

# Gluon Polarisation Measurements

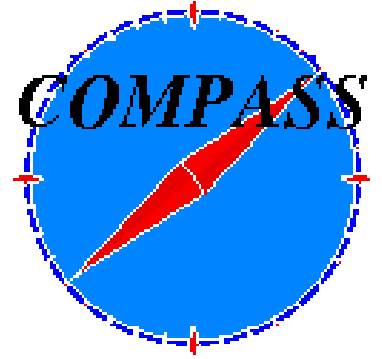
## @ COMPASS

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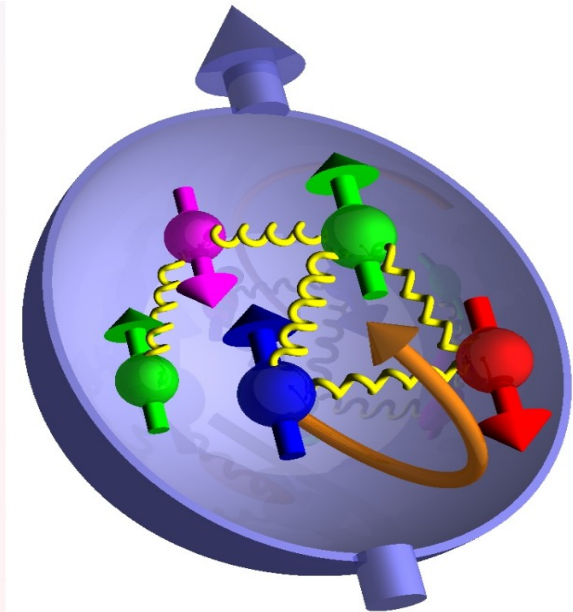
### Outline:

- Brief Motivation
- High  $p_T$  analysis
- Open Charm (LO and NLO) analyses
- $\Delta G/G$  results
- Summary and Conclusion



On behalf of the COMPASS Collaboration

# The Nucleon Spin



$$S_N = 1/2 = 1/2 \Delta\Sigma + \Delta G + L$$

Partons  
Angular  
Orbital  
Momenta

Quarks

Gluons

Future !  
GPDs

## Well known !

In 1988 EMC measured

$$\Delta\Sigma = 0.12 \pm 0.17 \text{ (Phys.Lett.B206,364)}$$

A recent result, including COMPASS, gives:

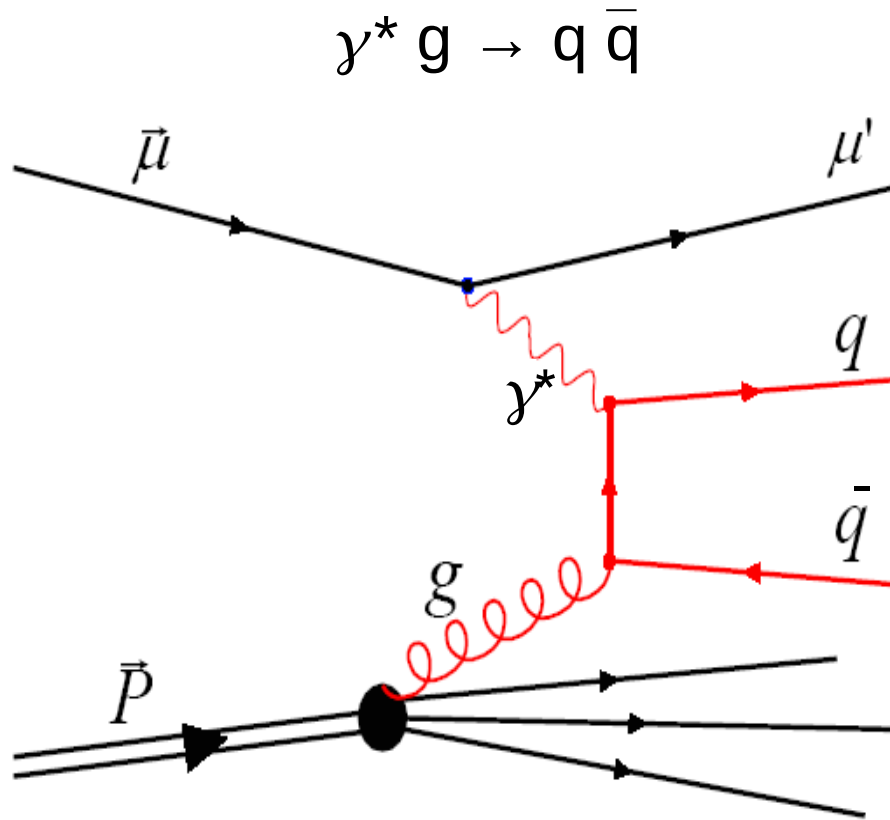
$$\Delta\Sigma = 0.30 \pm 0.01(\text{stat.}) \pm 0.02(\text{evol.}) \text{ Phys.Lett.B647,8}$$

## Poorly known

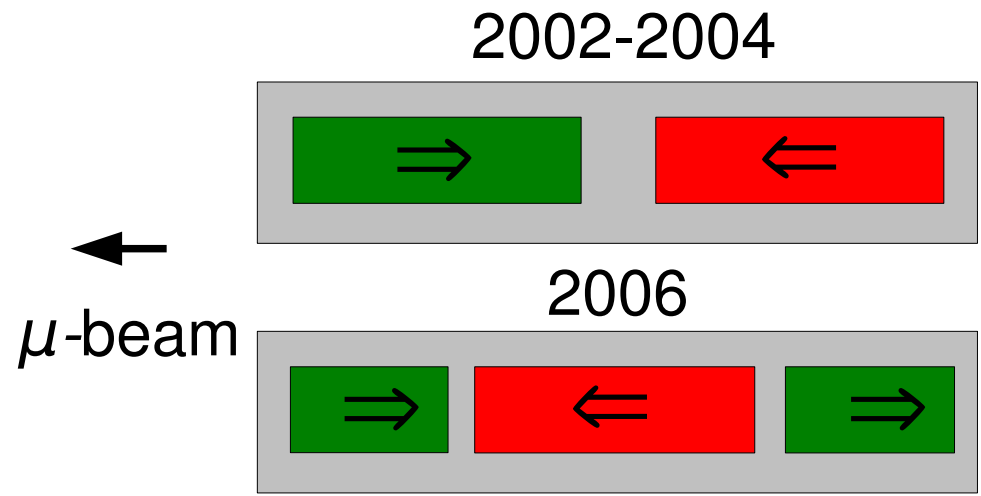
Exploratory and discovery stage.  
Some experiments and data  
might give hints.

COMPASS, HERMES, CLAS,  
STAR, PHOENIX

# Direct measurement of $\Delta G/G$



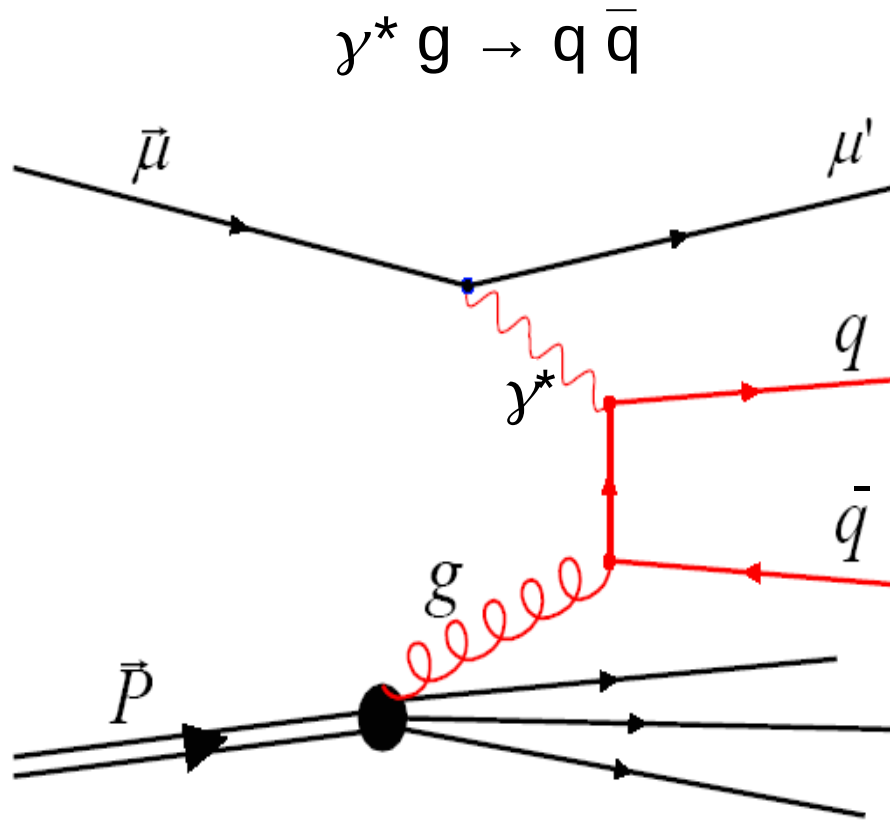
Photon-gluon fusion process (PGF)



$$A_{PGF} = \frac{N_{PGF}^{\rightarrow\leftarrow} - N_{PGF}^{\leftarrow\leftarrow}}{N_{PGF}^{\rightarrow\leftarrow} + N_{PGF}^{\leftarrow\leftarrow}}$$



Experiments with polarised beam and target are sensitive to gluon helicity



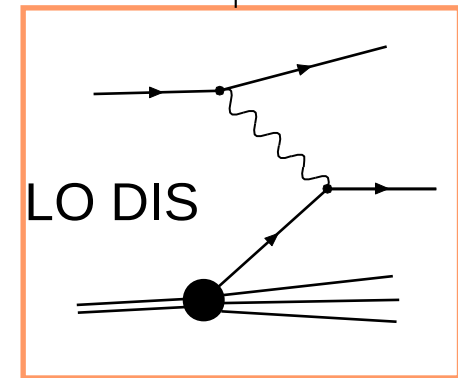
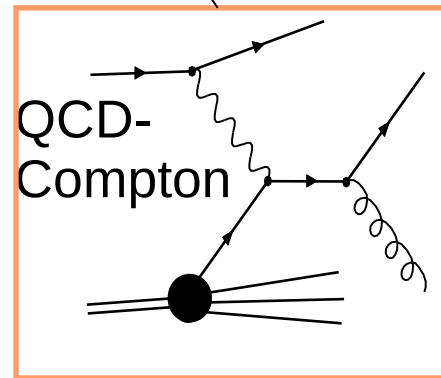
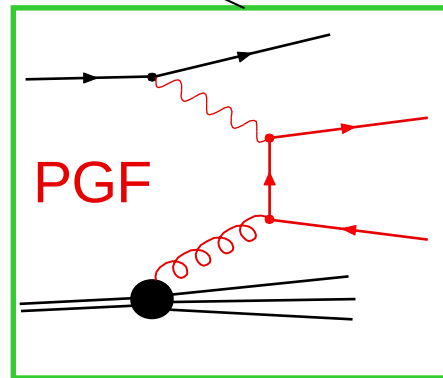
Photon-gluon fusion process (PGF)

To select this process there are two methods :

- **High transverse momentum hadrons** ( $Q^2 < 1$  and  $Q^2 > 1$  (GeV/c) $^2$ )
  - ☺ Much more statistics.
  - ☹ Physical background: strongly model dependent, requires a very good agreement between MC and Data.
- **Open-charm meson** (D mesons)
  - ☺ Provides the purest sample of PGF events, almost free from background contamination. Small dependence on MC.
  - ☹ Low statistics.

$$A_{LL}^{2h}(x) = \frac{\Delta G}{G}(x_g) a_{LL}^{PGF} \frac{\sigma^{PGF}}{\sigma^{Tot}} + A_1^{LO}(x_C) a_{LL}^C \frac{\sigma^C}{\sigma^{Tot}} + A_1^{LO}(x_{Bj}) D \frac{\sigma^{LO}}{\sigma^{Tot}}$$

$A_1^{LO}$  : estimated by an  
inclusive sample



Final formula for  
the gluon  
polarisation

$$\frac{\Delta G}{G}(x_g^{av}) = \frac{1}{\beta} \left[ A_{LL}^{2h}(x_{Bj}) + A_{corr} \right]$$

$$\beta = a_{LL}^{PGF} R_{PGF} - a_{LL}^{PGF, incl} R_{PGF}^{incl} \frac{R_{LO}}{R_{LO}^{incl}} - a_{LL}^{PGF, incl} \frac{R_C R_{PGF}^{incl}}{R_{LO}^{incl}} \frac{a_{LL}^C}{D}$$

