorbed into the PDF f
at a scale μ_F

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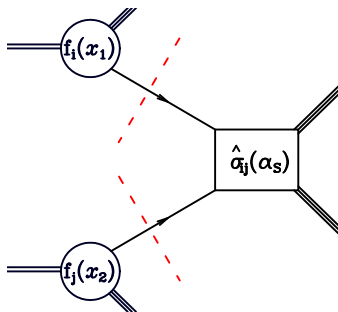


absorbed into the PDF f
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Evolution of PDF given by
DGLAP equation

$$\sigma(Q) = \sum_{ij} \underbrace{\hat{\sigma}_{ij}(Q, \mu_F)}_{\text{Partonic cross section}} \otimes \overbrace{f_i(\mu_F) \otimes f_j(\mu_F)}^{\text{Parton Distribution Functions (PDFs)}}$$



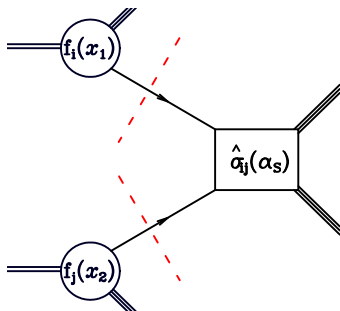
Divergence due to **collinear** emission of gluon

↓
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↓
Evolution of PDF given by **DGLAP equation**
 $f = q, \bar{q}, g$

$$\frac{d}{d \ln \mu_F^2} \begin{pmatrix} q_i \\ g \end{pmatrix} = \sum_{q_k, \bar{q}_k} \begin{pmatrix} P_{ik}(\alpha_s(\mu_F)) & P_{ig}(\alpha_s(\mu_F)) \\ P_{gk}(\alpha_s(\mu_F)) & P_{gg}(\alpha_s(\mu_F)) \end{pmatrix} \otimes \begin{pmatrix} q_k \\ g \end{pmatrix}$$

$$\sigma(Q) = \sum_{ij} \left(\sum_{n=0}^{\infty} \hat{\sigma}_{ij}^{(n)}(Q, \mu_F, \mu_R) \alpha_s^{(n+\lambda)}(\mu_R) \right) \otimes f_i(\mu_F) \otimes f_j(\mu_F)$$



Divergence due to collinear emission of gluon



absorbed into the PDF f
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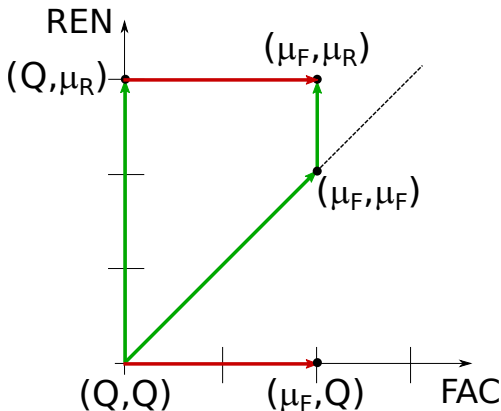
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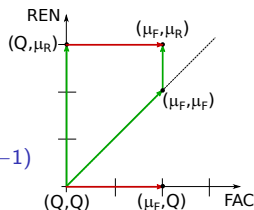
$$\hat{\sigma}_{ij}^{(n)}(Q, Q, Q) \rightarrow \hat{\sigma}_{ij}^{(n)}(Q, \mu_F, \mu_F) \rightarrow \hat{\sigma}_{ij}^{(n)}(Q, \mu_F, \mu_R)$$

$$\hat{\sigma}_{ij}^{(n)}(Q, \mu_F, \mu_F) = \sum_{l=0}^n \hat{\sigma}_{ij}^{(n,l)} \left(\ln \frac{\mu_F^2}{Q^2} \right)^l$$

$$\hat{\sigma}_{ij}^{(n,l)} = \frac{1}{l} \sum_{m=l-1}^{n-1} \beta_{n-1-m} (m + \lambda) \hat{\sigma}_{ij}^{(m,l-1)}$$

$$- \left\{ \sum_k \hat{\sigma}_{jk}^{(m,l-1)} \otimes P_{ki}^{(n-1-m)} + i \leftrightarrow j \right\}$$

$$\hat{\sigma}_{ij}^{(n,0)} \equiv \hat{\sigma}_{ij}^{(n)}(Q, Q, Q)$$



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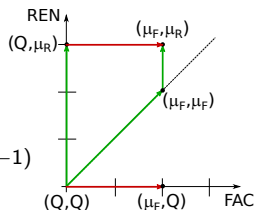
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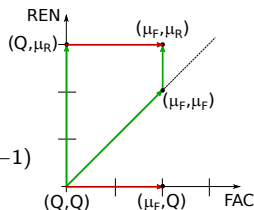
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$$\hat{\sigma}_{ij}^{(n,0)} \equiv \hat{\sigma}_{ij}^{(n)}(Q, Q, Q)$$

→ only symbolic, they must be properly convoluted with PDFs



An example: short-distance coefficients in DIS

Only one convolution (here q is quark singlet PDF):

$$F(Q) \supset \sum_{n=0}^{\infty} C_q^{(n)}(Q, \mu_F, \mu_R) \alpha_s^n(\mu_R) \otimes q(\mu_F)$$

First terms, with $L_M = \ln(\mu_F^2/Q^2)$ and $L_R = \ln(\mu_R^2/\mu_F^2)$:

$$C_q^{(0)}(Q, \mu_R, \mu_F) = c_q^{(0,0)}$$

$$C_q^{(1)}(Q, \mu_R, \mu_F) = c_q^{(1,0)} + L_M (c_q^{(0,0)} \otimes P_{qq}^{(0)})$$

$$\begin{aligned} C_q^{(2)}(Q, \mu_R, \mu_F) = & c_q^{(2,0)} + L_M [c_q^{(0,0)} \otimes P_{qq}^{(1)} + c_q^{(1,0)} \otimes (P_{qq}^{(0)} - \beta_0 \mathbb{I}) + c_g^{(1,0)} \otimes P_{gq}^{(0)}] \\ & + L_M^2 \left[\frac{1}{2} (c_q^{(0,0)} \otimes P_{qq}^{(0)}) \otimes (P_{qq}^{(0)} - \beta_0 \mathbb{I}) + \frac{1}{2} (c_q^{(0,0)} \otimes P_{gq}^{(0)}) \otimes P_{gq}^{(0)} \right] \\ & + L_R \beta_0 c_q^{(1,0)} + L_R L_M \beta_0 c_q^{(0,0)} \otimes P_{qq}^{(0)} \end{aligned}$$

where $c_i^{(n,0)} \equiv C_i^{(n)}(Q, Q, Q)$ (note that $c_g^{(0)} = 0$).

Conclusions

- Recurrence relations:
 - general way to obtain scale dependence;
 - straightforward implementation into any HEP generator;
 - checked against literature (e.g. top pair production);
 - code in preparation with a series of examples and *ready-to-use* scale dep. for some important observables.
- Applications?
 - theoretical uncertainty estimates;
 - useful for studies about scale choice.