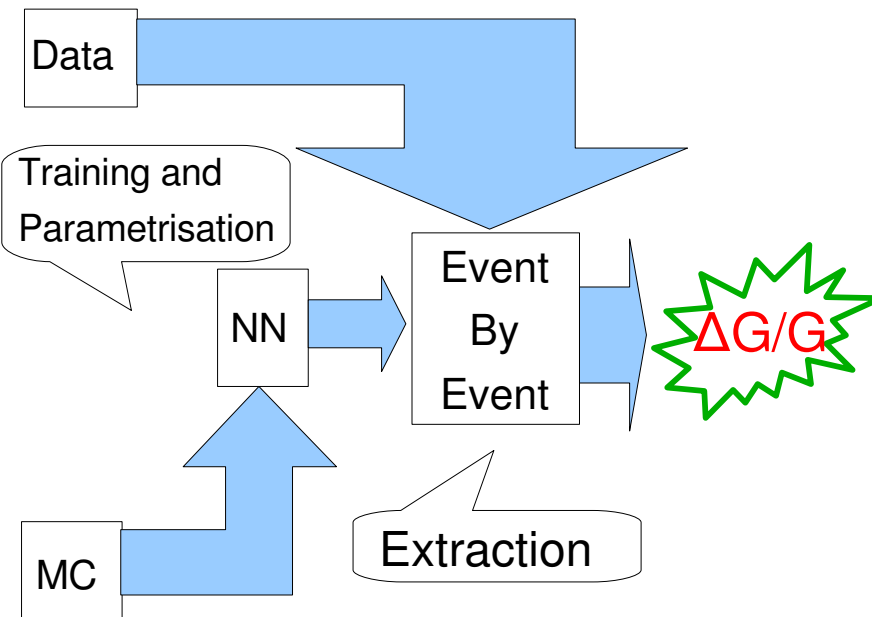
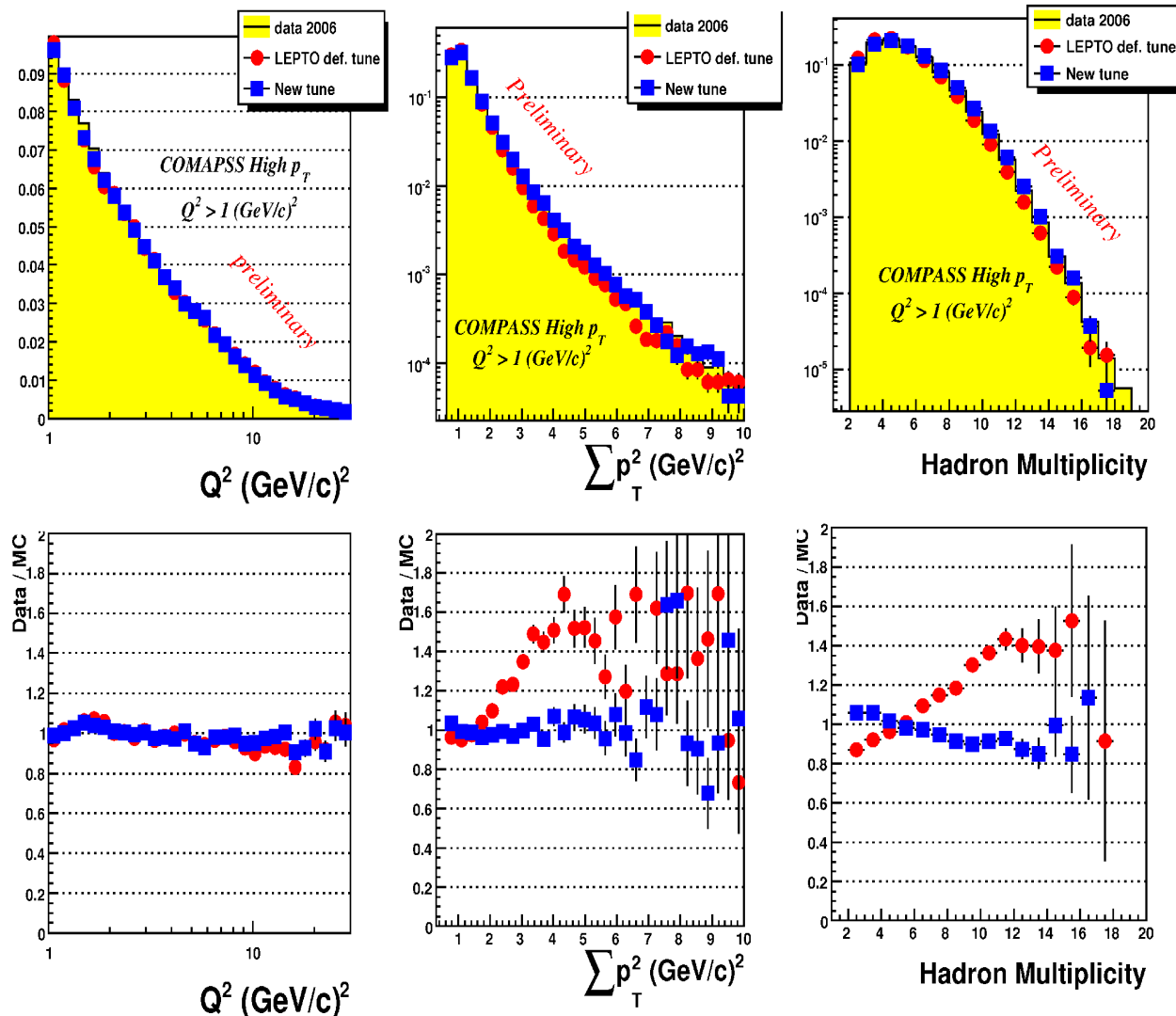
$$A_{corr} = - \left( A_1(x_{Bj}) D \frac{R_{LO}}{R_{LO}^{incl}} - A_1(x_C) \beta_1 + A_1(x_C') \beta_2 \right)$$

- $A_{LL}^{2h}$  : measured from the two hadron sample.
- $a_{LL}^i$  and  $R_i$  : estimated from MC and parametrised using a Neural Network.

- Full chain of MC has been used: Generator (LEPTO) + Apparatus Simulation (GEANT) + Reconstruction Program.
- PDF: MSTW2008LO.
- High  $p_T$  sample:
  - MC with **parton shower ON**.
  - A **new tuning** was performed to **improve** the hadron description.

Data-MC comparison:

$Q^2$ ,  $p_T$  and Hadron Multiplicities.



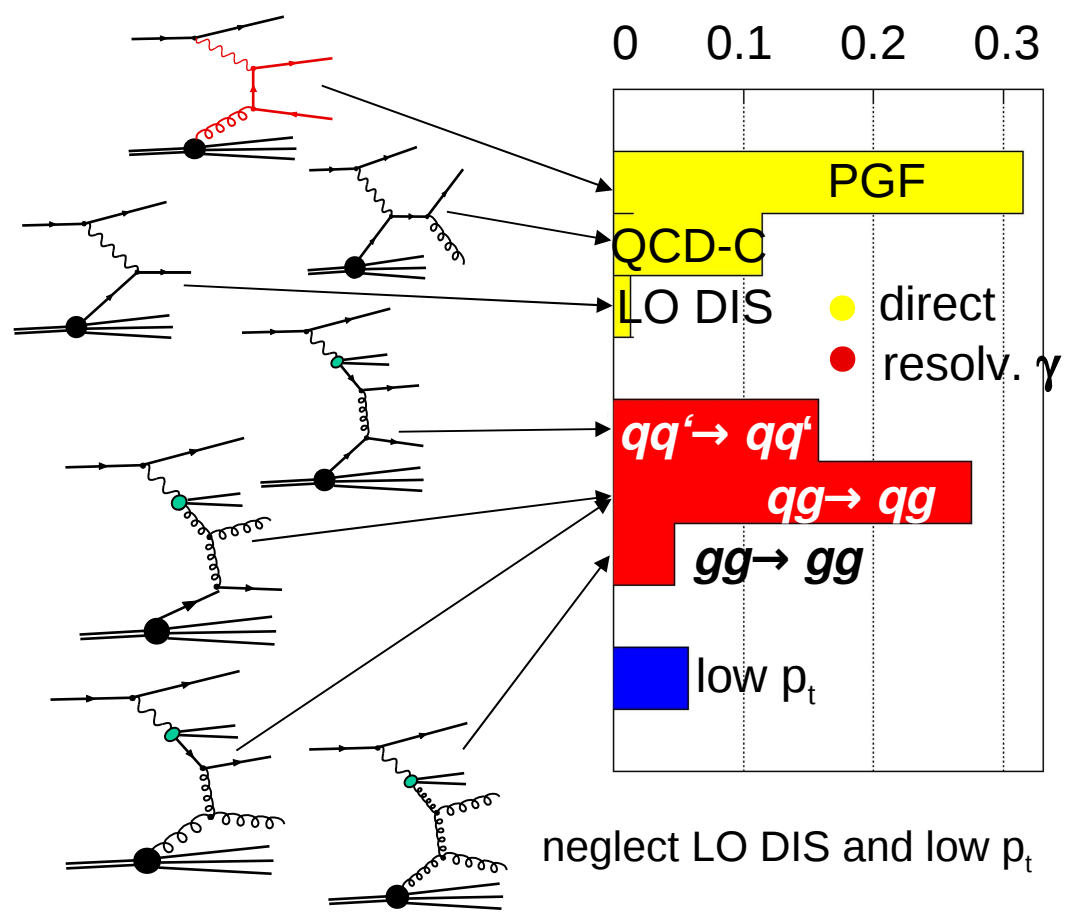
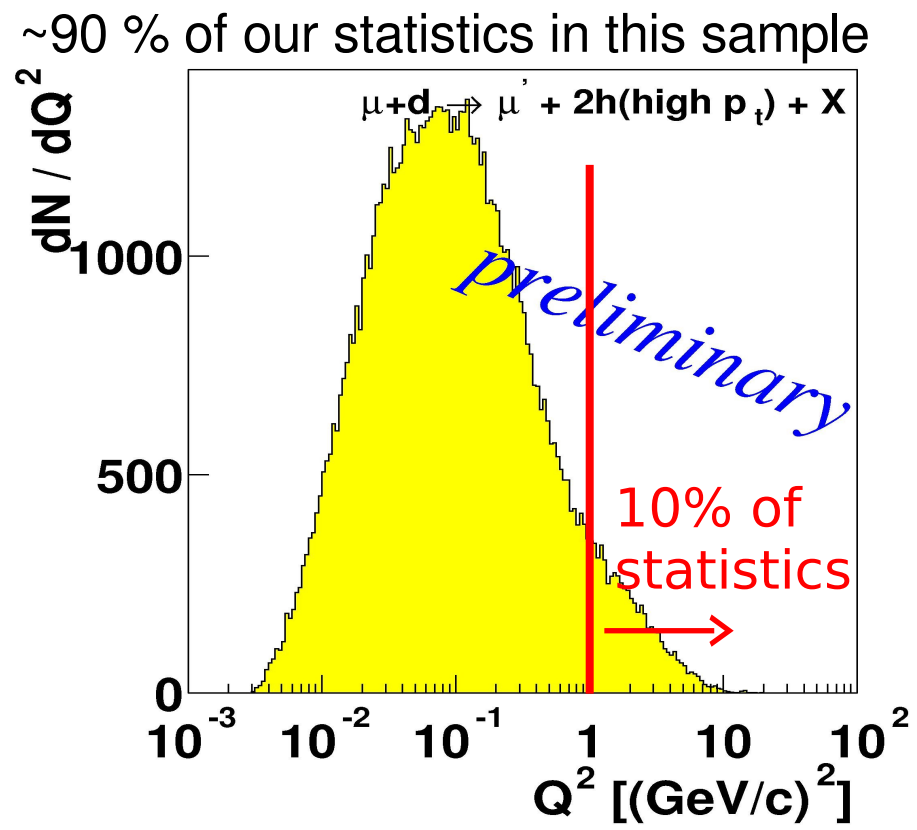


$$\frac{\Delta G}{G} = 0.125 \pm 0.060 \pm 0.063 \quad x_g = 0.09^{+0.08}_{-0.04} \quad \langle \mu^2 \rangle_g = 3.4 \text{ (GeV/c)}^2$$

- The whole statistics was divided, for the first time, in 3 independent samples, having each one its own  $x_g$  distribution.

	1 <sup>st</sup> point	2 <sup>nd</sup> point	3 <sup>rd</sup> point
$\Delta G/G$	$0.147 \pm 0.091 \pm 0.088$	$0.079 \pm 0.096 \pm 0.082$	$0.185 \pm 0.165 \pm 0.143$
$\langle x_g \rangle$	$0.07^{+0.05}_{-0.03}$	$0.10^{+0.07}_{-0.04}$	$0.17^{+0.10}_{-0.06}$

⇒ Within the errors the 3 points show no  $x_g$  dependence



2002-2004 Preliminary:

$$\Delta G/G = 0.016 \pm 0.058(\text{stat}) \pm 0.055(\text{syst})$$

2002-2003 Published:

$$\Delta G/G = 0.024 \pm 0.089(\text{stat}) \pm 0.057(\text{syst})$$

*Phys. Lett. B 633,25*



- The relation between the number of reconstructed  $D^0$  (for each target cell configuration) and  $\Delta G/G$  is given by:

$$N_t = a \phi n (S + B) \left( 1 + f P_T P_\mu \left[ a_{LL} \frac{S}{S+B} \frac{\Delta G}{G} + D \frac{B}{S+B} A^{bg} \right] \right), \quad t = (u, d, u', d')$$

acceptance, muon flux, number of target nucleons

Open Charm event probability

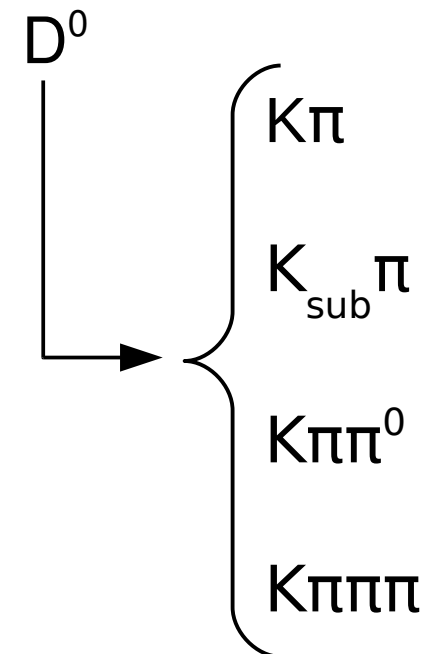
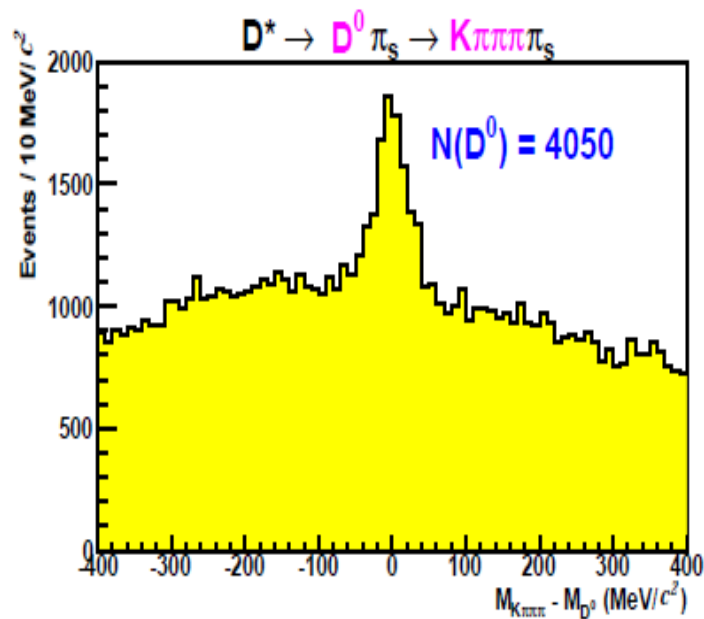
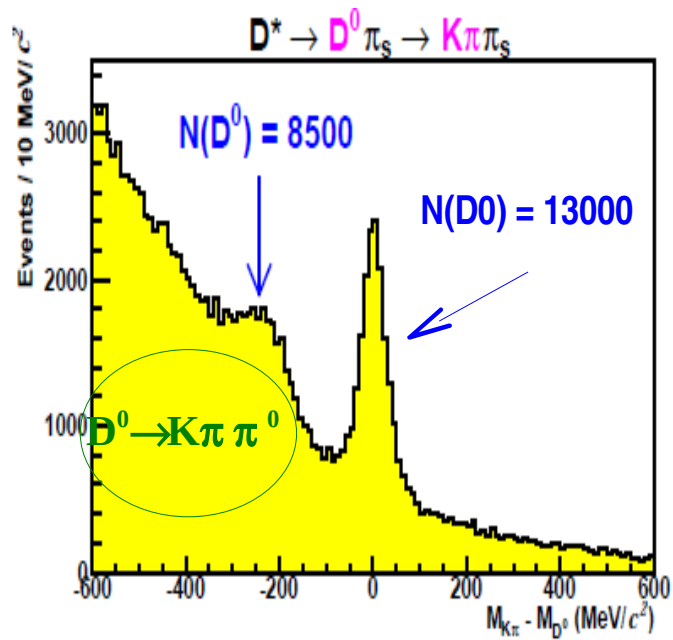
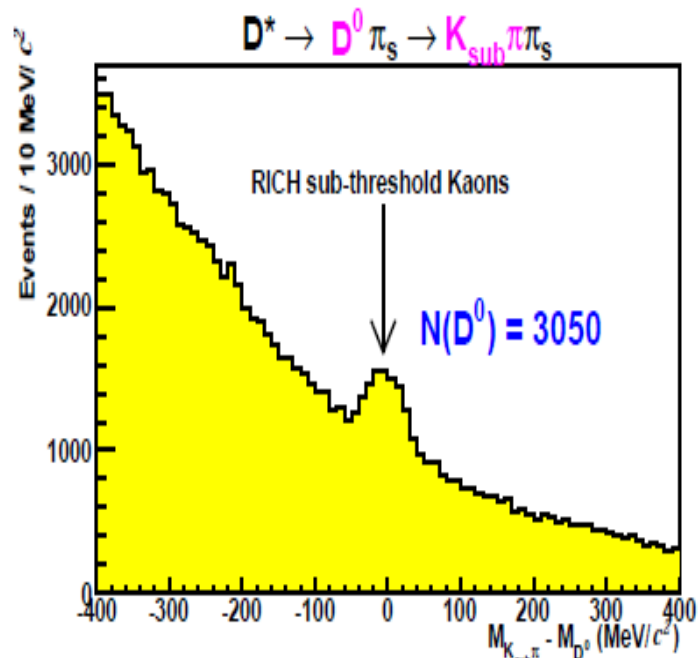
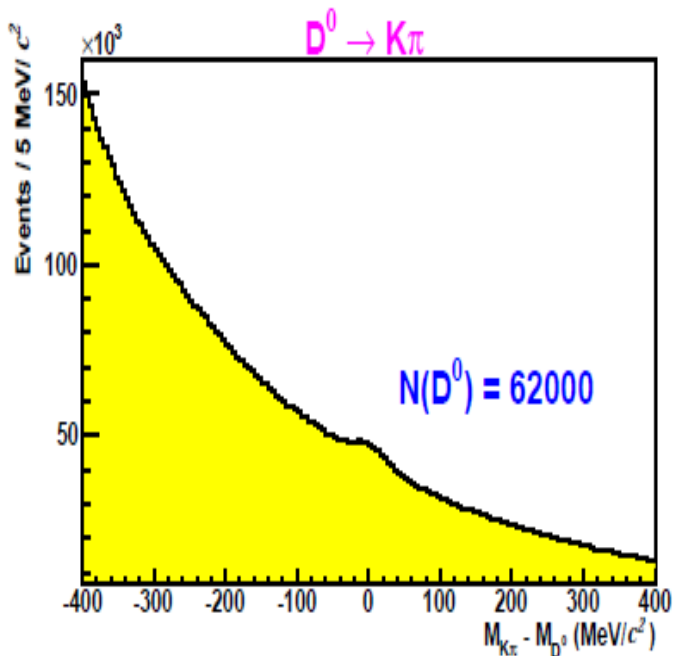
- Each equation is weighted with a signal weight  $w_S = f P_m a_{LL} S/(S+B)$  and also with a background weight  $w_B = f P_m D B/(S+B)$ :

**8 equations with 7 unknowns:**  $\Delta G/G$ ,  $A^{bg}$  + 5 independent  $\alpha = (a\phi n)$  factors

The system is solved by a  $\chi^2$  minimisation



# D<sup>0</sup> invariant mass spectra: 2002-2007 data



## Number of $D^0$ :

- Total = 90600
- ${}^6\text{LiD}$  = 65600
- $\text{NH}_3$  = 25000

- Two real data samples (with the same cuts applied) are compared by a Neural Network (using some kinematic variables as a learning vector):

- **Signal model**

$$gcc = K^+ \pi^- \pi_s^- + K^- \pi^+ \pi_s^+$$

( $D^0$  spectrum: signal + background)

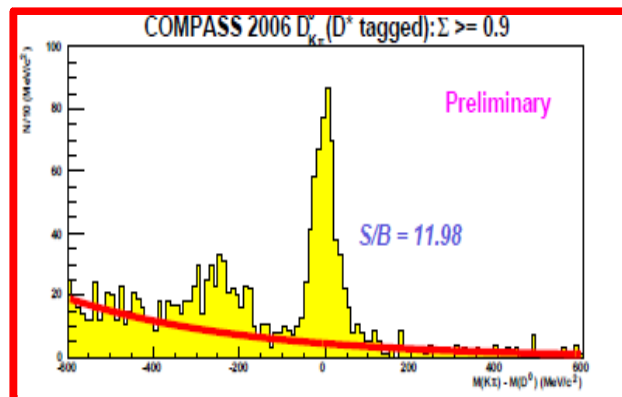
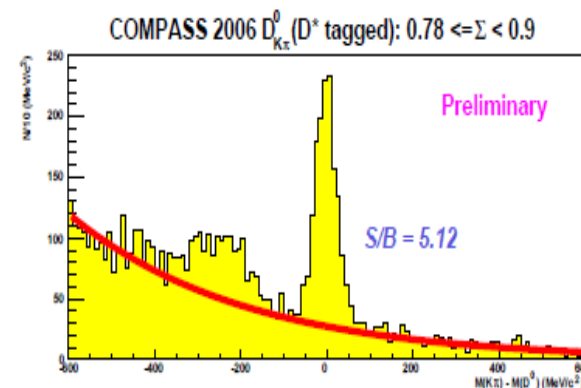
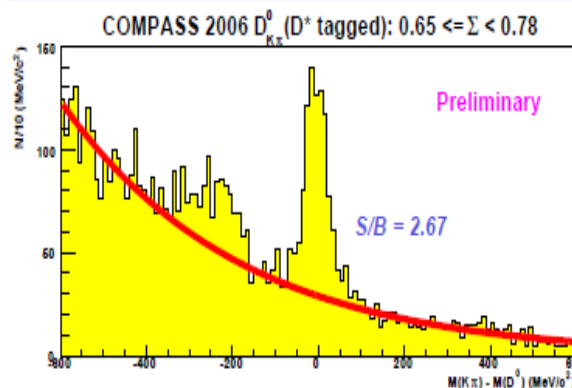
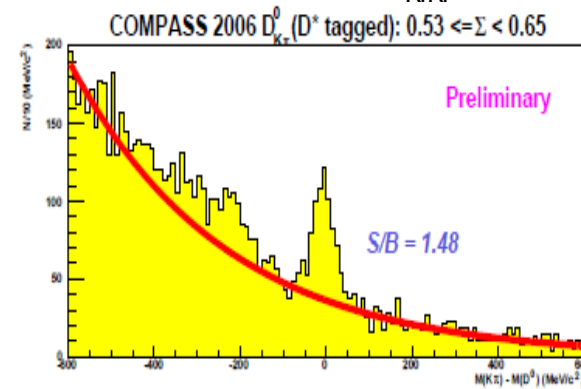
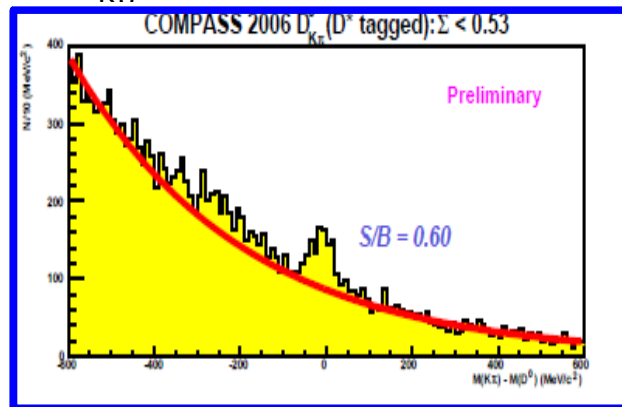
- **Background model**

$$wcc = K^+ \pi^+ \pi_s^- + K^- \pi^- \pi_s^+$$

(no  $D^0$  is allowed)

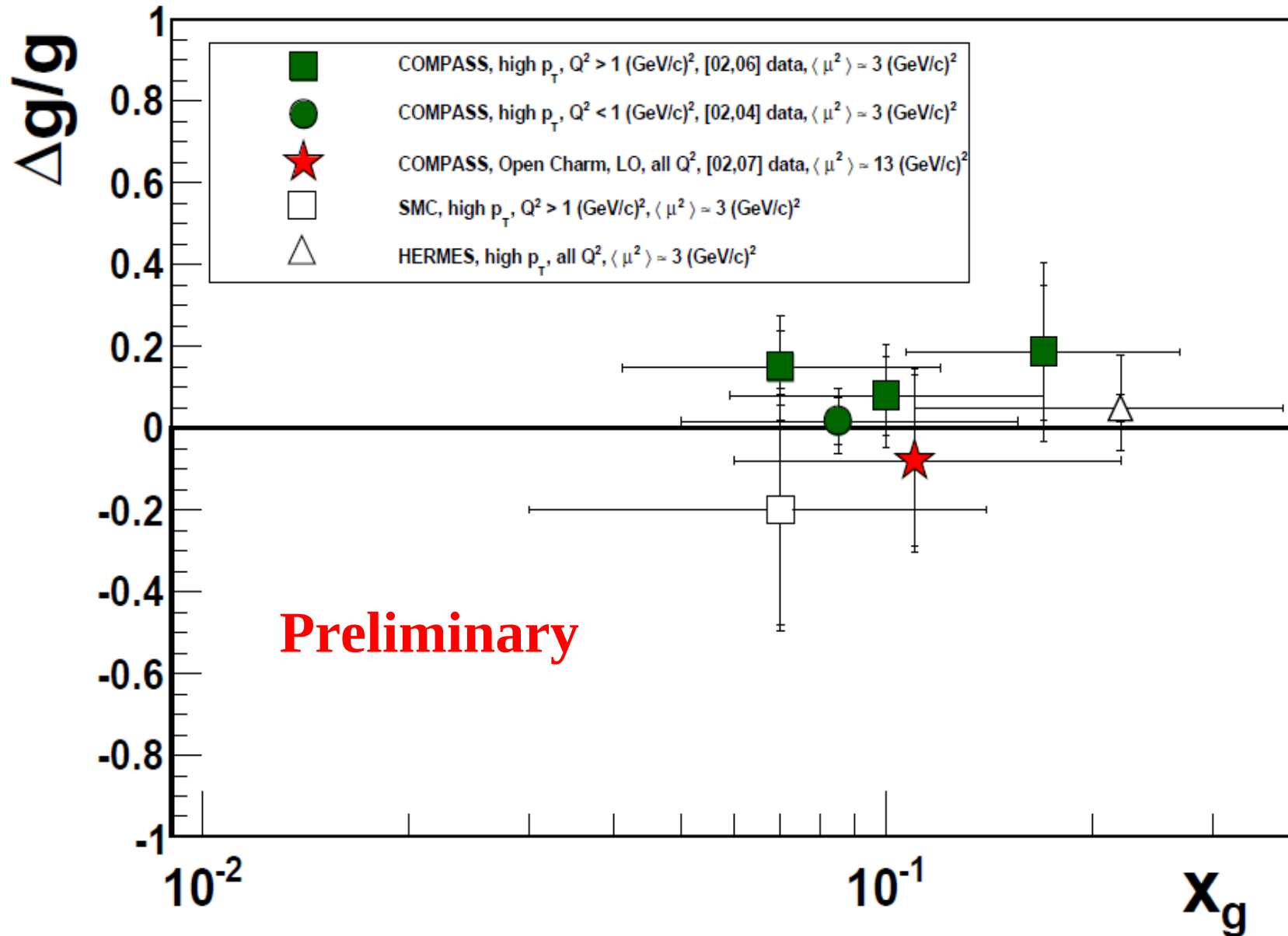
- If the background model is good enough: The Neural Network is able to distinguish the signal from the combinatorial background on a event by event basis (inside gcc)

$D_{K\pi}^0$  tagged spectrum in bins of  $\Sigma = S/(S+B)_{NN}$

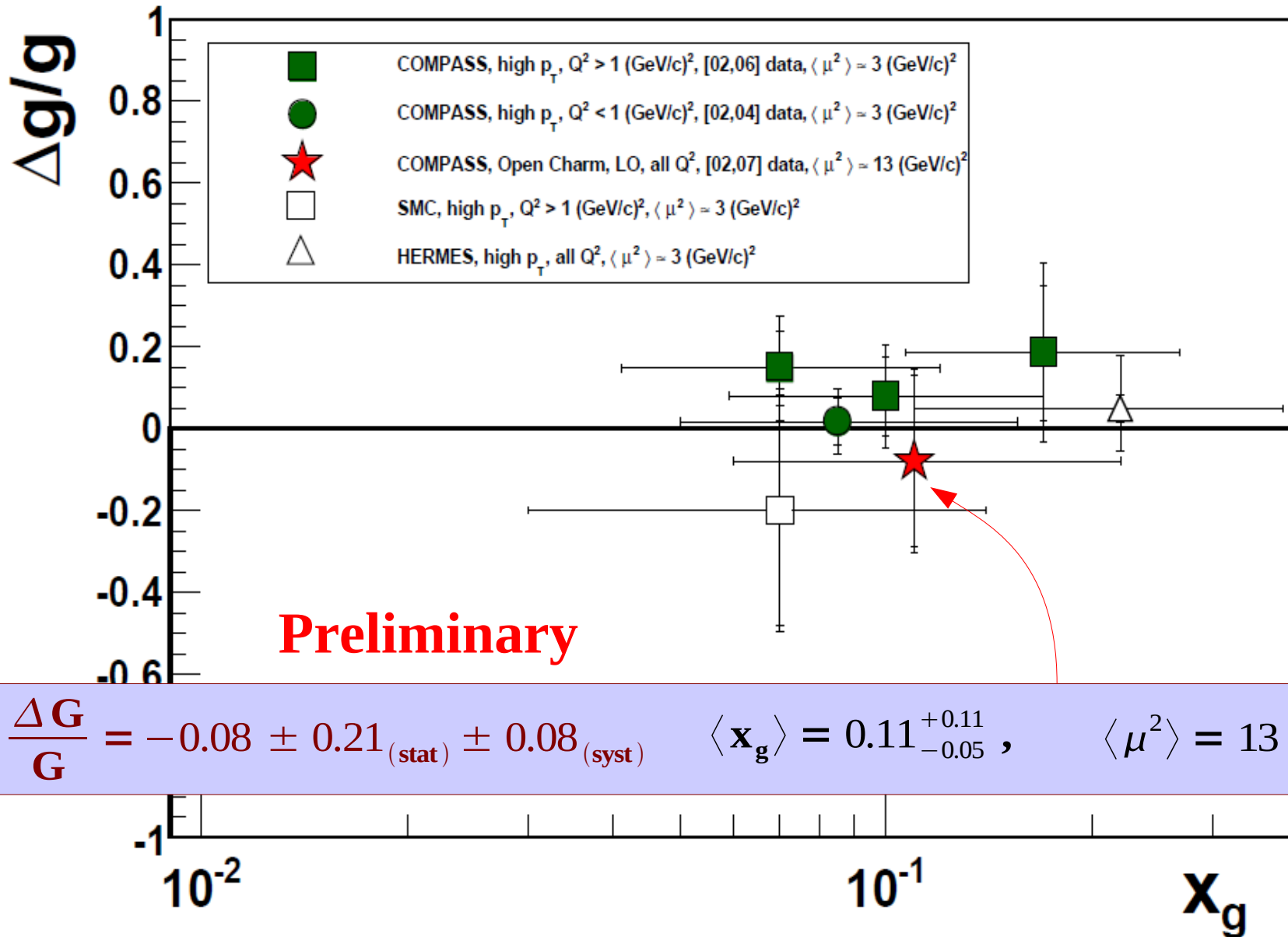


$$\delta \left( \frac{\Delta G}{G} \right) \propto \frac{1}{FOM}$$

# $\Delta G/G$ Results (LO)



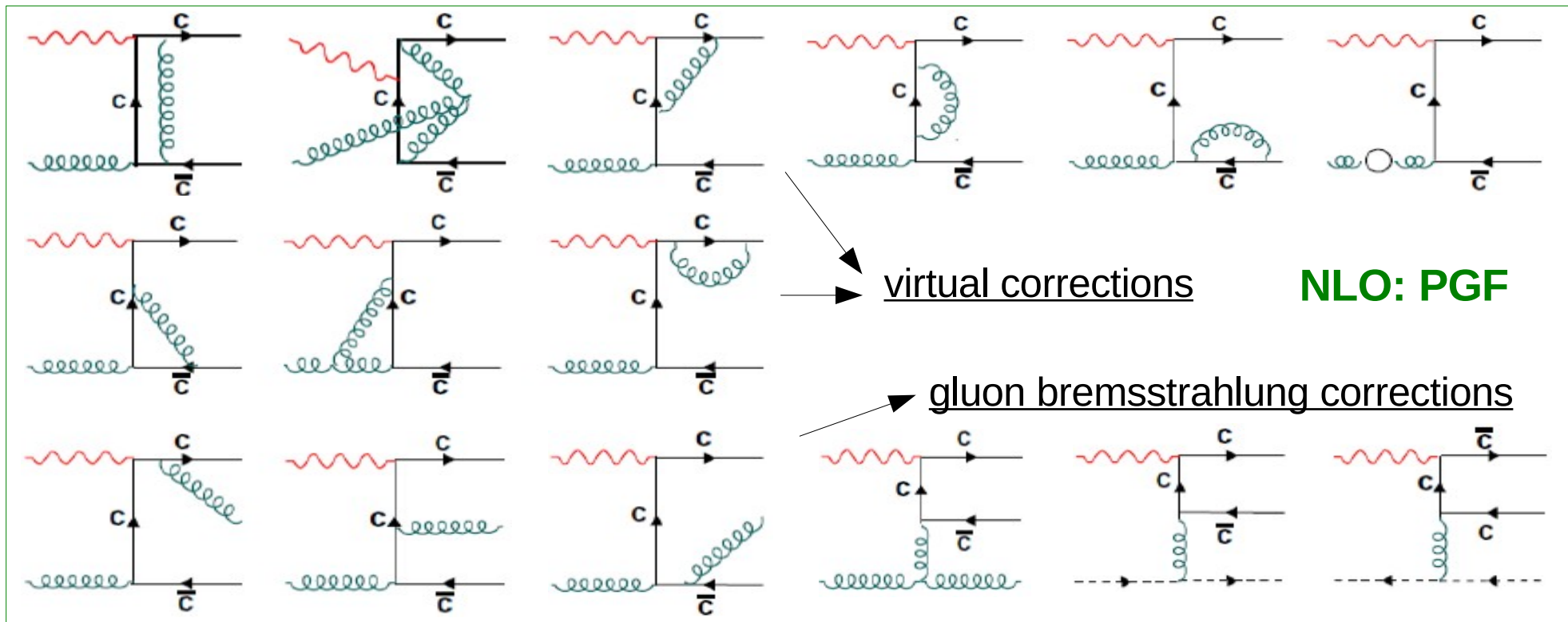
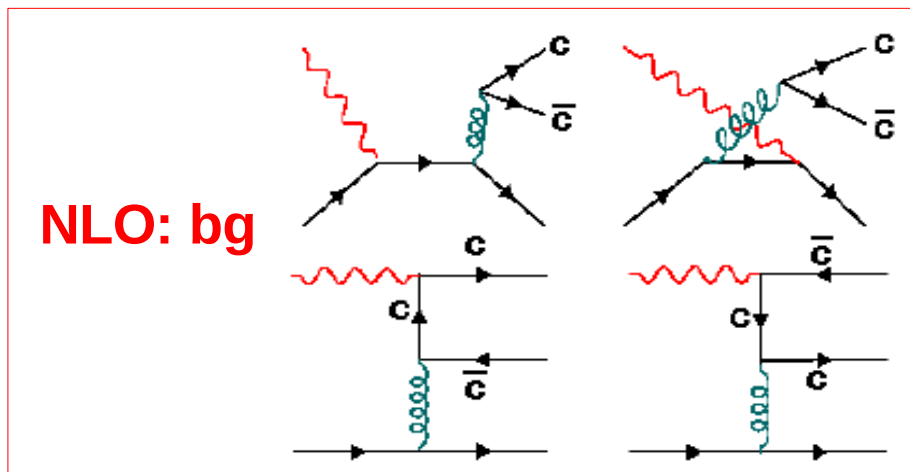
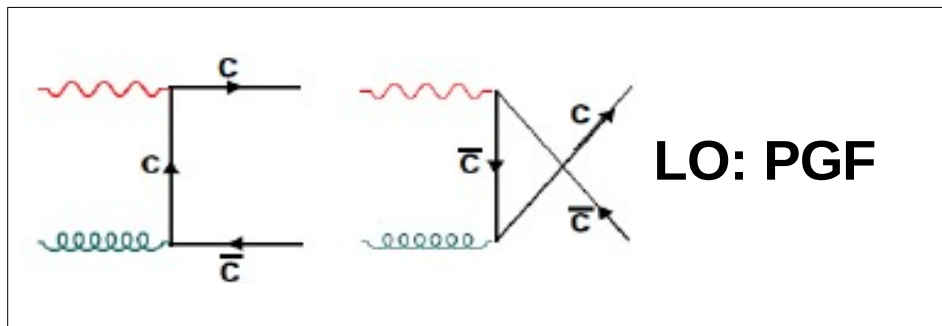
# $\Delta G/G$ Results (LO)



$$\frac{\Delta G}{G} = -0.08 \pm 0.21_{(stat)} \pm 0.08_{(syst)} \quad \langle x_g \rangle = 0.11^{+0.11}_{-0.05}, \quad \langle \mu^2 \rangle = 13 \text{ (GeV/c)}^2$$

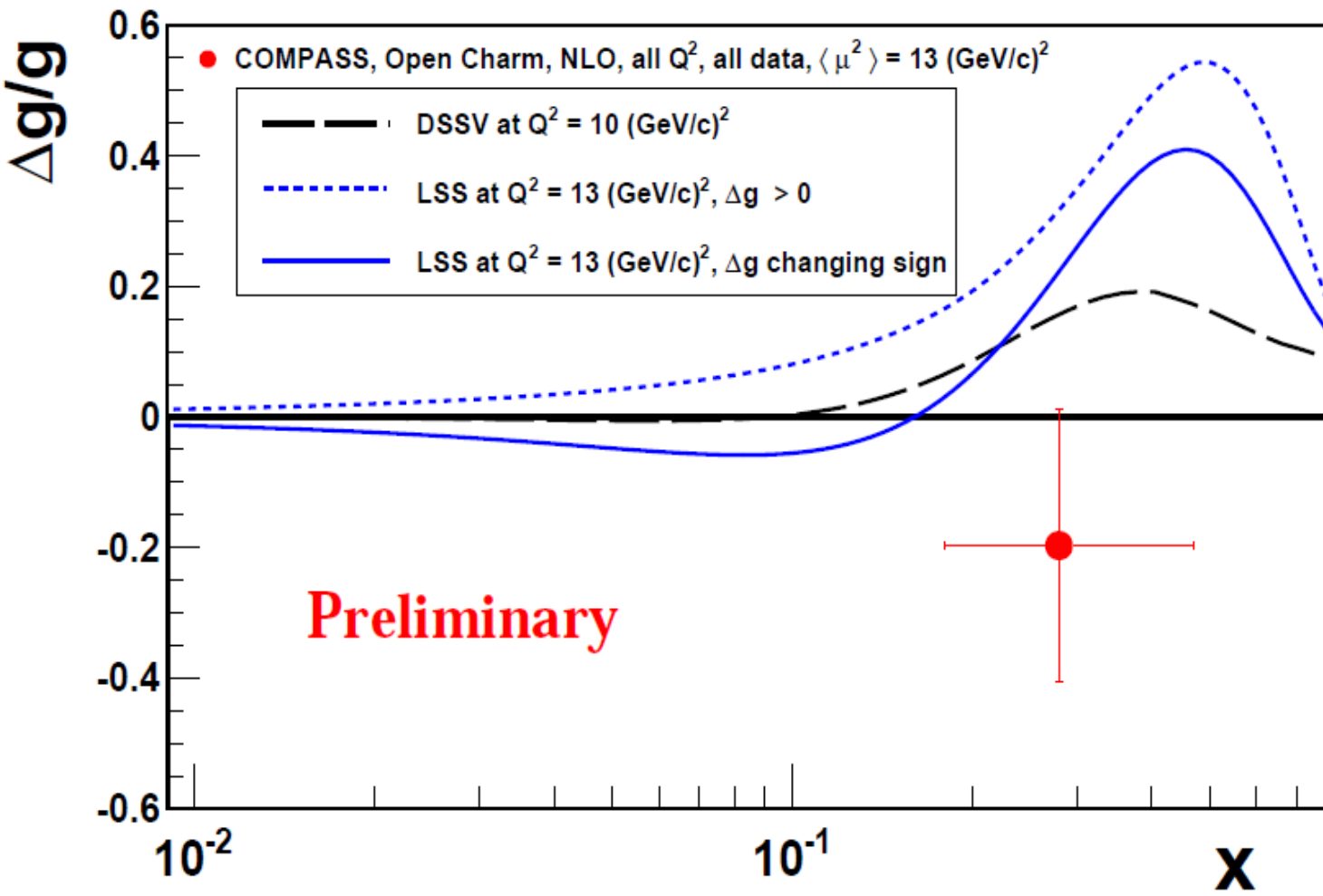
NLO corrections to the analysing power

$a_{LL}$





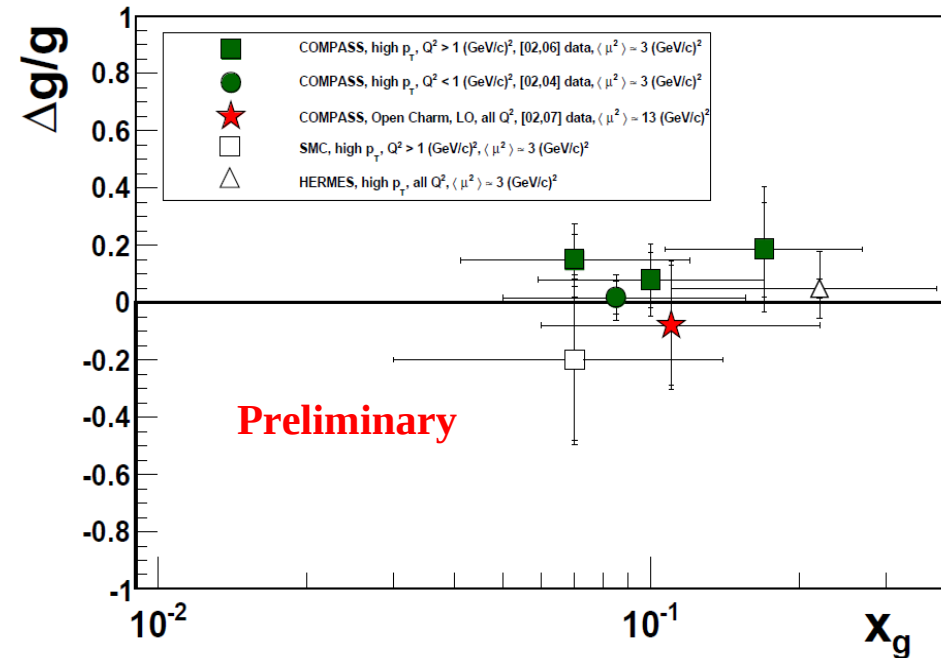
# $\Delta G/G$ new Result (NLO)



$$\frac{\Delta G}{G} = -0.20 \pm 0.21 \pm 0.08 \text{ (syst)} \quad @ \langle x_g \rangle = 0.28^{+0.19}_{-0.10}, \langle \mu^2 \rangle = 13 \text{ (GeV/c)}^2$$

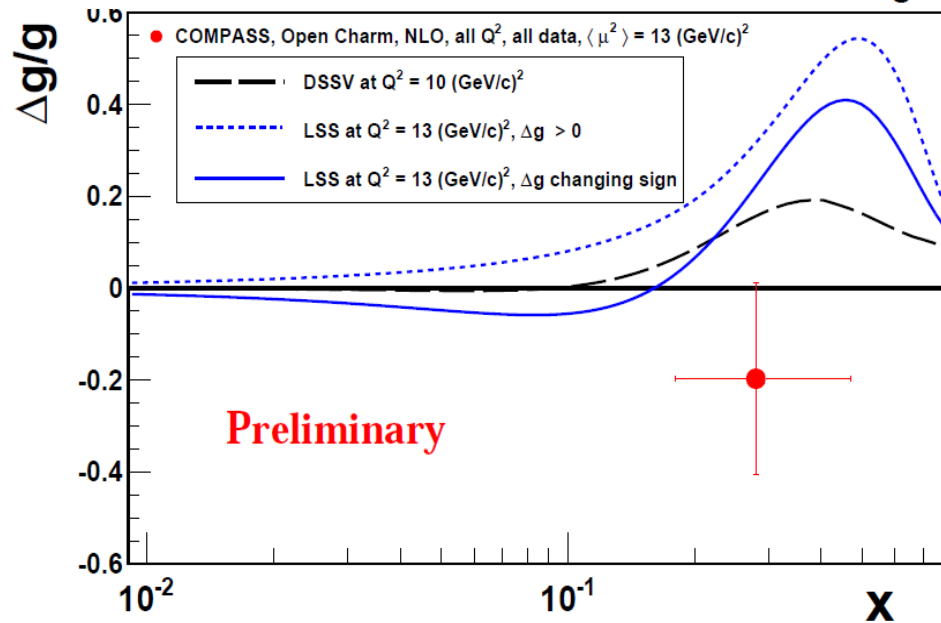
**Preliminary:** theoretical uncertainties still under study ( $a_{LL}$ )





## Summary:

- The importance of the gluon polarisation measurement concerning the nucleon spin structure was emphasised.
- The direct measurement methods were explained.
- The gluon polarisations results are presented.



## Conclusion:

- All measurements of  $\Delta G/G$  are compatible with zero, around  $X_g \sim 0.1$
- The  $\Delta G$  seems to be small contribution.
- The missing contribution could be in  $L_{partons}$ .
- COMPASS-II program foresees to measure  $L_{partons}$  via GPDs.

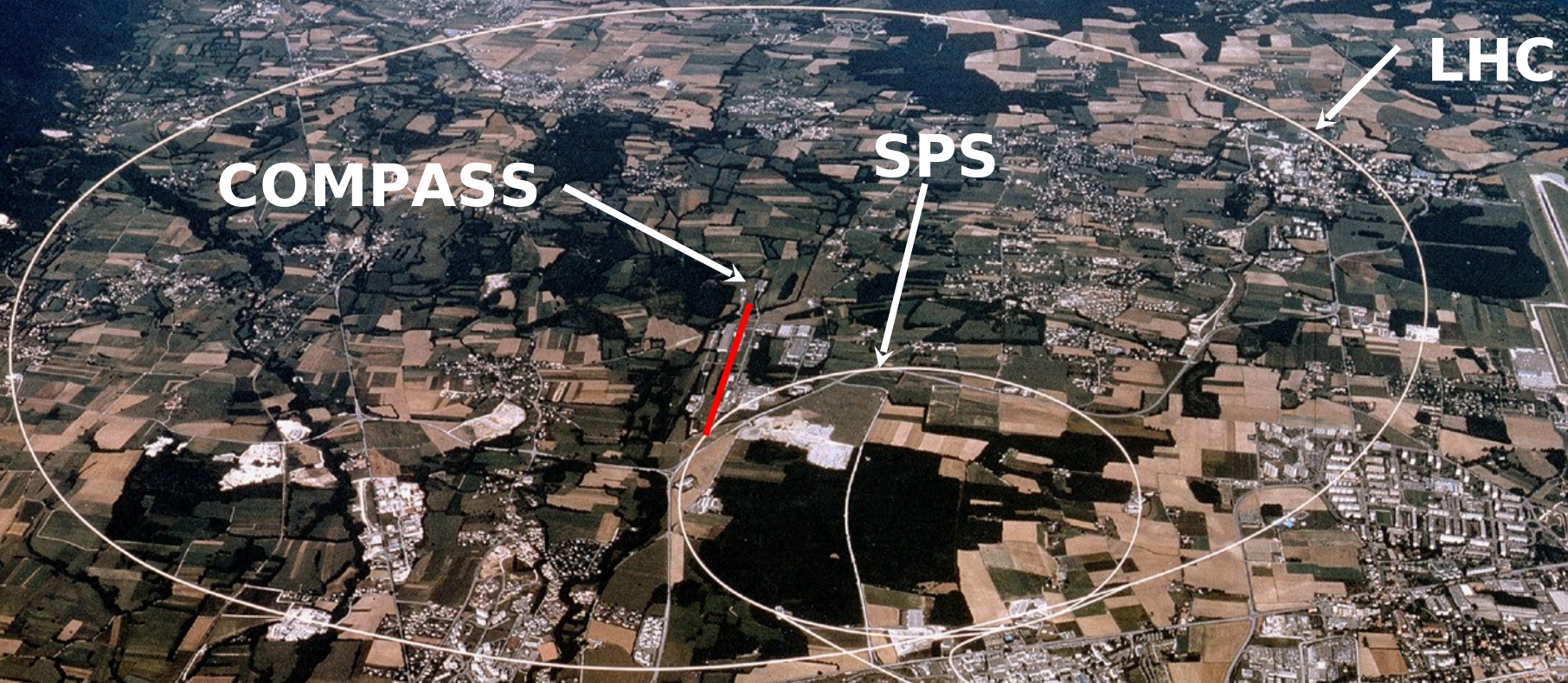
# Spares



# THE COMPASS EXPERIMENT

Beam:  $2 \cdot 10^8 \mu^+$ / spill  
Luminosity:  $\sim 5 \cdot 10^{32} \text{ cm}^{-2} \text{ s}^{-1}$   
Beam polarisation: 80%  
Beam momentum: 160 GeV/c








Data taken: 2002 - 2010, ...



~250 physicists  
25 institutes  
11 countries



## Common Muon and Proton Apparatus for Structure and Spectroscopy

-  Trackers
-  Magnets
-  RICH
-  Electromagnetic Calorimeters
-  Hadronic Calorimeters
-  Absorbers
-  Target

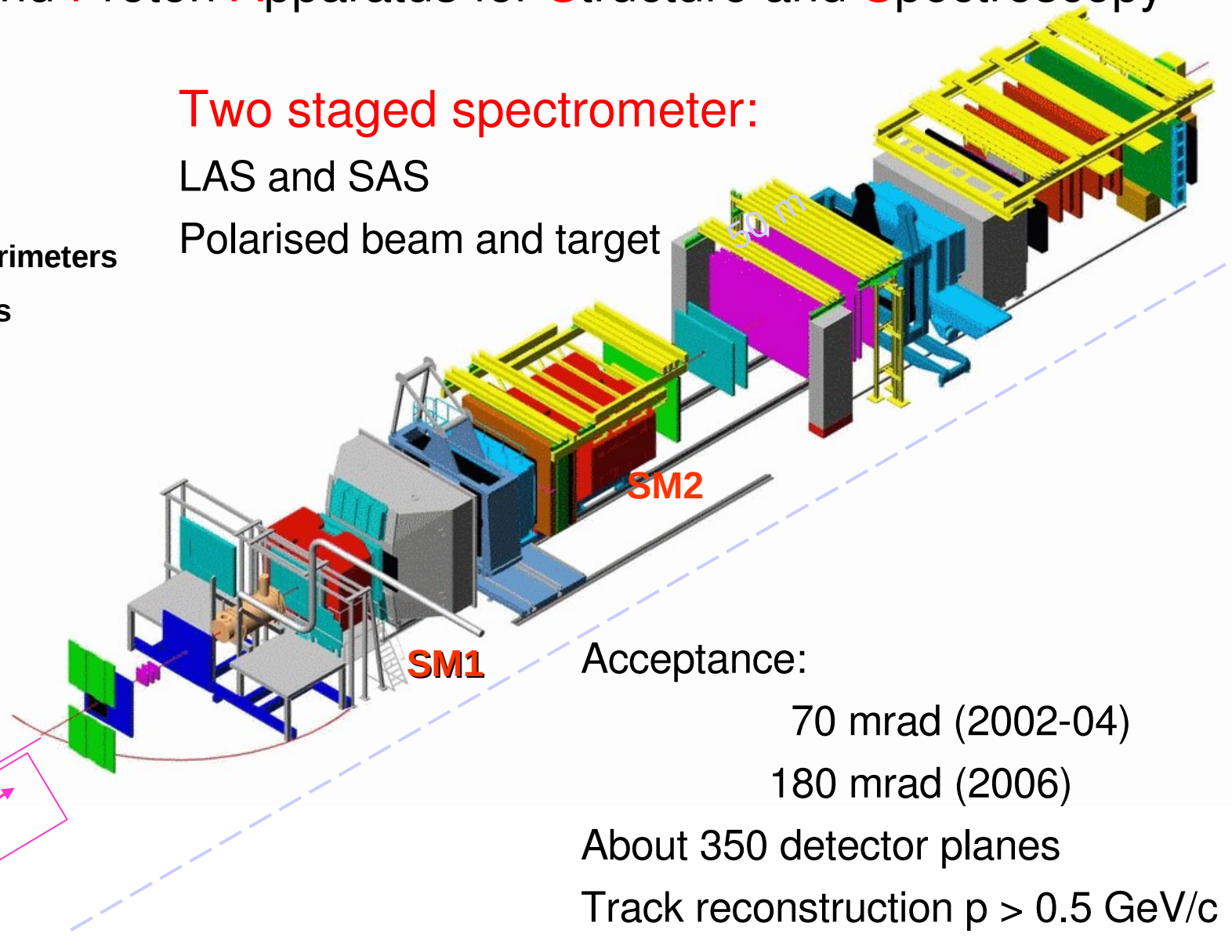
Two staged spectrometer:

LAS and SAS

Polarised beam and target

NIM A577 (2007) 455

160 GeV  $\mu^+$



Acceptance:

70 mrad (2002-04)

180 mrad (2006)

About 350 detector planes

Track reconstruction  $p > 0.5 \text{ GeV}/c$

This analysis uses **information** from the **MC**, thus a **strong effort** and **care** to ensure that the MC simulation describes as **good** as possible the **data** was undertaken.

Two **MC samples** were used in the analysis: **high  $p_T$**  and **inclusive** samples.

- Full chain of MC has been used:

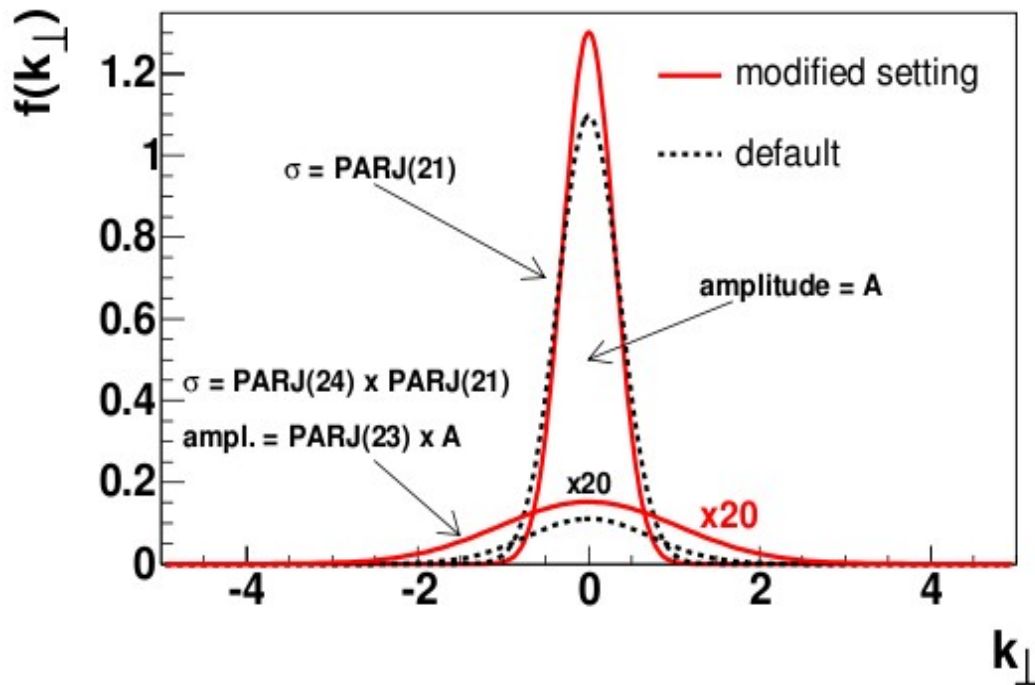
Generator (LEPTO) + Apparatus Simulation (GEANT) + Reconstruction Program.

- PDF: MSTW2008LO.
- High  $p_T$  sample:
  - MC with **parton shower ON** has been used in the analysis.
  - A **new tuning** was performed to **improve** the hadron description.

- The purpose of the **MC tuning** is to **correct** the shapes of the **hadron variables** (momenta) and **fragmentation** (multiplicity).
- In **LEPTO** this can be **achieved** by changing **JETSET** parameters:

PARJ(21)	PARJ(23)	PARJ(24)	PARJ(41)	PARJ(42)
Transverse momentum of the hadron fragmentation			Fragmentation function	

- These **parameters** can be **divided** into **two sets** regarding the **component** of the **trajectory** of the particles: **Transverse** and **longitudinal** variable components.
  - The **sets** can be **tuned independently**.
- ⇒ The tuning improves substantially the Data-MC agreement.



$a = \text{PARJ}(41)$

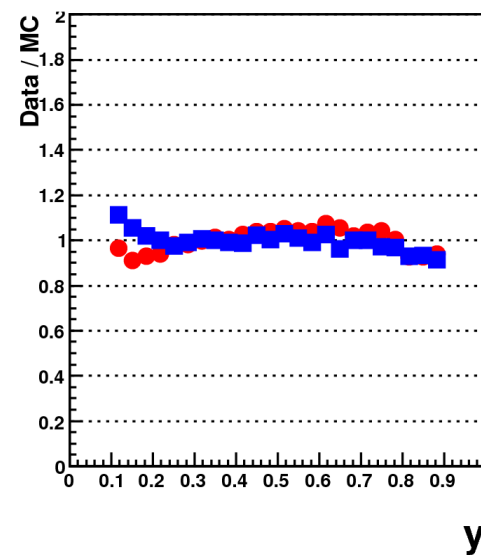
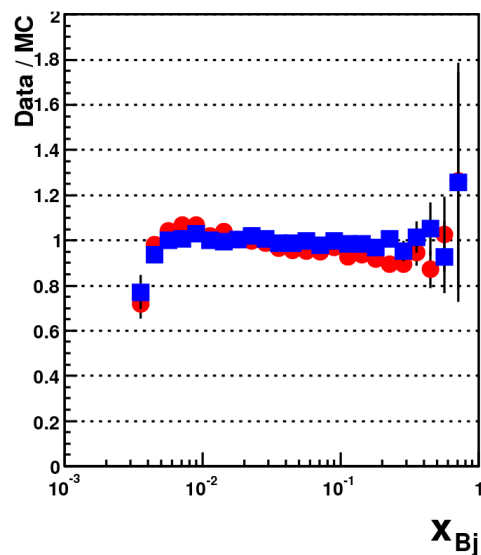
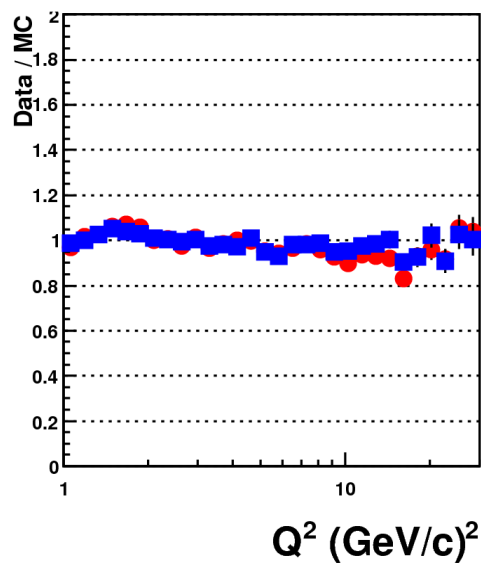
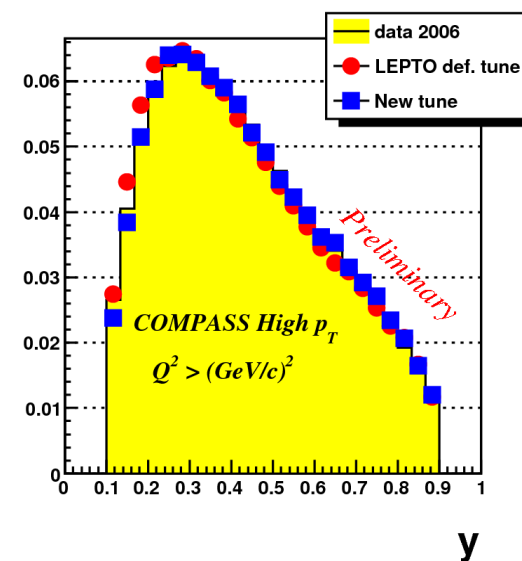
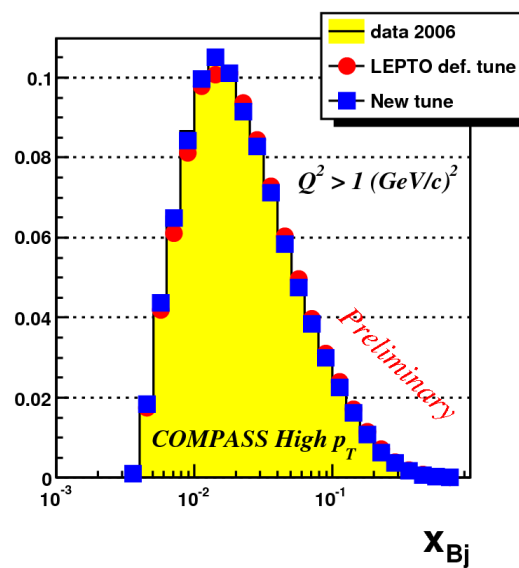
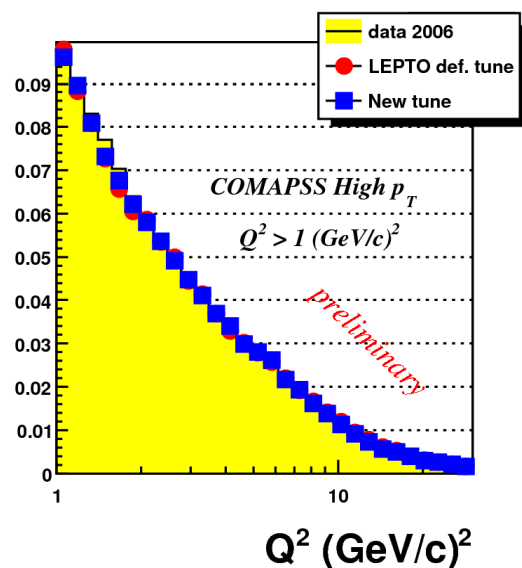
$b = \text{PARJ}(42)$

$$f(z) \propto \frac{1}{2} (1-z)^a \exp\left(-\frac{b m_T^2}{z}\right)$$

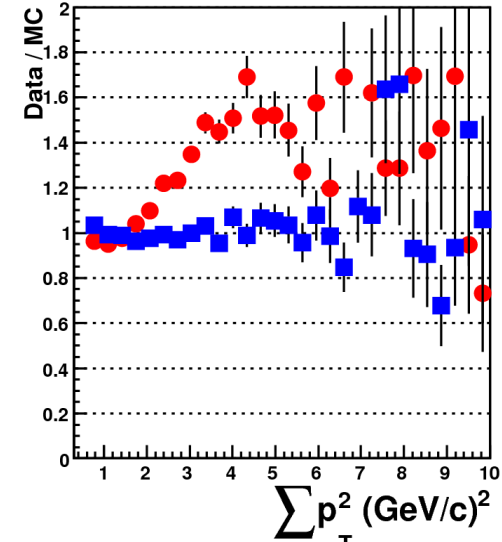
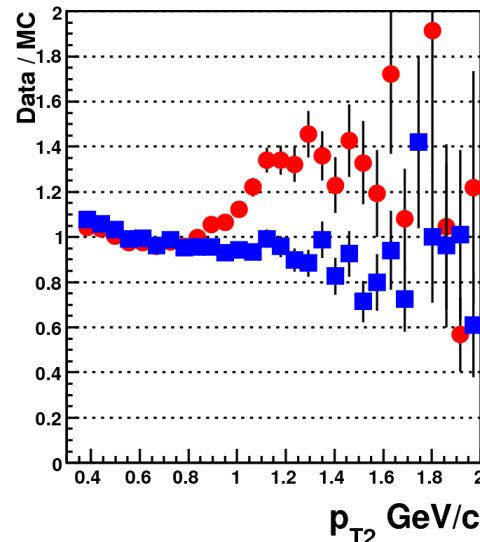
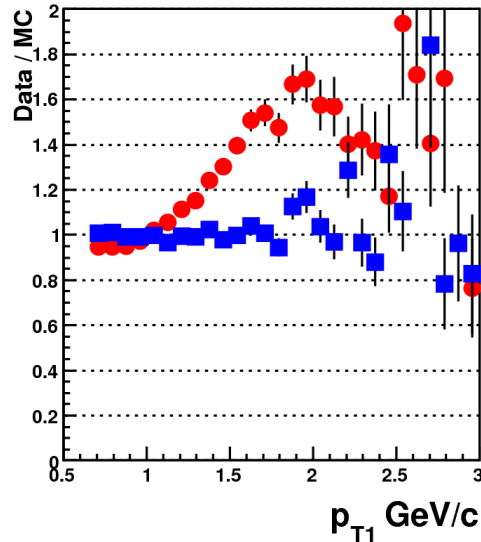
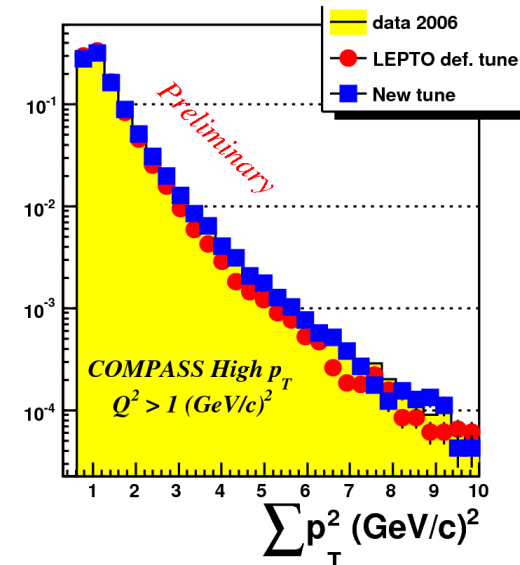
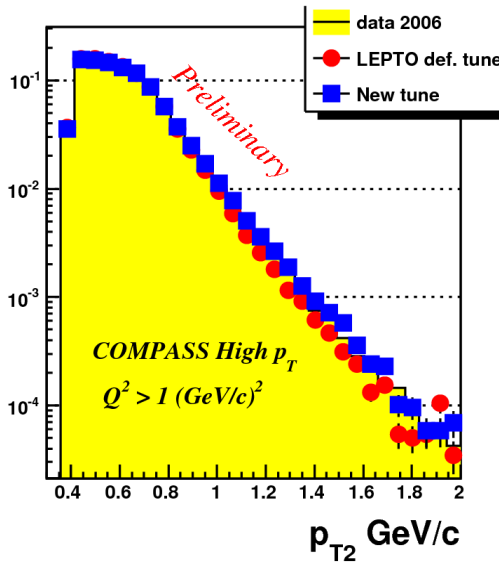
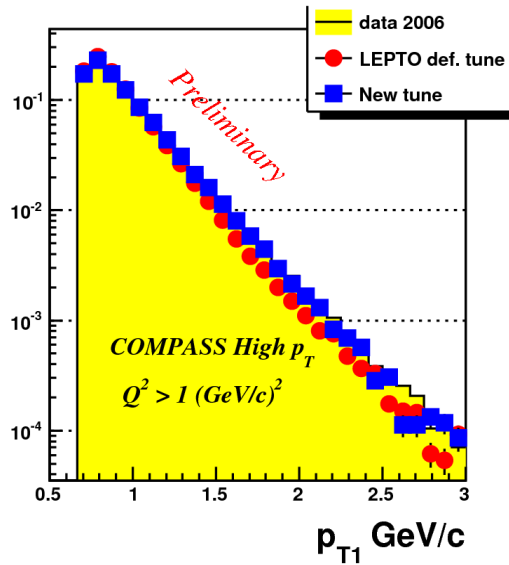
COMPASS new tuning  
LEPTO default tuning

PARJ(21)	PARJ(23)	PARJ(24)	PARJ(41)	PARJ(42)
0.34	0.04	2.8	0.025	0.075
0.36	0.01	2.0	0.3	0.58
Transverse momentum of the hadron fragmentation			Fragmentation function	

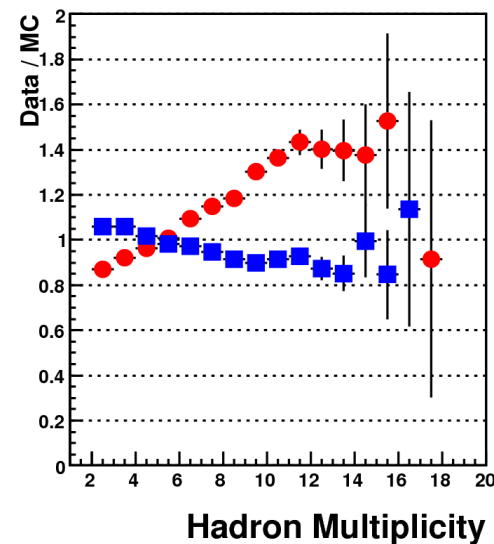
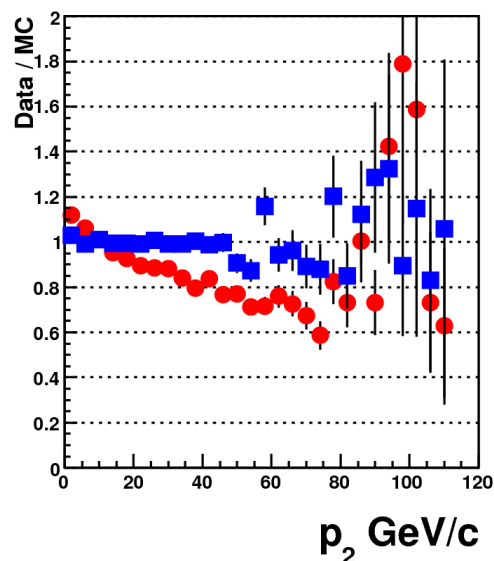
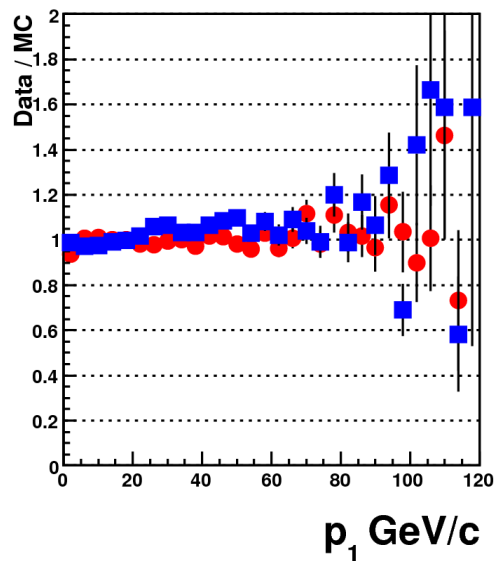
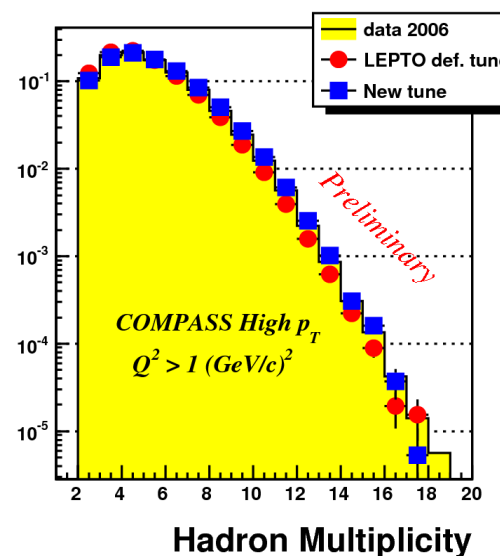
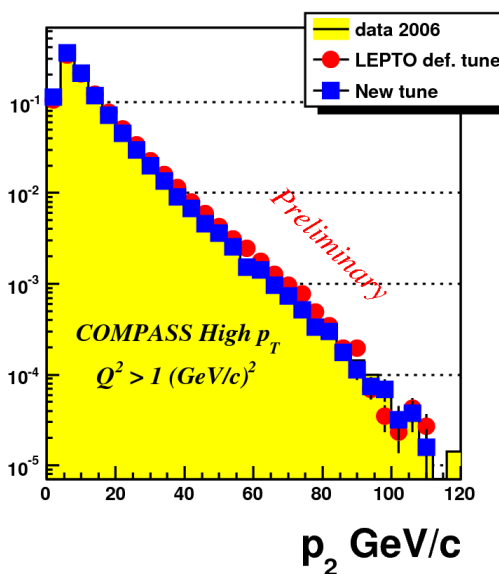
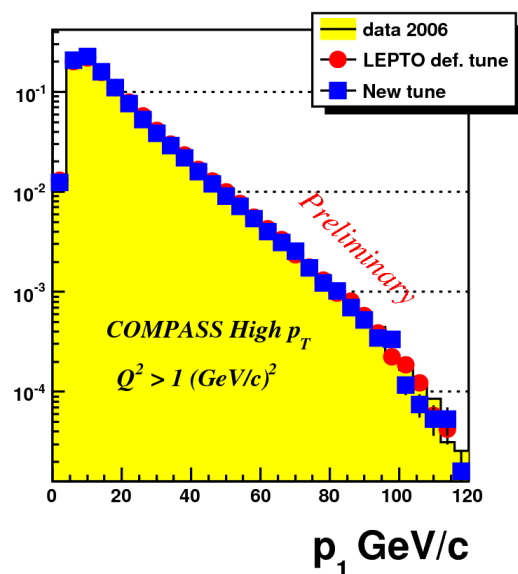




high- $p_T$  sample: hadron variables ( $p_{T1}$ ,  $p_{T2}$  and  $\Sigma p_T^2$ )

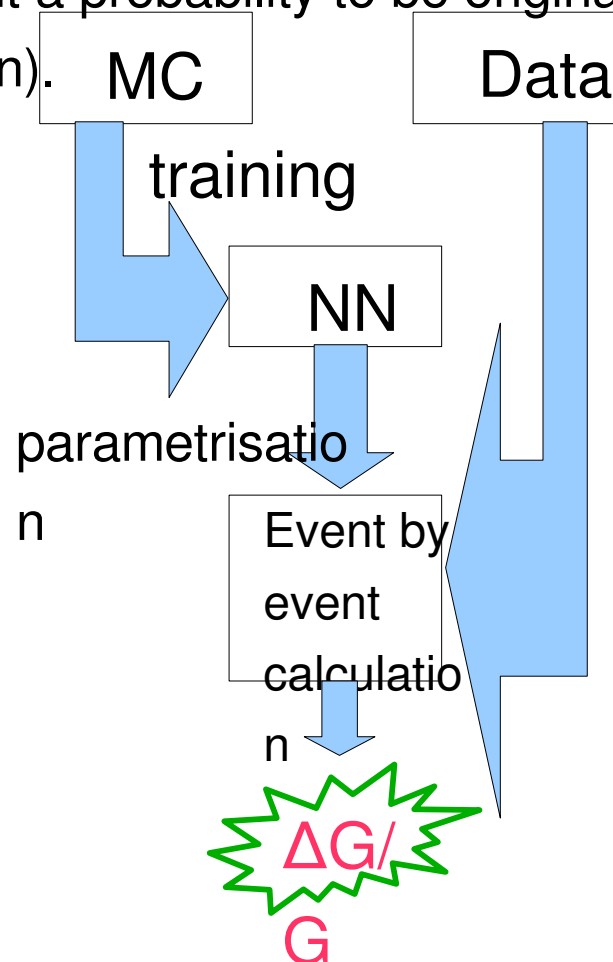


high- $p_T$  sample: hadron variables ( $p_1$ ,  $p_2$  and multiplicity)



A Neural Network is used to assign to each event a probability to be originated from one of the three processes (LO, PGF or Compton).

- A **MC** sample is used to train the Neural Network (NN).
- A parametrisation is constructed for all variables involved in the weight.
- A **Data** sample is weighted on an event-by-event basis.



Optimal usage of the data sample statistics

- A weight is applied on event-by-event basis:

$$W = fDP_b\beta \quad , \text{ where } \beta \text{ is a factor depending on } a_{LL}^i \text{ and } R^i$$

- Therefore for every event we have to know:

$$R_{PGF}, R_C, R_{LO}, R_{PGF}^{incl}, R_C^{incl}, R_{LO}^{incl},$$

$$a_{LL}^{PGF}, a_{LL}^C, a_{LL}^{PGF,incl}, a_{LL}^{C,incl},$$

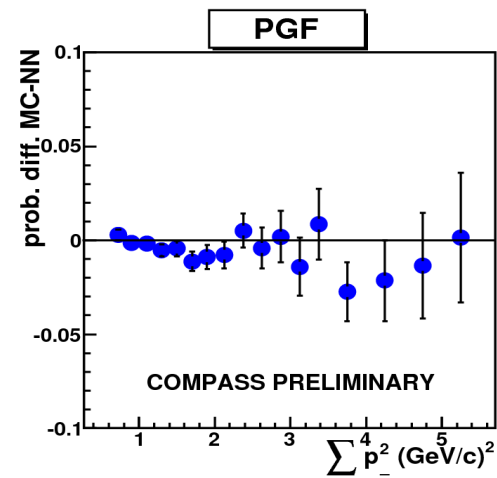
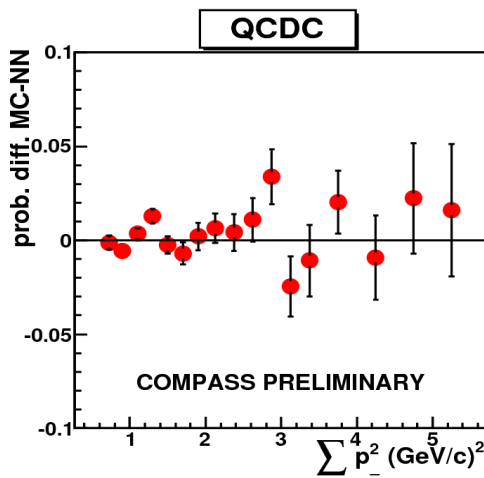
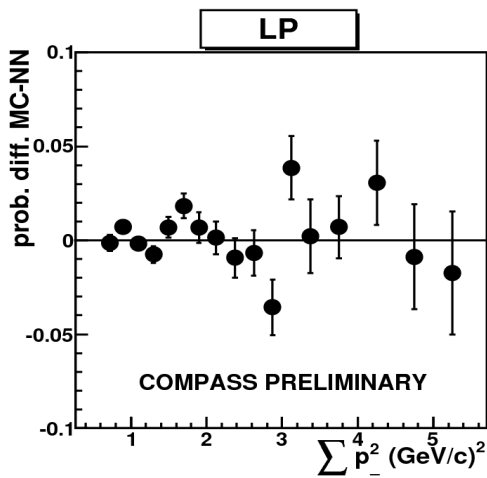
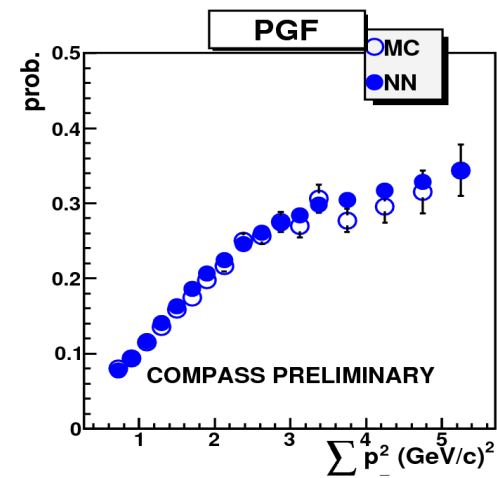
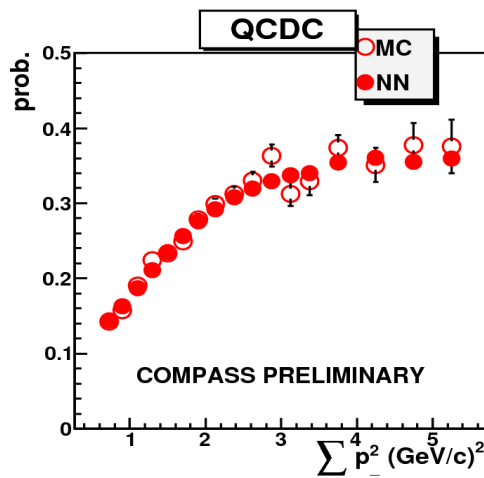
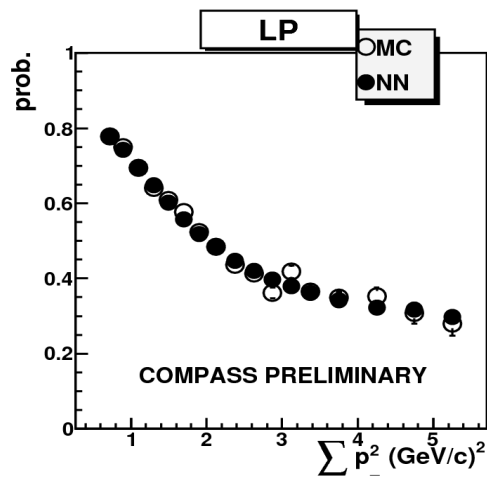
$$X_C, X_G,$$

$$f, D, P_b$$

$f, D, P_b$  are directly obtained from data.

The all the others variables have to be estimated/parametrised.

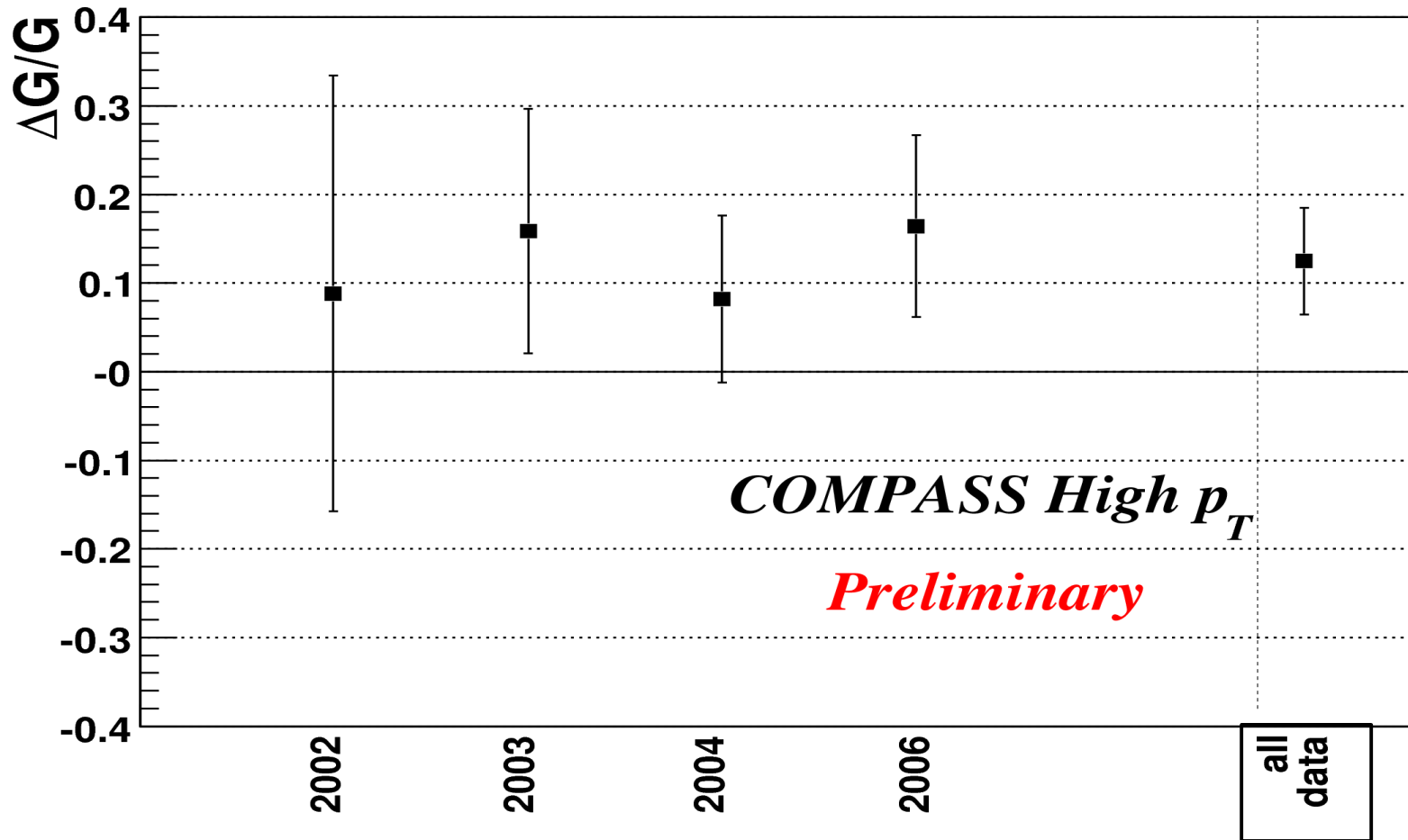
We parametrise the  $R^i$  fractions as probabilities.



$$\frac{\Delta G}{G} = 0.125 \pm 0.060 \pm 0.063$$

$$x_G = 0.09^{+0.08}_{-0.04}$$

$$\langle \mu^2 \rangle = 3.4 \text{ (GeV/c)}^2$$





Sources of Systematic Uncertainties	$\delta(\Delta G/G)$	
	High pT	Open Charm
MC Simulation	<b>0.05</b>	
Formula Simplification	<b>0.04</b>	
False Asymmetries	0.02	<b>0.08</b>
$A_1$ Parametrisation	0.02	
NN Parametrisation	0.01	
$P_B, P_T, f$	0,004	0.01
$a_{LL}$		0.01
$s/(s+b)$		0.01
Total	0.06	0.08

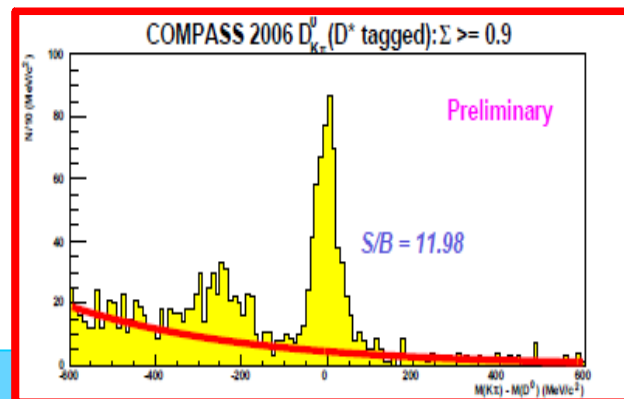
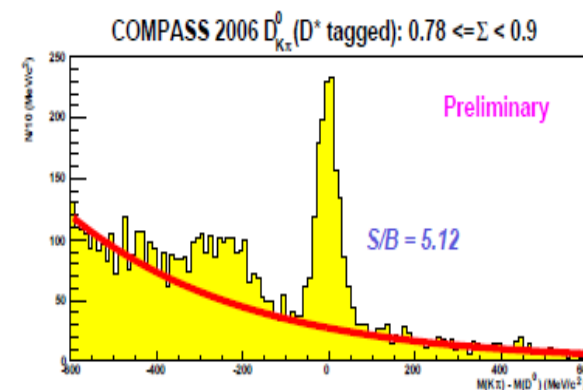
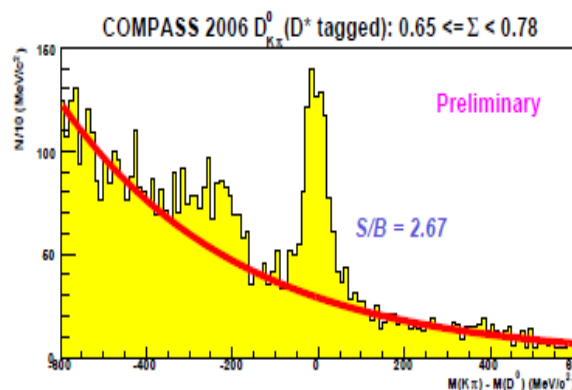
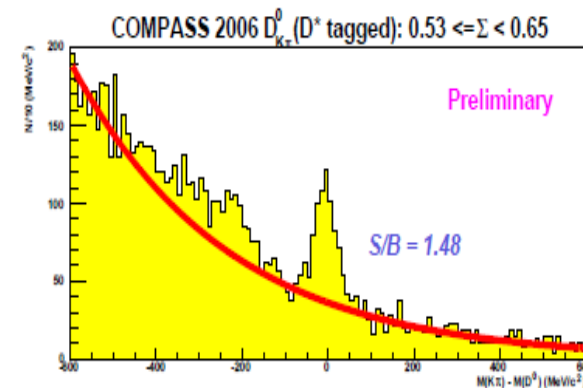
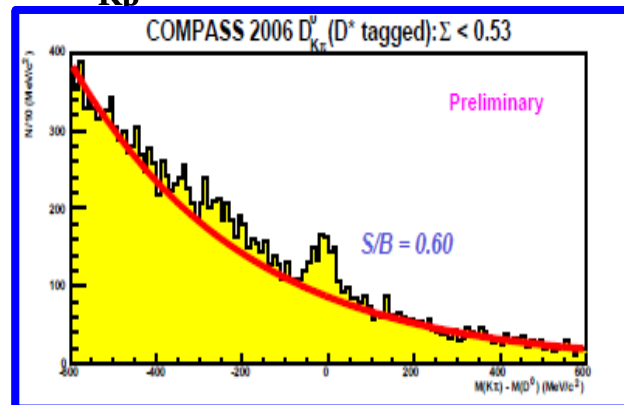
# S/(S+B): Obtaining final probabilities for a D<sup>0</sup> candidate

- Events with small  $S/(S+B)_{NN}$ 
  - Mostly combinatorial background is selected

S/(S+B) is obtained from a fit inside this bins (correcting with the NN parameterisation)

- Events with large  $S/(S+B)_{NN}$ 
  - Mostly Open Charm are selected

D<sup>0</sup><sub>K<sub>D</sub></sub> tagged spectrum in bins of  $\Sigma = S/(S+B)_{NN}$

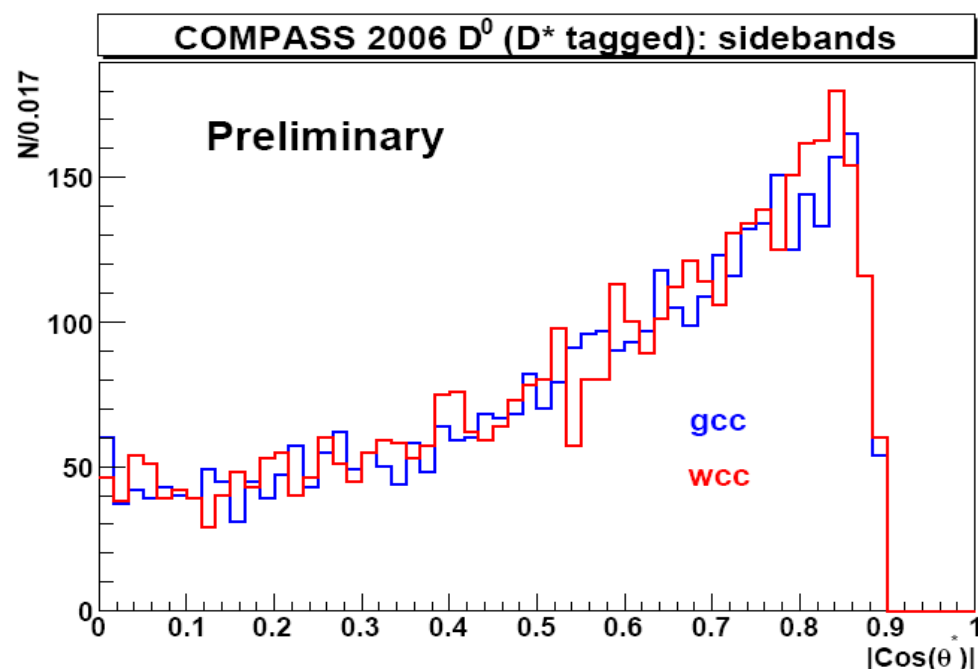
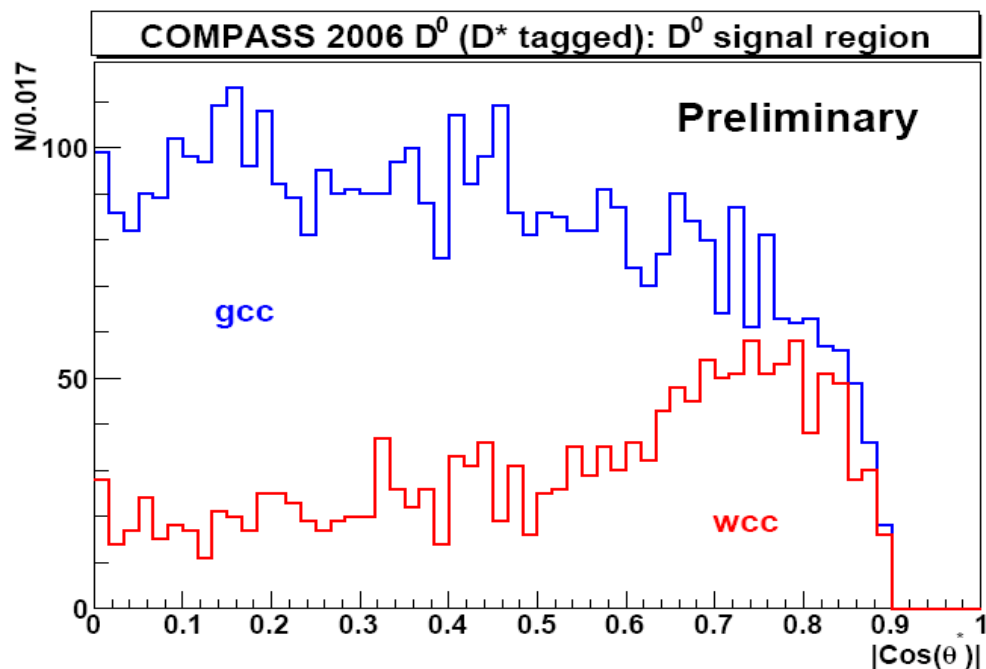


$$\delta \left( \frac{\Delta G}{G} \right) \propto \frac{1}{\text{FOM}}$$

# Neural Network qualification of events

- **Two real data samples** (with the same cuts applied) **are compared by a Neural Network** (using some kinematic variables as a learning vector):
  - **Signal model**  $\rightarrow$  **gcc** =  $\mathbf{K}^+\pi^-\pi_s^- + \mathbf{K}^-\pi^+\pi_s^+$  ( $D^0$  spectrum: signal + background)
  - **Background model**  $\rightarrow$  **wcc** =  $\mathbf{K}^+\pi^+\pi_s^- + \mathbf{K}^-\pi^-\pi_s^+$  (no  $D^0$  is allowed)
- **If the background model is good enough:** The Neural Network is able to distinguish the signal from the combinatorial background on a event by event basis (inside gcc)

## Example of a good learning variable



# Analysing power (muon-gluon asymmetry $a_{LL}$ )

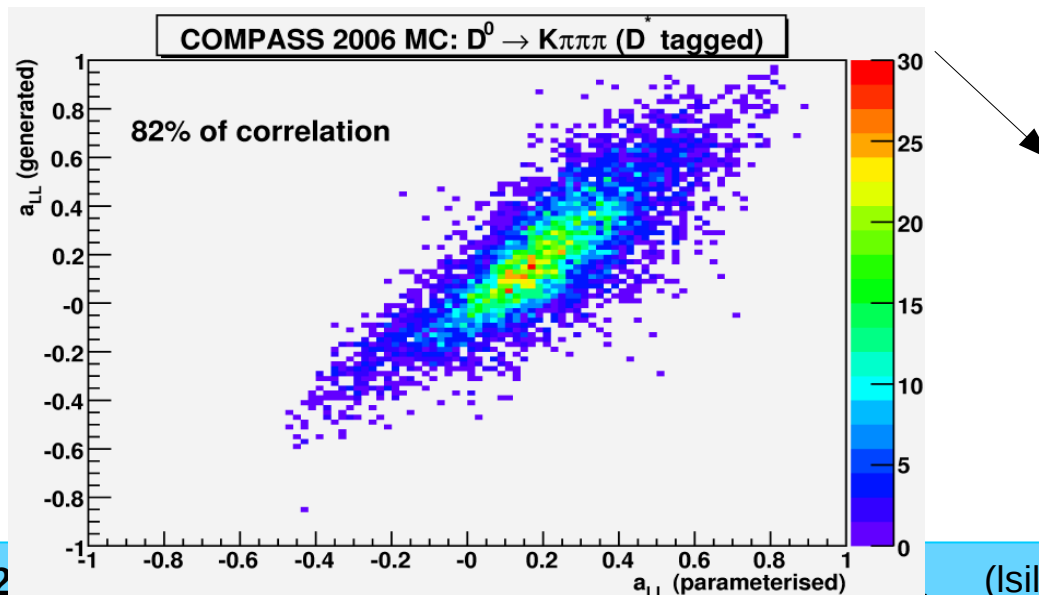
- $a_{LL}$  is dependent on the full knowledge of the partonic kinematics:

$$a_{LL} = \frac{\Delta \sigma^{PGF}}{\sigma_{PGF}}(y, Q^2, x_g, z_C, \phi)$$



Can't be experimentally obtained: only one charmed meson is reconstructed

- $a_{LL}$  is obtained from Monte-Carlo (in LO), to serve as input for a Neural Network parameterisation on some reconstructed kinematical variables:  $y$ ,  $x_{Bj}$ ,  $Q^2$ ,  $z_D$  and  $p_T$



Parameterised  $a_{LL}$ , shows a strong correlation with the generated one (using AROMA)

- The AROMA generator is used to simulate the phase space for the **NLO (PS on) / LO (PS off)** calculations of  $a_{LL}$ . The resulting  $D^0$  mesons are reconstructed in the COMPASS spectrometer like real events. The respective  $a_{LL}$  distributions are:

