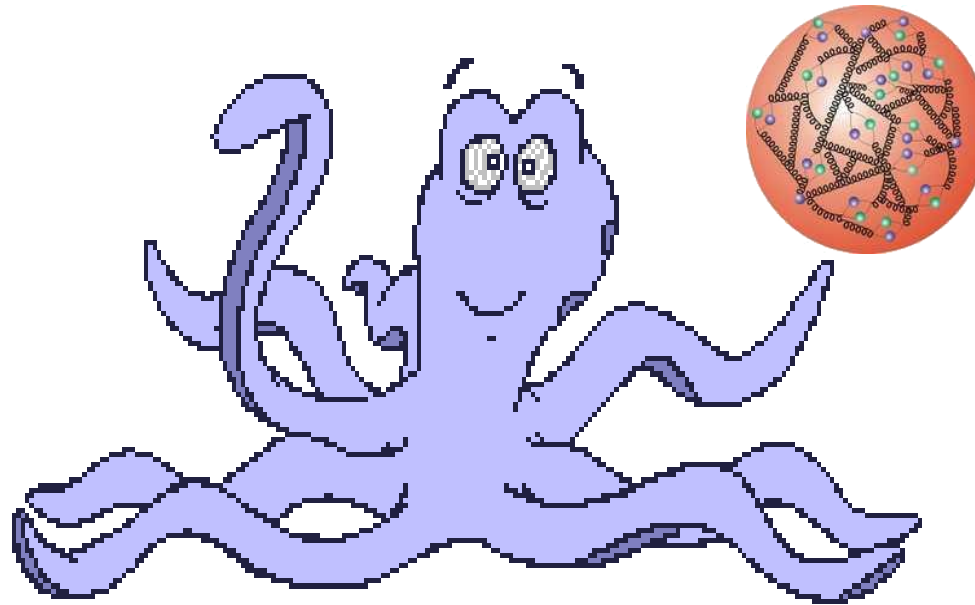


Multiple Parton Interactions in Photoproduction at H1 at HERA.



Lluís Martí Magro (DESY)
DPG, Heidelberg. 7th of March, 2007.

Multiple Parton Interactions (MI)

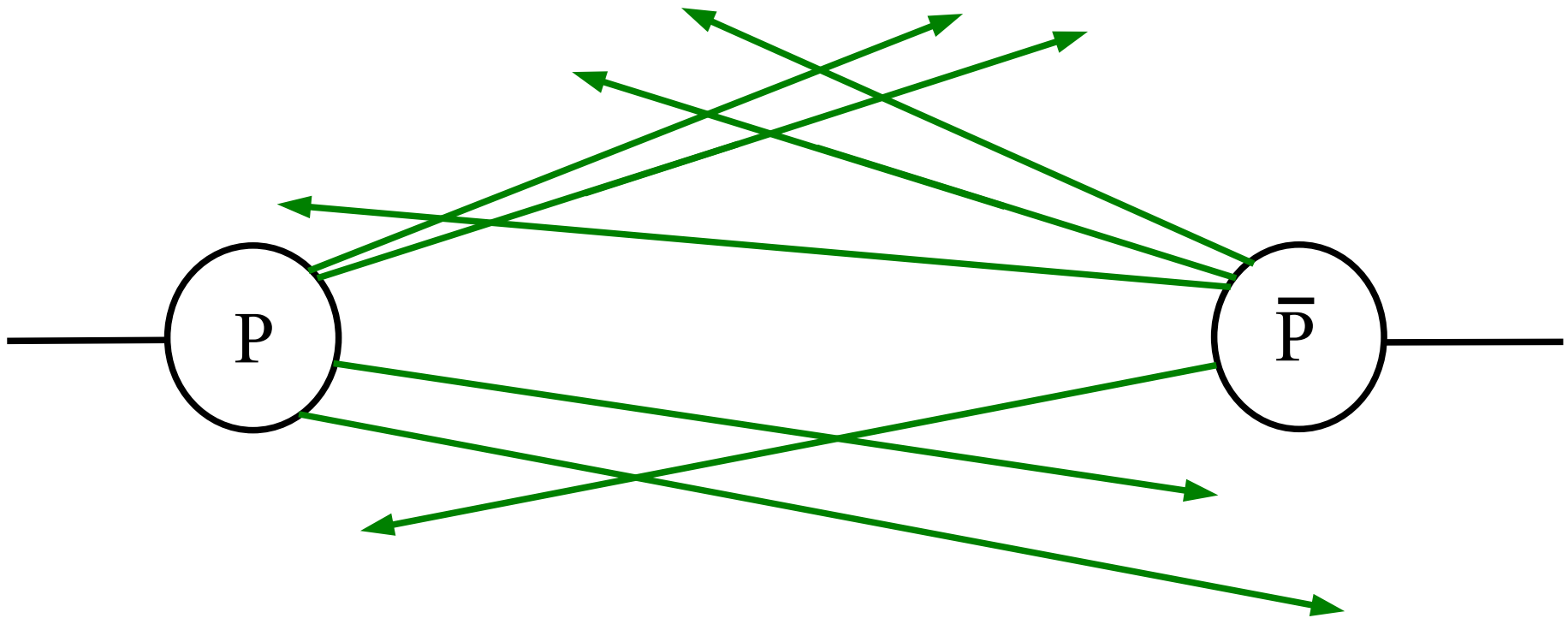
1

Contents:

- I. Introduction & motivation to MI (very short).
- II. Sample definition: all inclusive flavour and charm in PhP.
- III. New measurement at HERA, i.e. observable: multiplicity as a function of φ .

Introduction: hadron-hadron collision

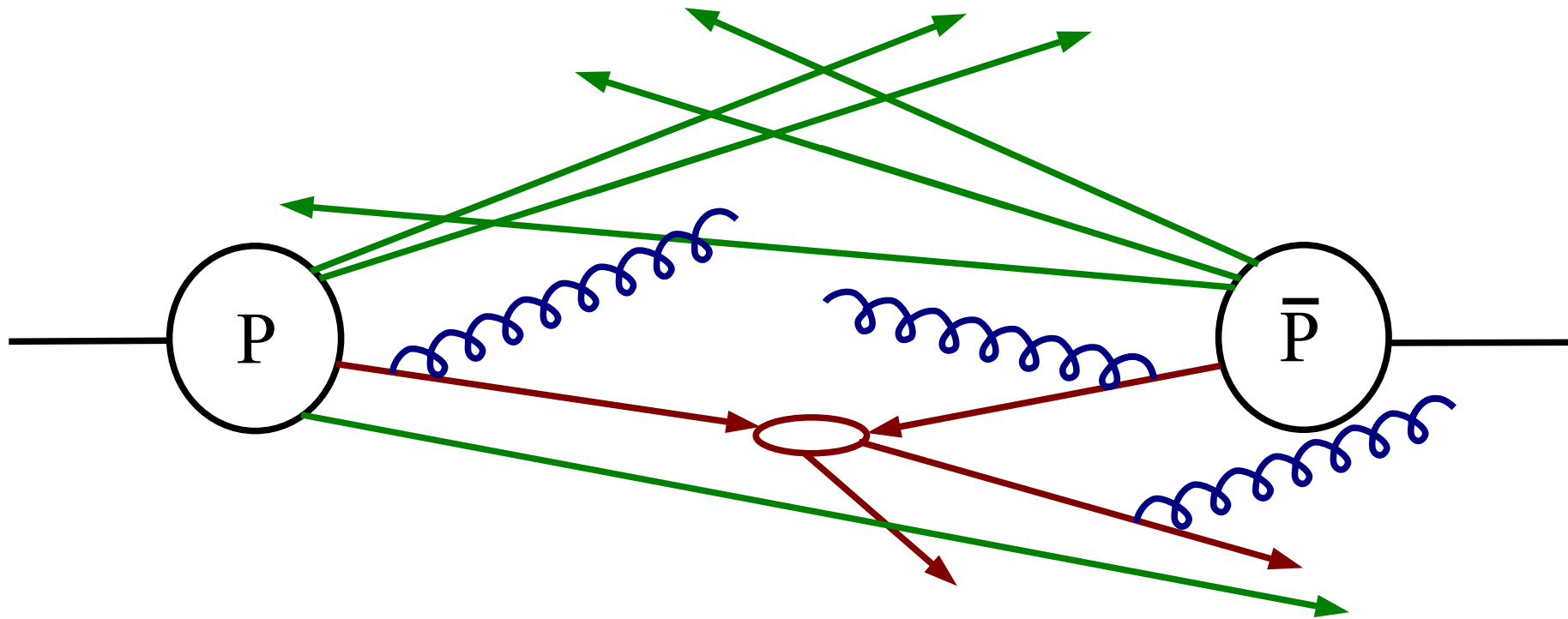
- ✗ “Minimum- Bias” event: only soft particles are produced.



No hard interaction \longrightarrow we cannot apply pQCD.

Introduction: hadron-hadron collision

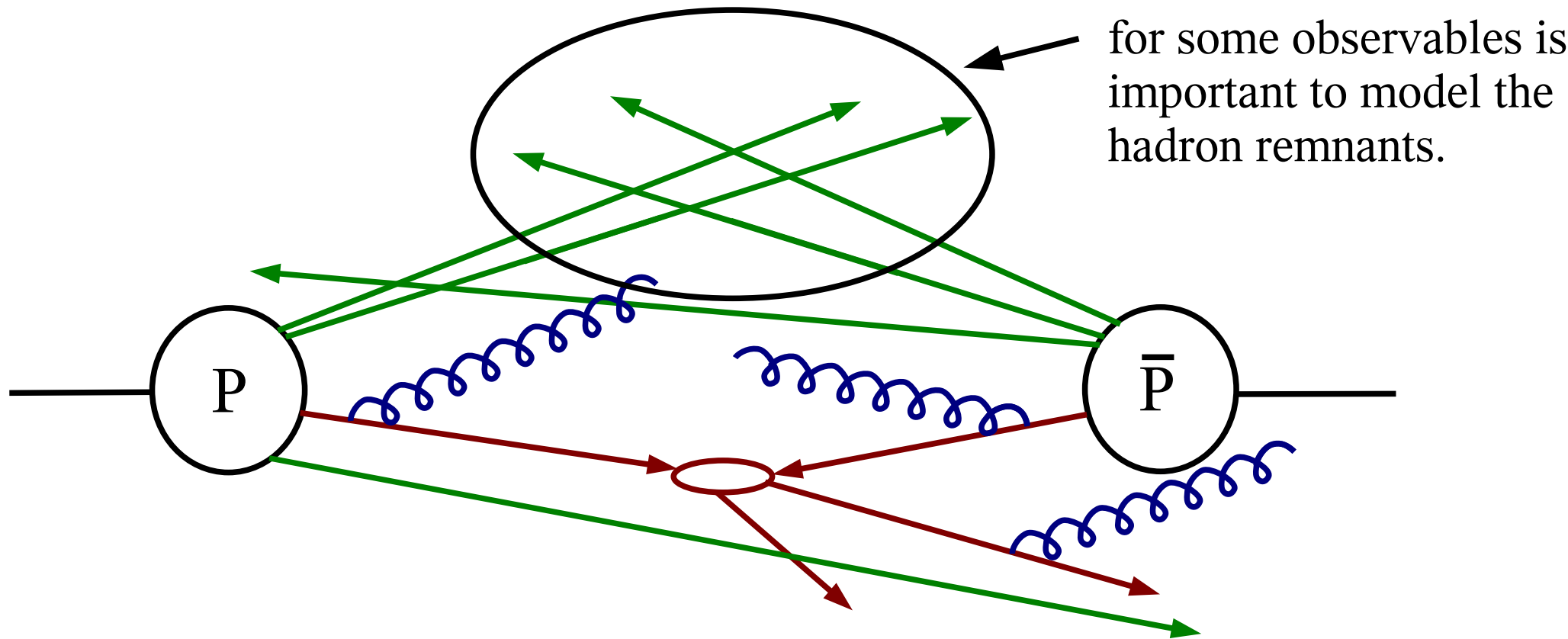
✘ Hard collision: apply pQCD



but there is also a soft (or maybe not so soft) component

Introduction: hadron-hadron collision

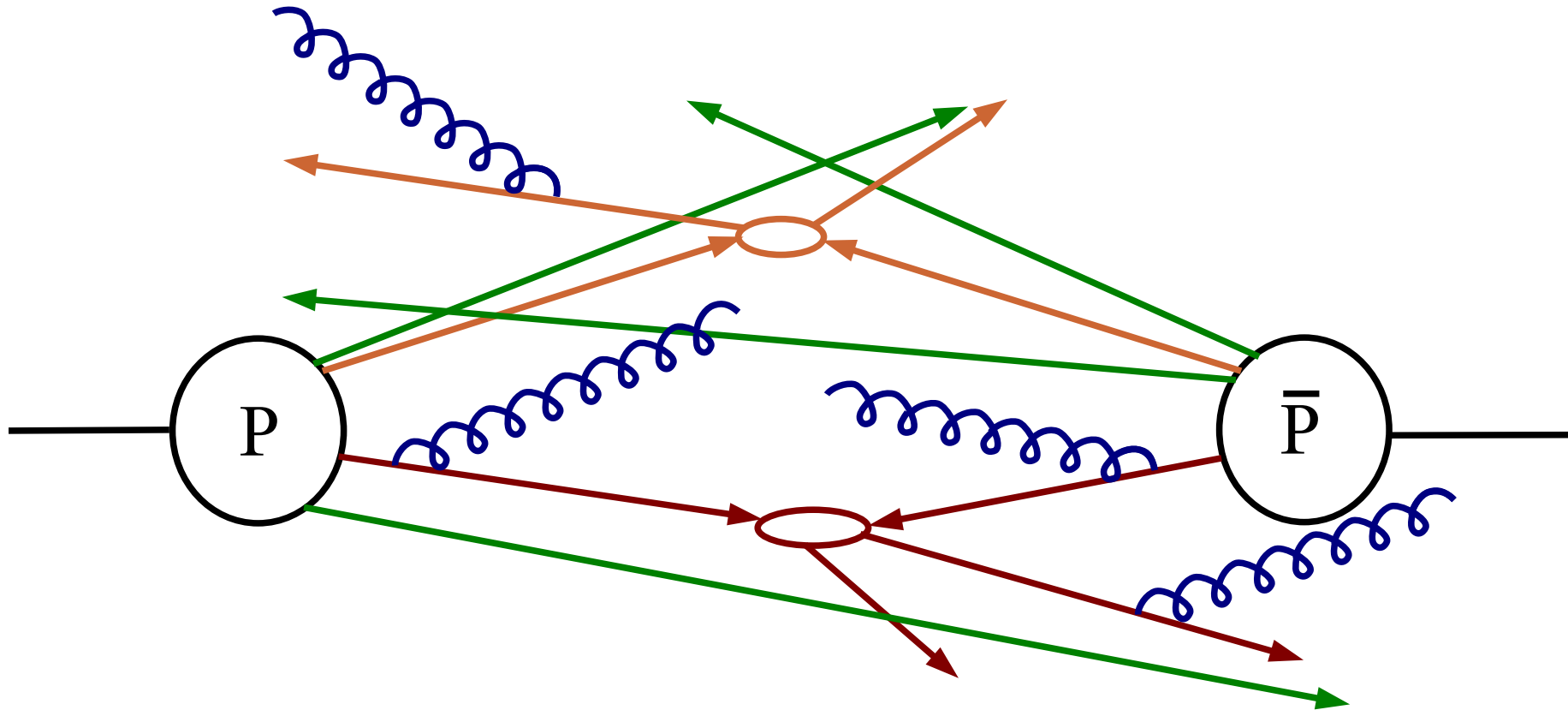
✗ Hard collision: apply pQCD



but there is also a soft (or maybe not so soft) component

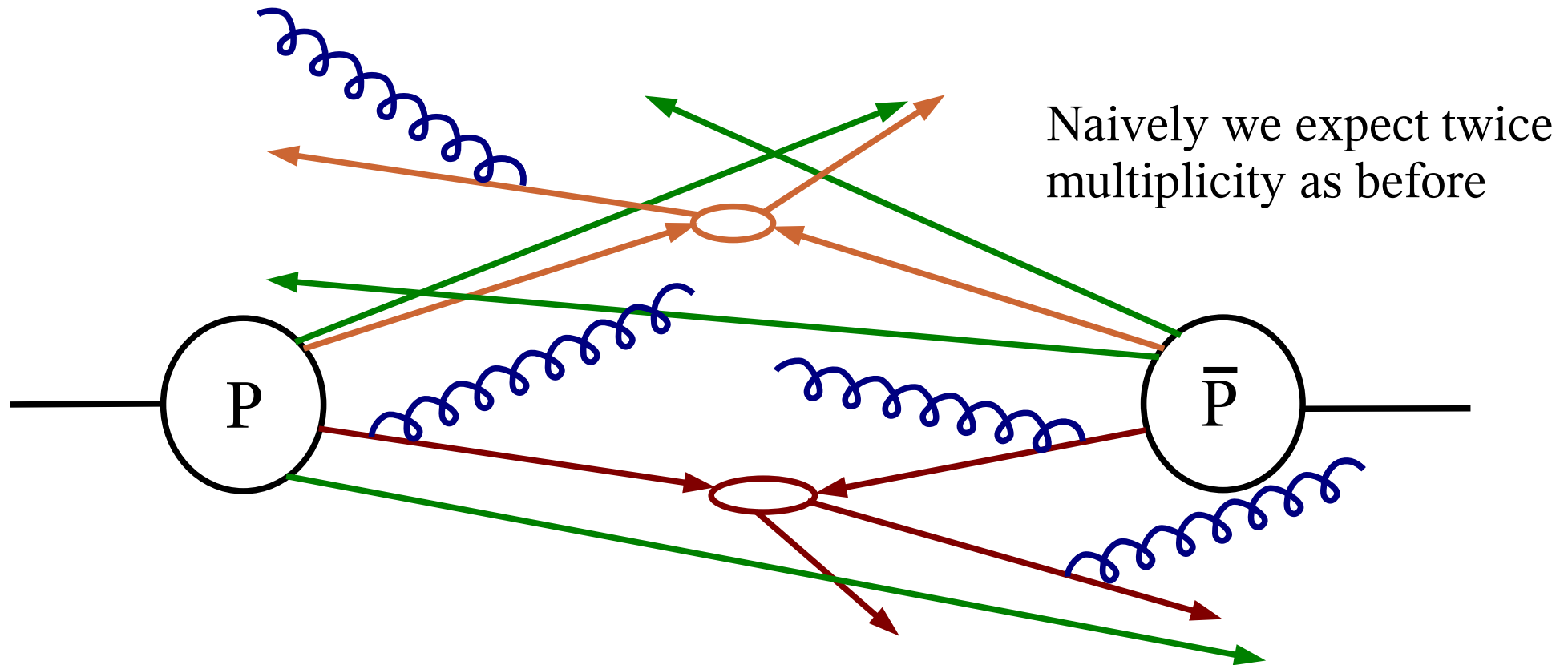
Introduction: hadron-hadron collision

- MI: go down to low transverse momenta (as far as possible) and model the remnants.



Introduction: hadron-hadron collision

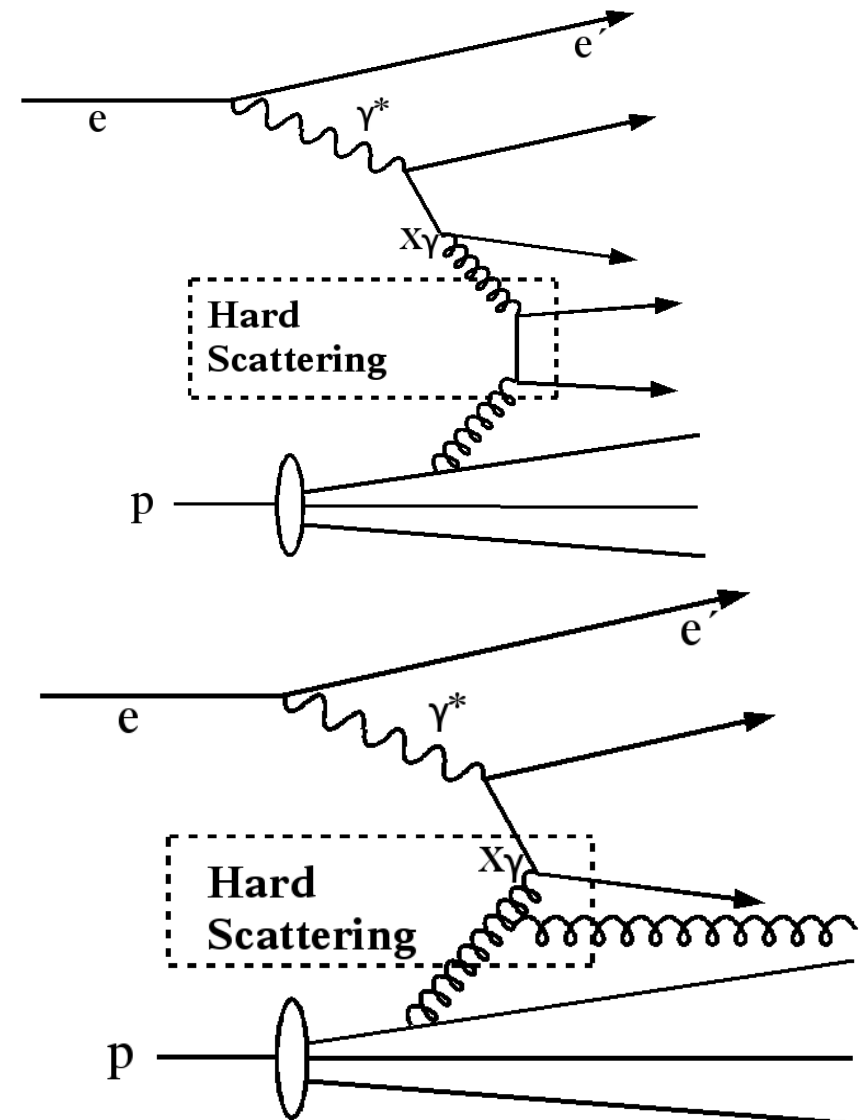
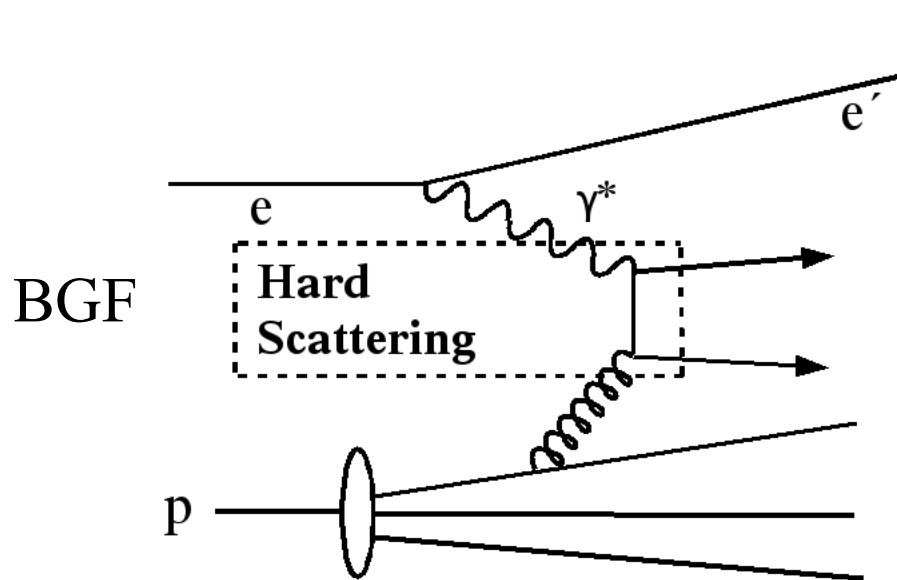
- ✗ MI: go down to low transverse momenta (as far as possible) and model the remnants.



the question is how to do it....

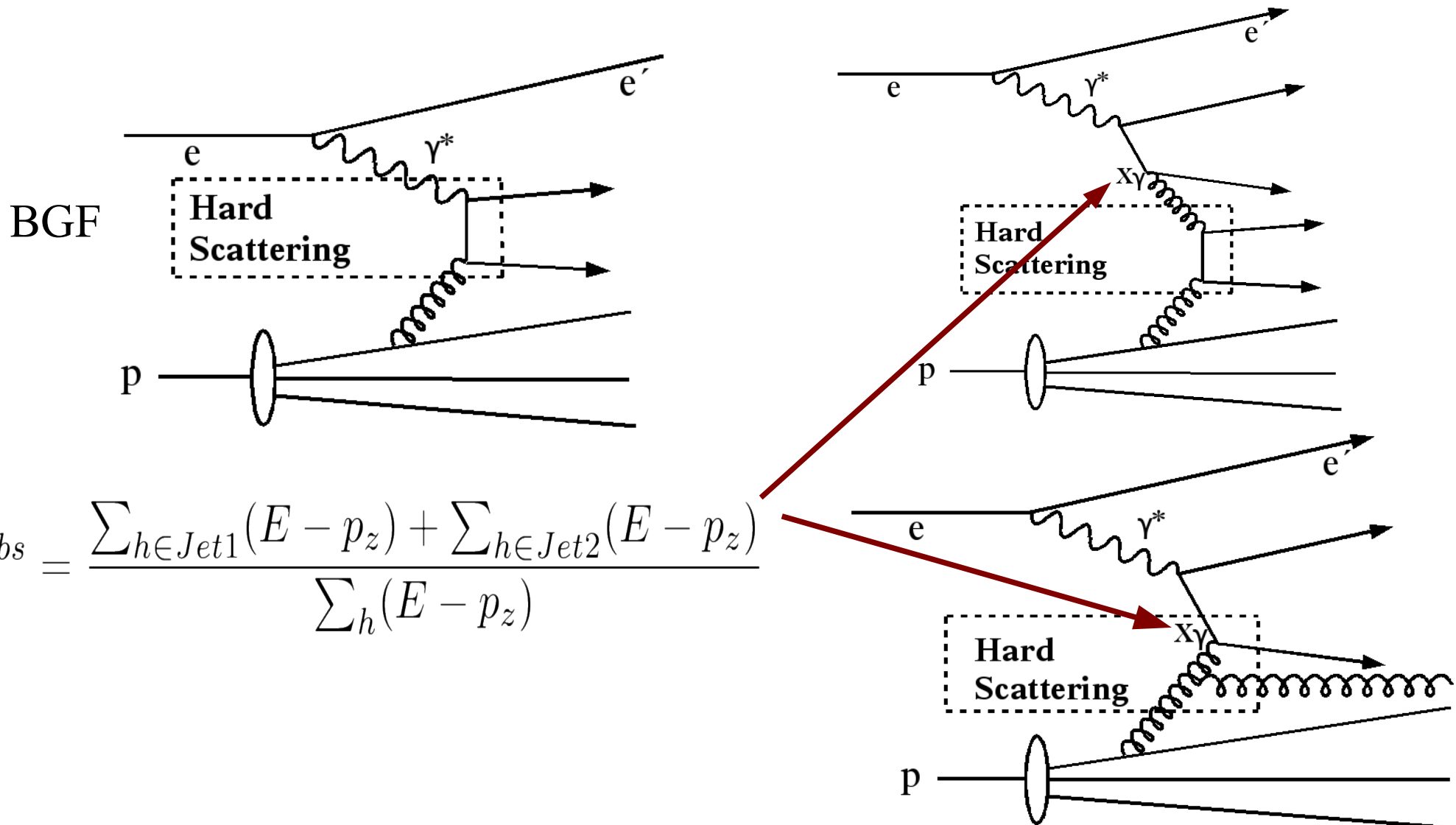
Introduction: Pythia

Three types of events are generated: BGF (direct), resolved and excitation events.



Introduction: Pythia

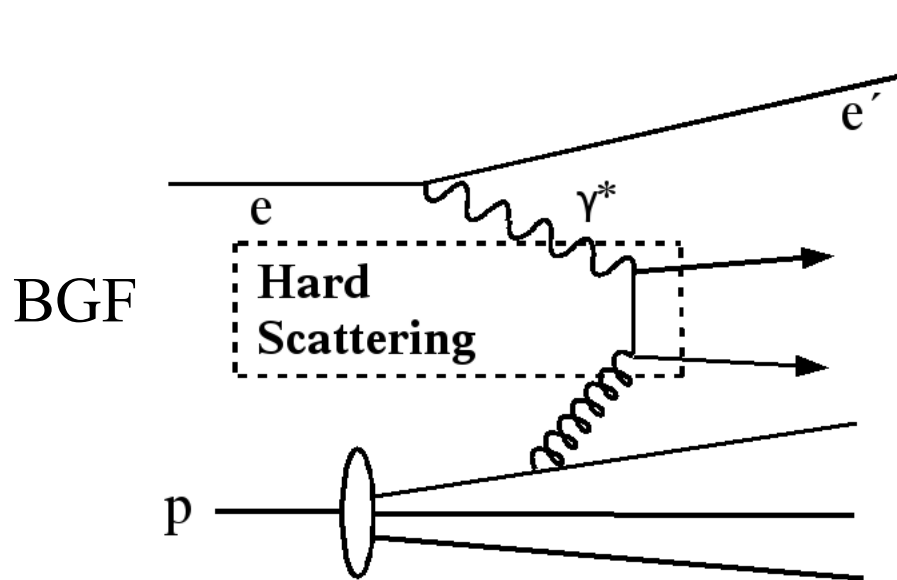
Three types of events are generated: BGF (direct), resolved and excitation events.



$$x_{\gamma}^{obs} = \frac{\sum_{h \in Jet1} (E - p_z) + \sum_{h \in Jet2} (E - p_z)}{\sum_h (E - p_z)}$$

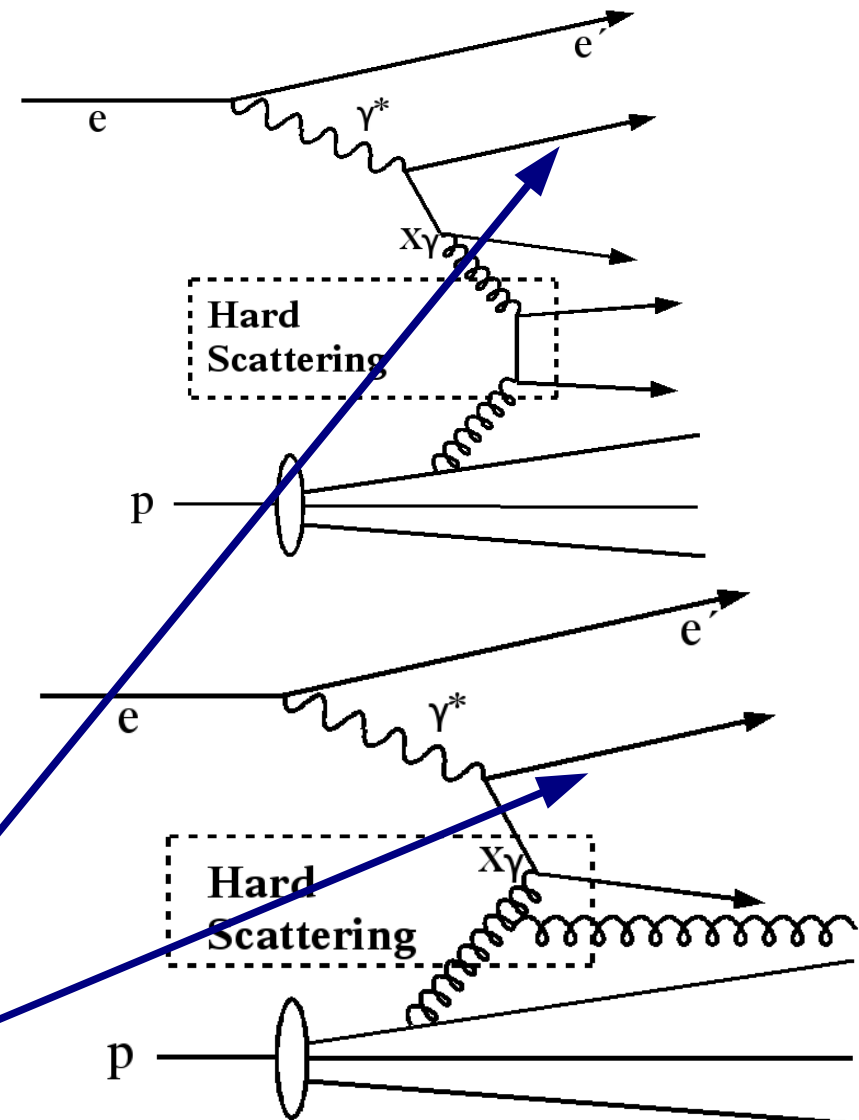
Introduction: Pythia

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$$x_{\gamma}^{obs} = \frac{\sum_{h \in Jet1} (E - p_z) + \sum_{h \in Jet2} (E - p_z)}{\sum_h (E - p_z)}$$

Photon remnants like in hadron-hadron collisions



Sample selections

Event Selection:

- ✗ Common selection: photoproduction
- ✗ DiJet sample with: $P_t > 7(6) \text{ GeV}$ Kt clustering algorithm (pt weighted recom. scheme)
 $|\eta| < 1.5$
- ✗ Charm sample:
 - Muon $P_t > 2.5 \text{ GeV}$
 - $|\eta| < 1.5$

The highest P_t Jet is the “Leading Jet”

Observables

Particle Multiplicity: average number of measured particles per event as a function of φ .

✗ Charged particles:

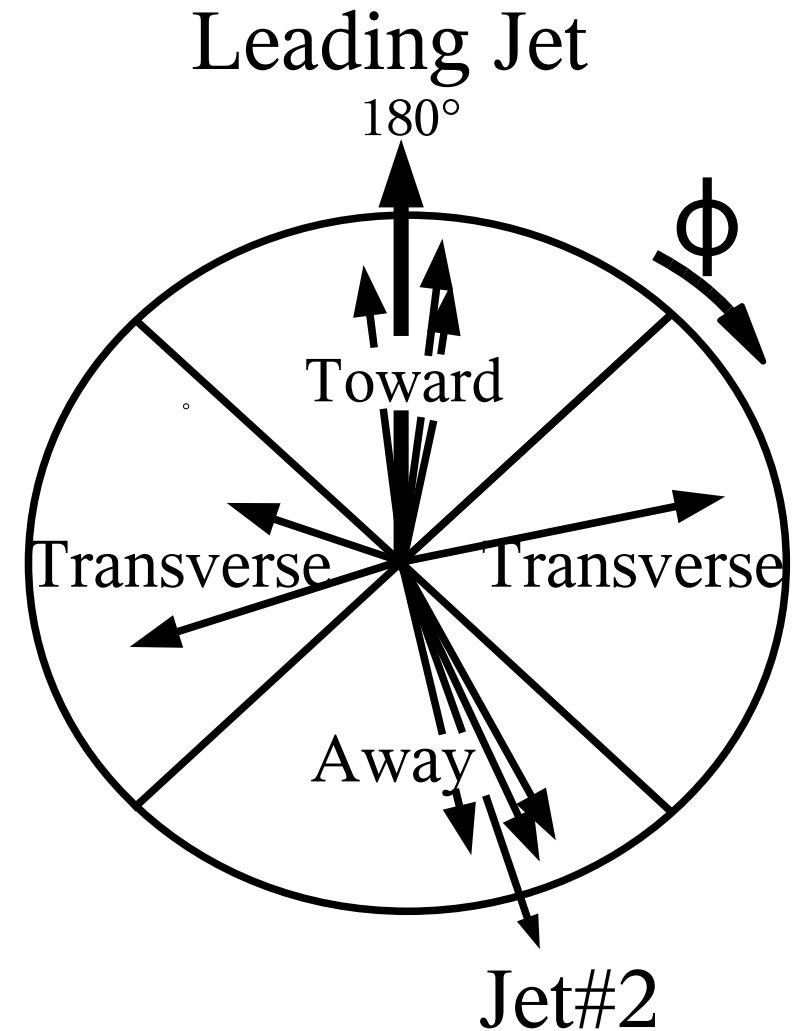
✗ with $P_t > 150\text{MeV}$ and $|\eta| < 1.5$

We define three regions:

✗ Toward: $120^\circ < |\varphi|$

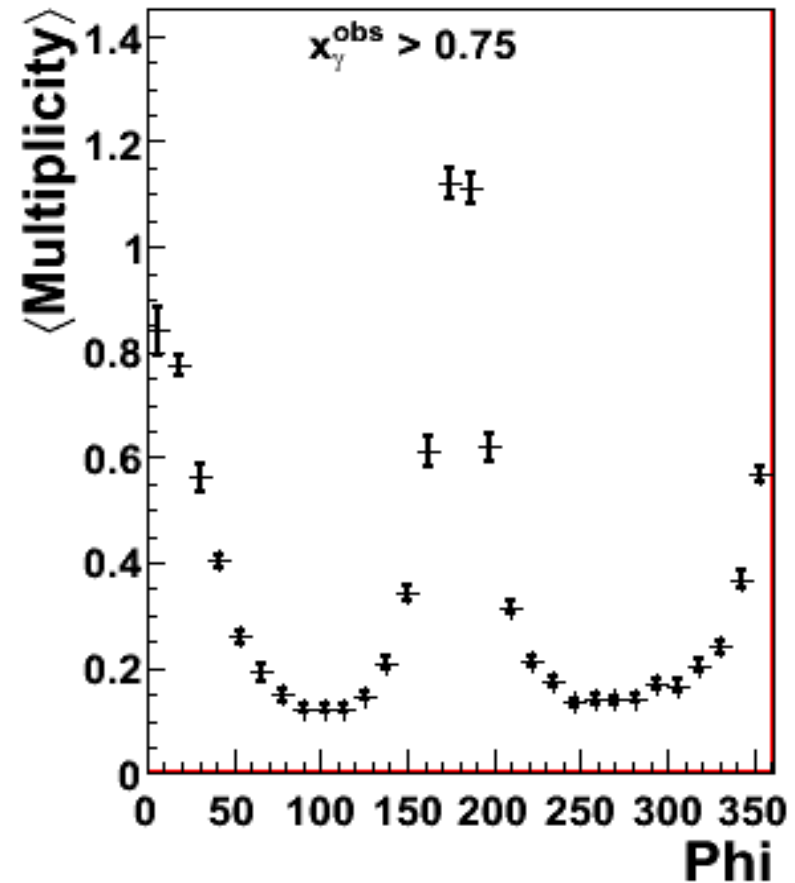
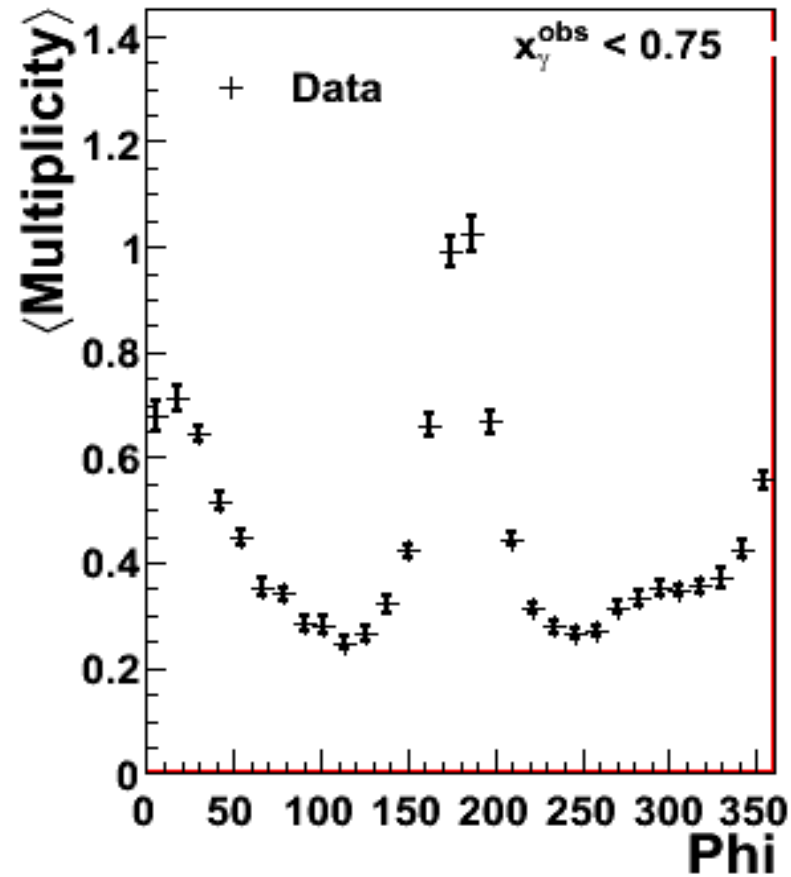
✗ Transverse: $60^\circ < |\varphi| < 120^\circ$

✗ Away: $|\varphi| < 60^\circ$



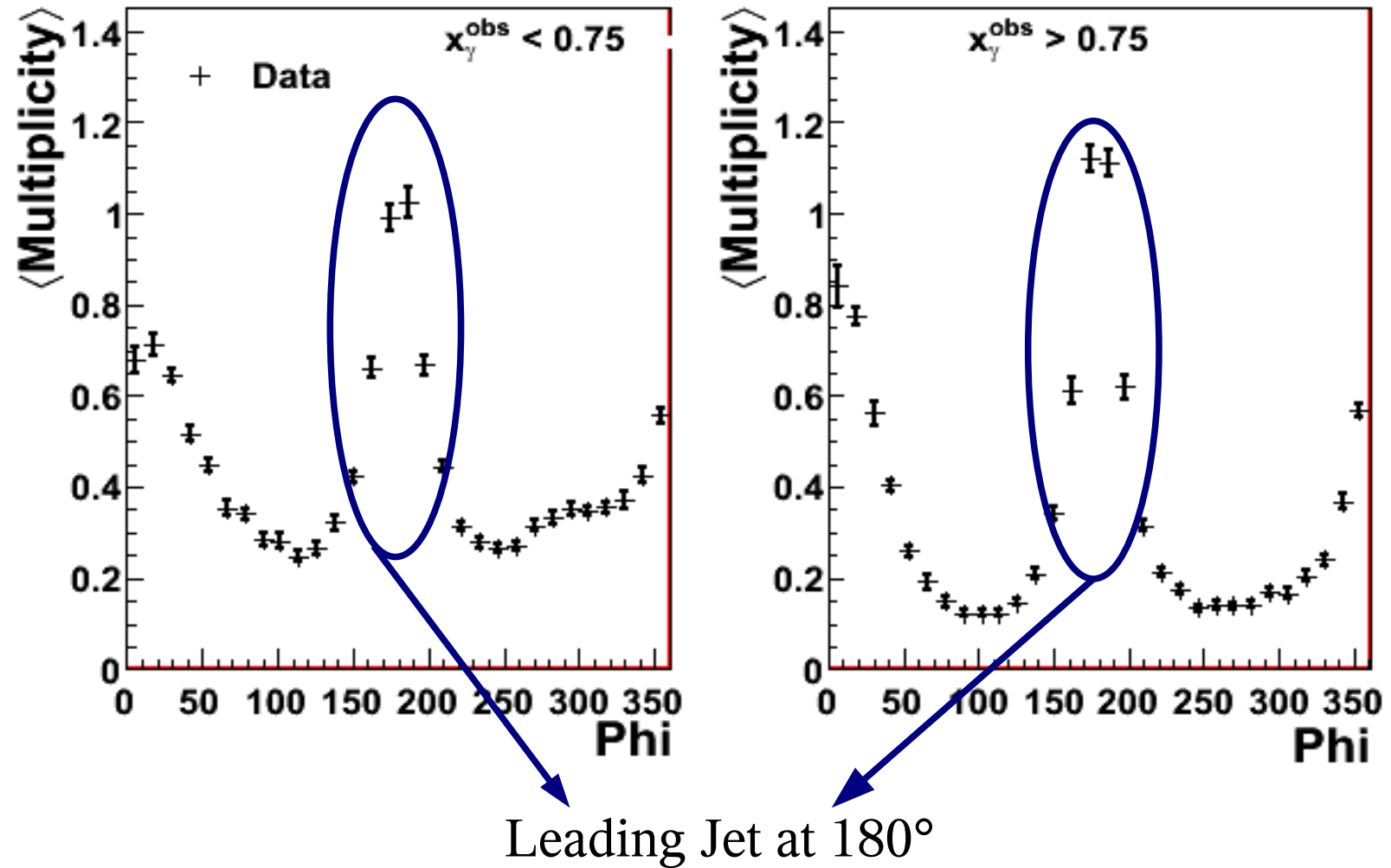
Multiplicity

Particle multiplicity: Average number of measured particles per event.



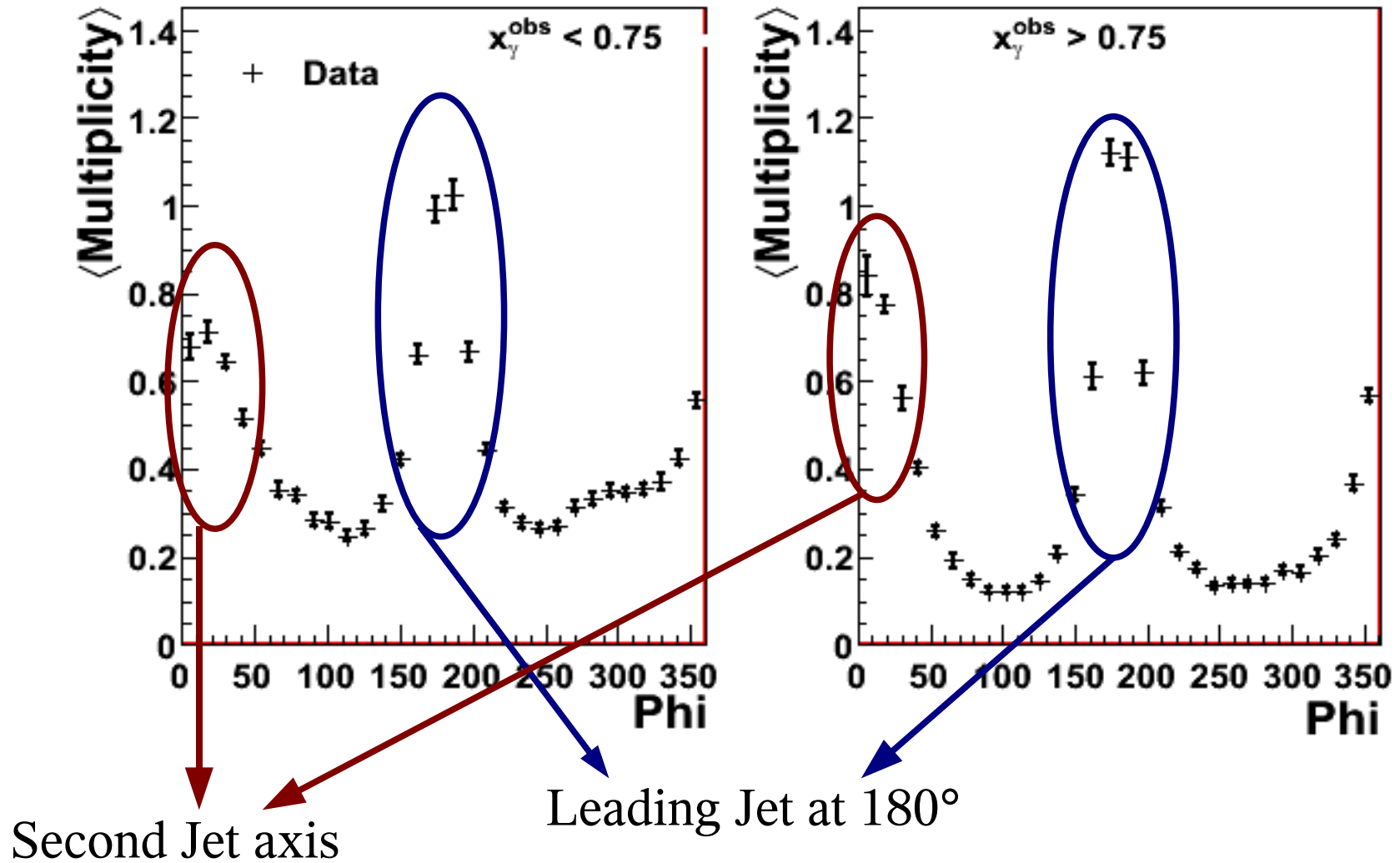
Multiplicity

Particle multiplicity: Average number of measured particles per event.



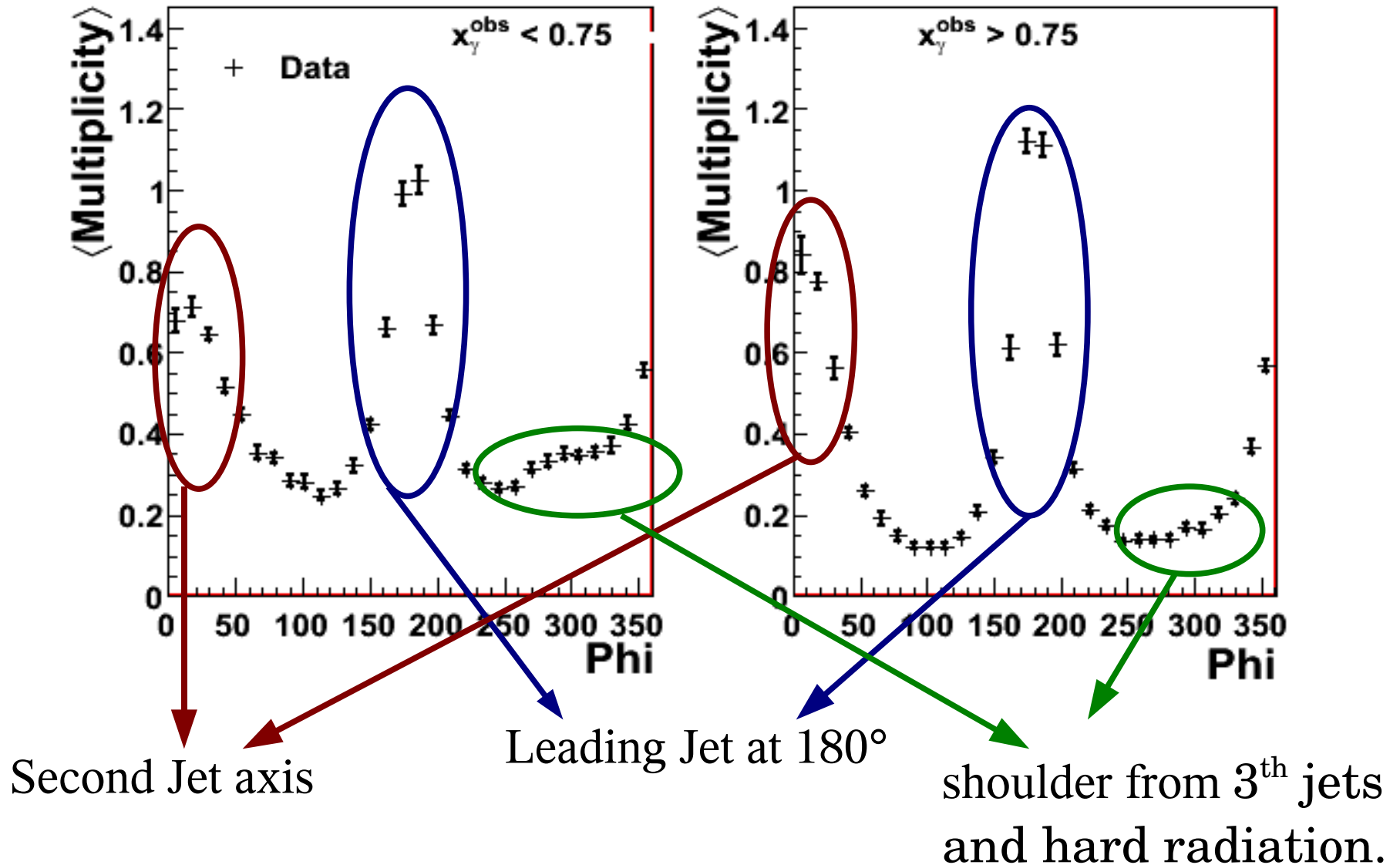
Multiplicity

Particle multiplicity: Average number of measured particles per event.



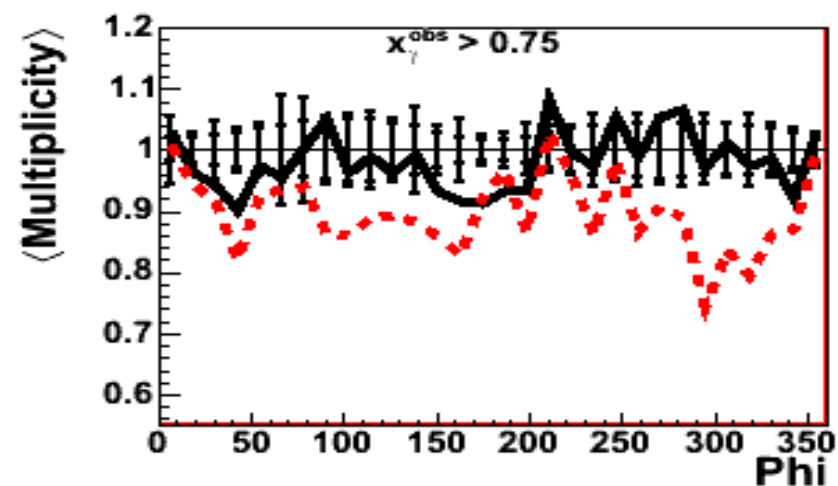
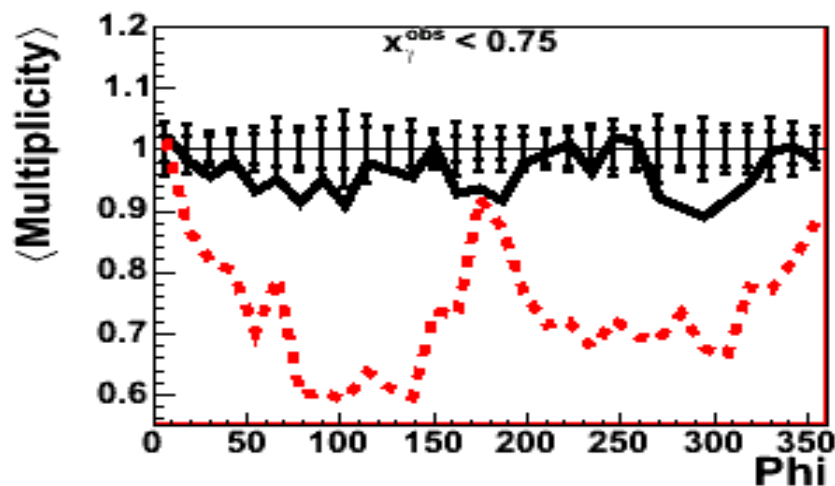
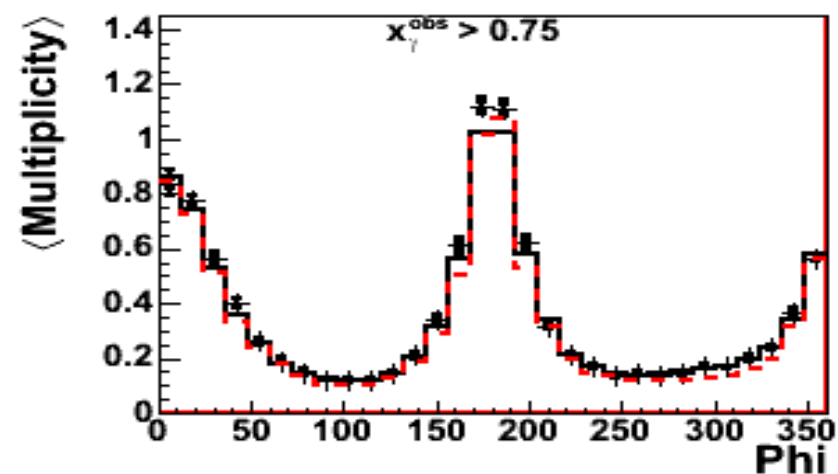
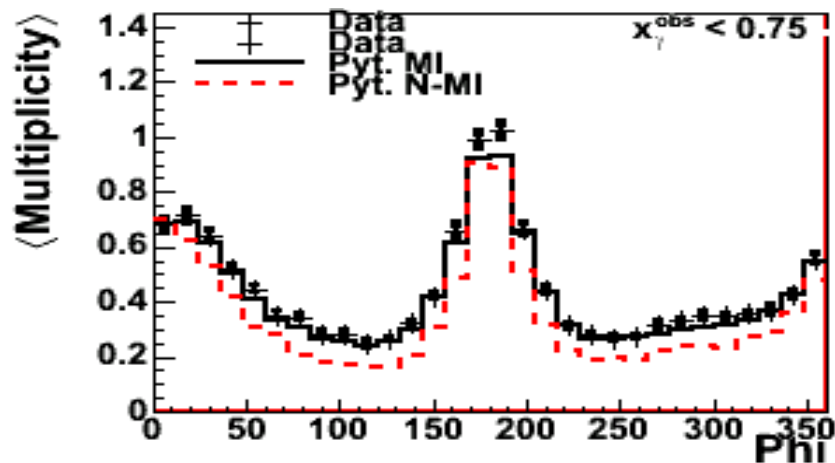
Multiplicity

Particle multiplicity: Average number of measured particles per event.



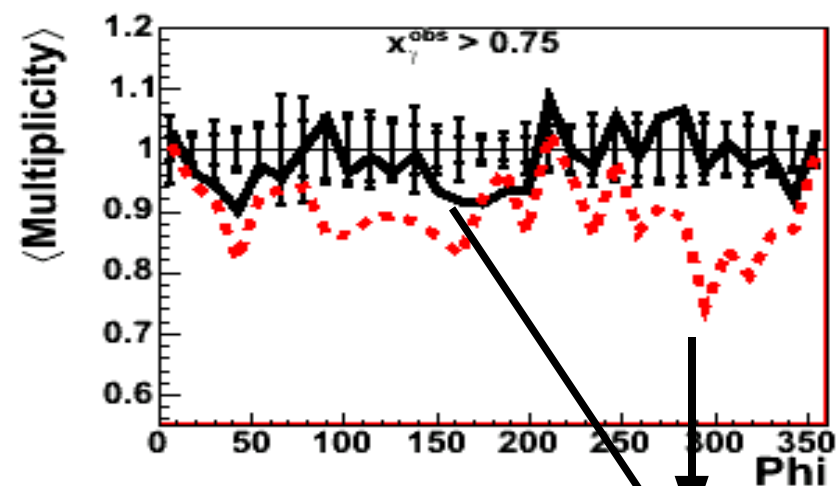
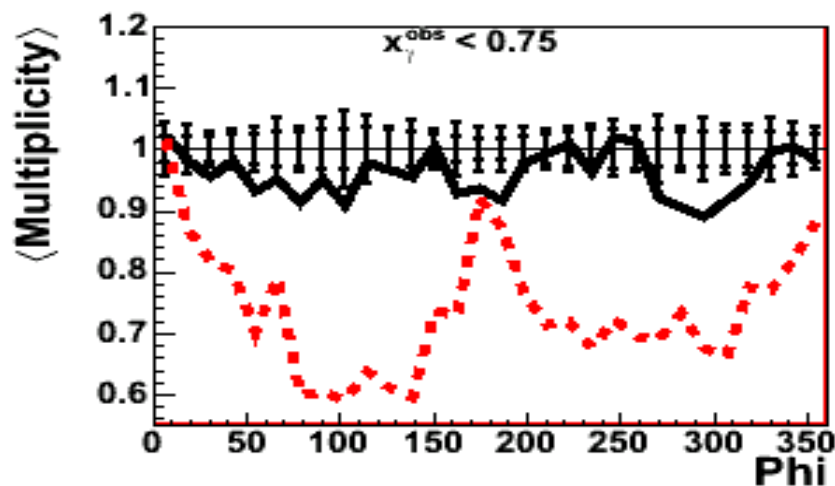
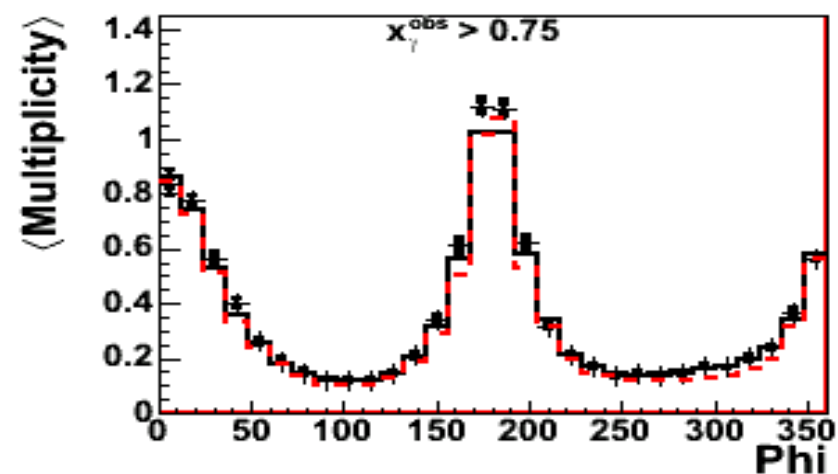
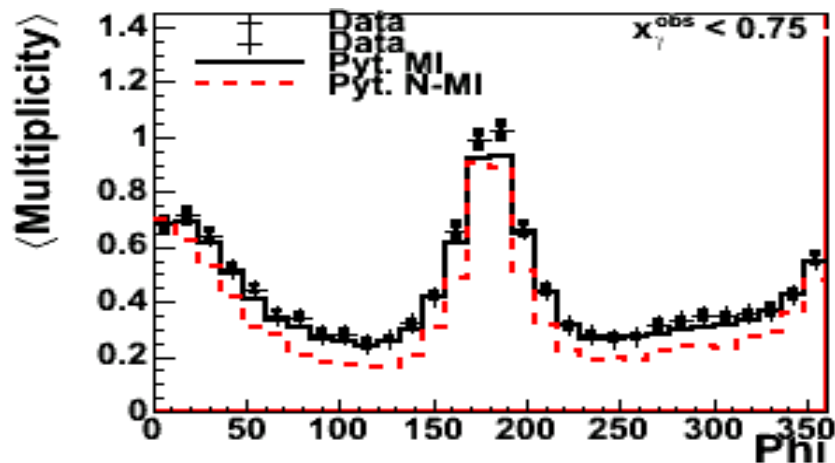
Dijet sample track multiplicity

Particle multiplicity: Average number of measured particles per event.



Dijet sample track multiplicity

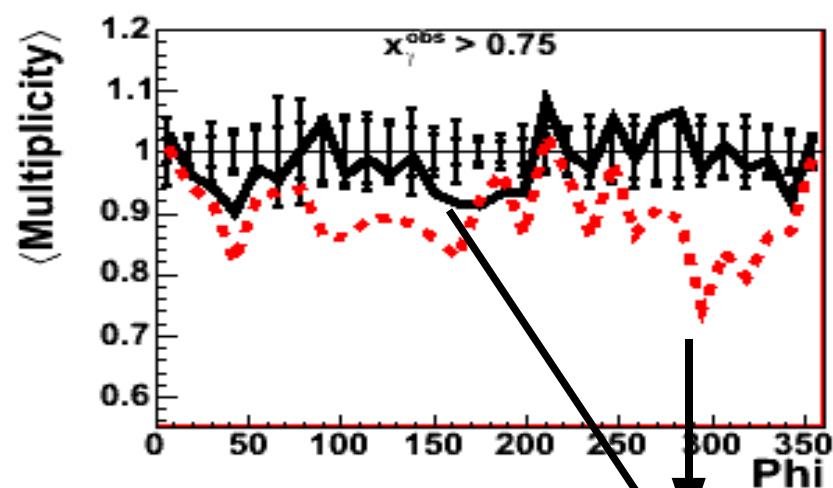
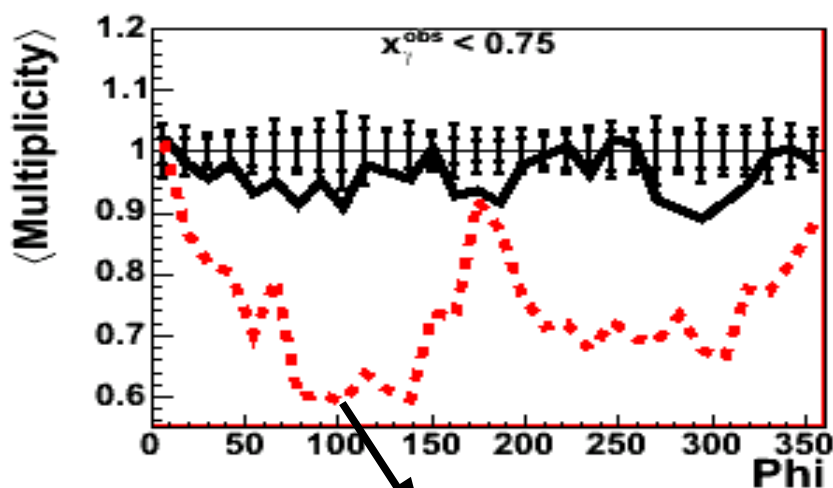
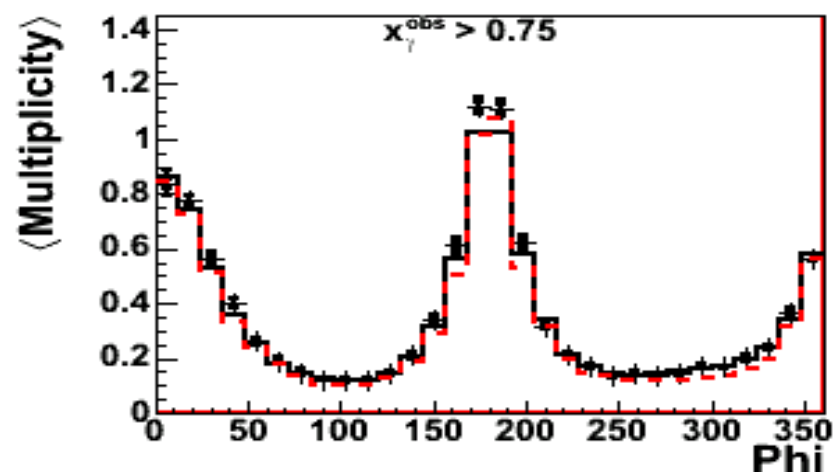
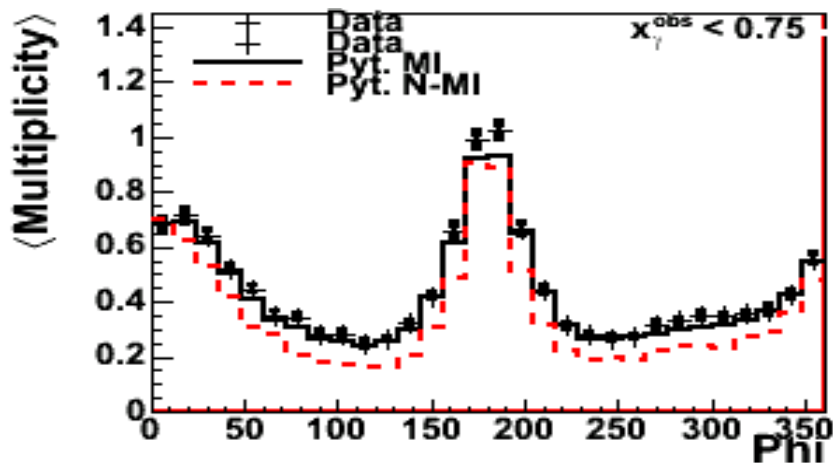
Particle multiplicity: Average number of measured particles per event.



Ratio Data/MC

Dijet sample track multiplicity

Particle multiplicity: Average number of measured particles per event.



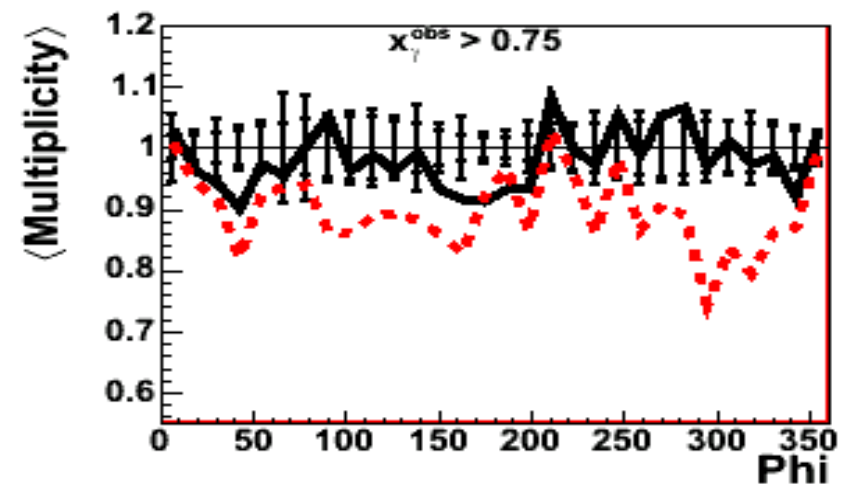
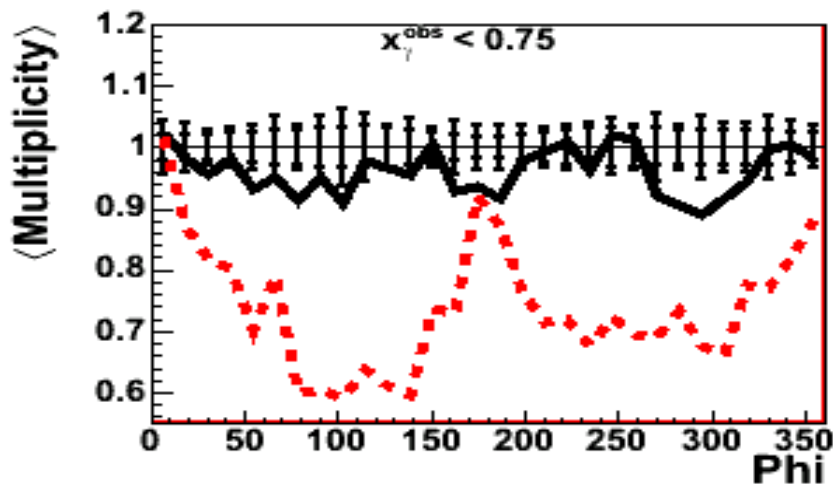
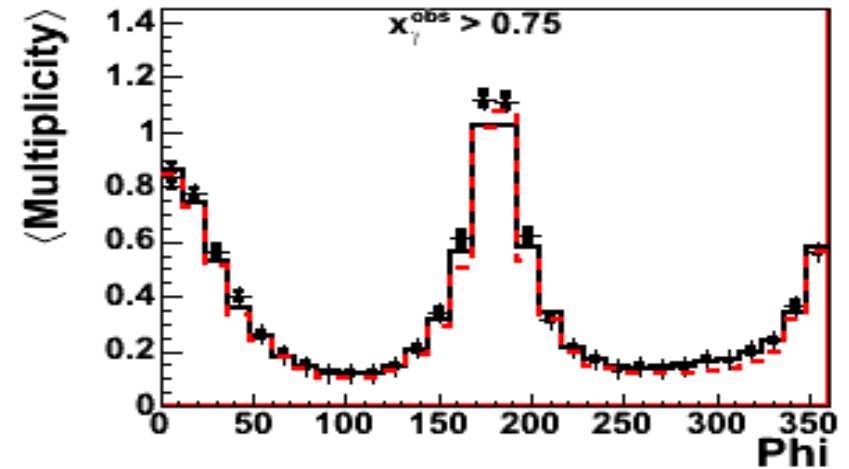
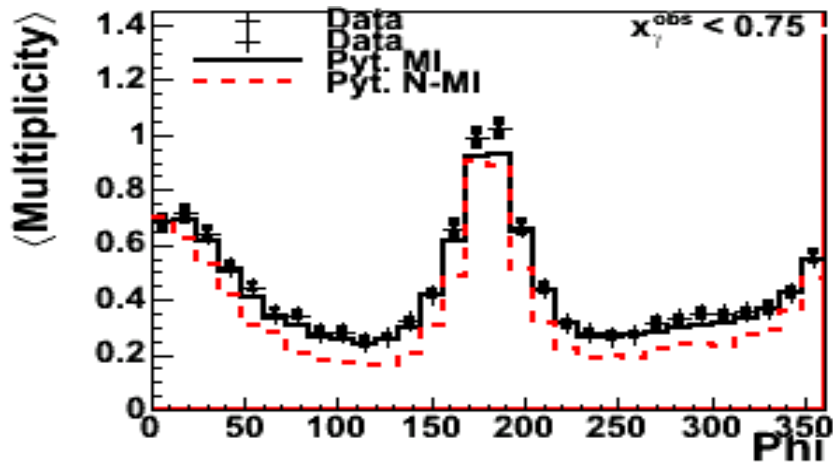
$\sim 6\sigma$ between data and
Pythia w/o MI

Total uncertainty is less
than 10%

Ratio Data/MC

Dijet sample track multiplicity

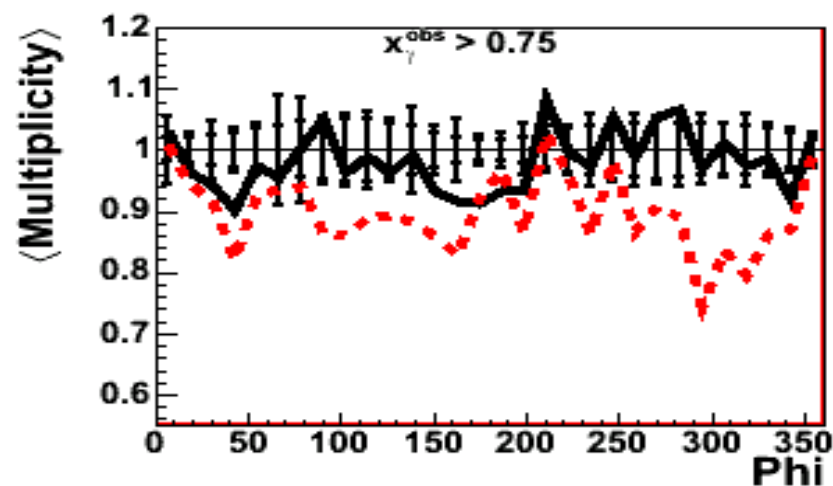
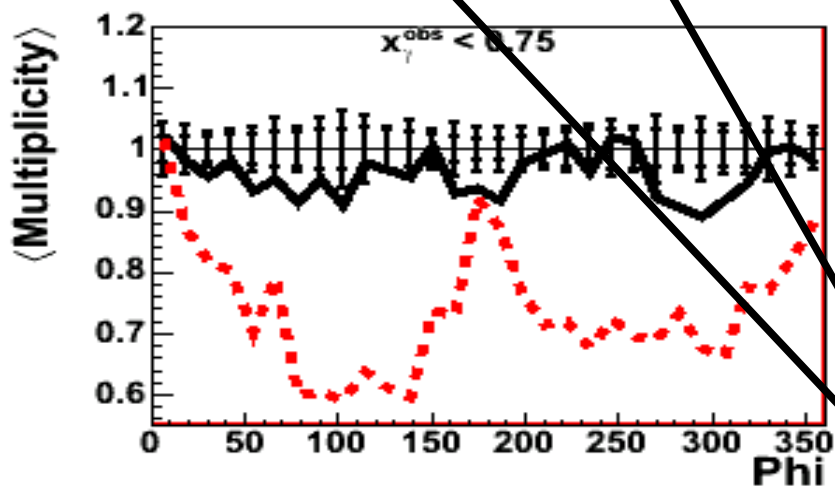
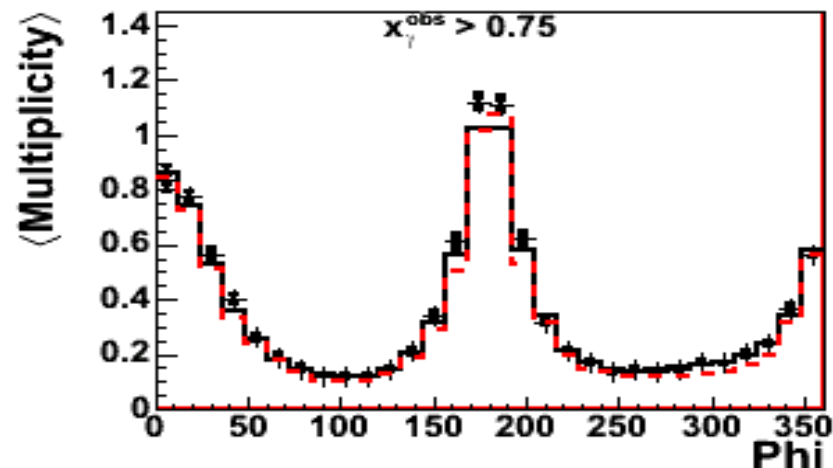
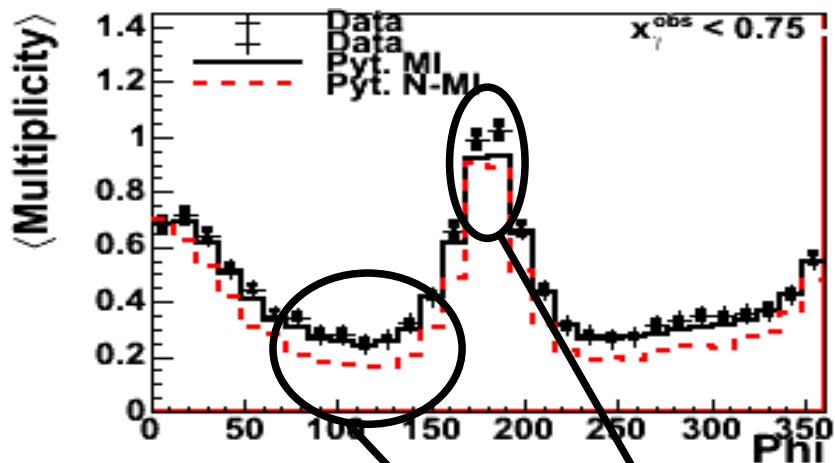
Particle multiplicity: Average number of measured particles per event.



Pythia MI describes all regions!

Dijet sample track multiplicity

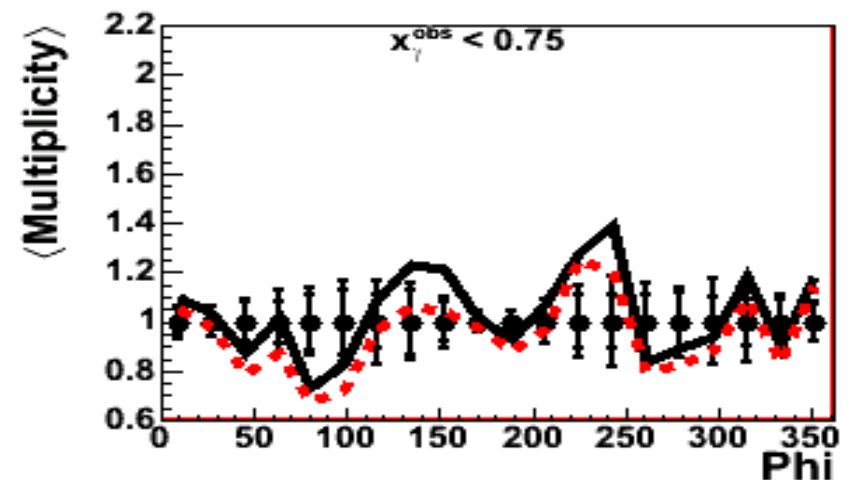
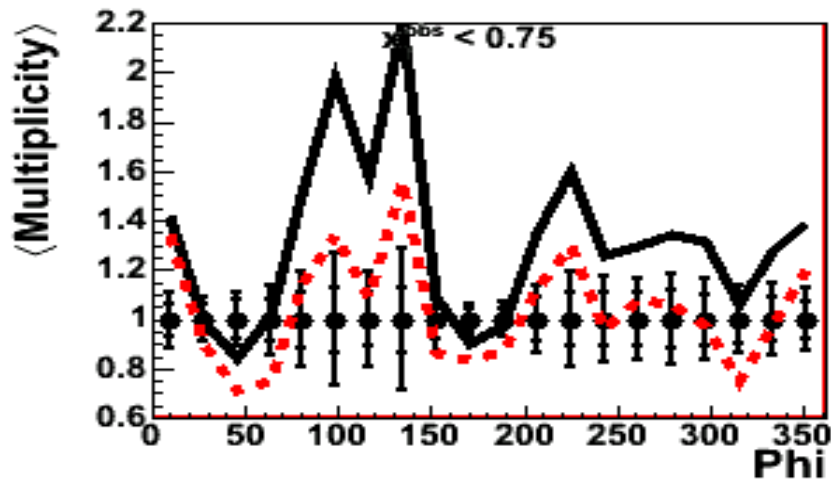
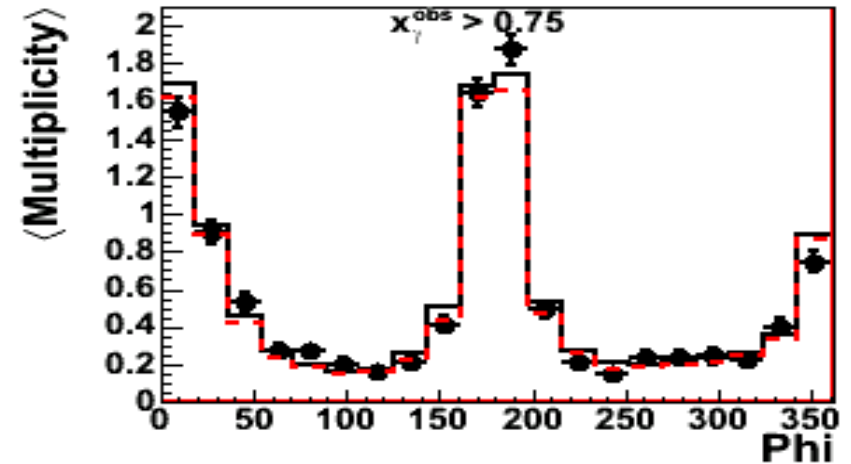
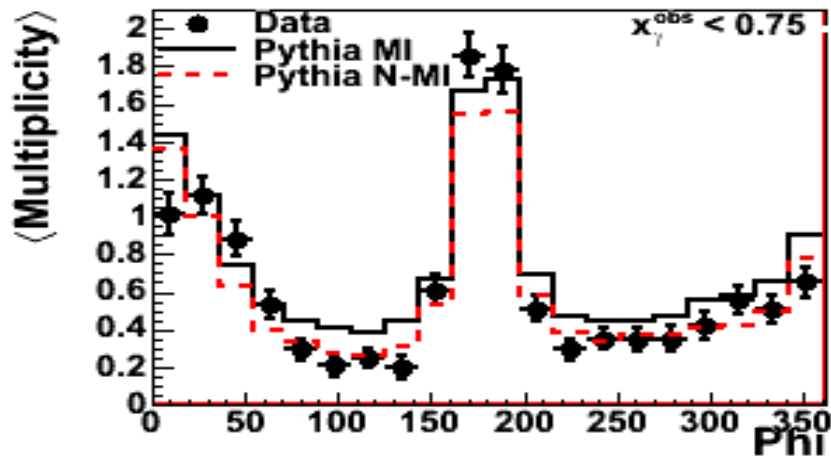
Particle multiplicity: Average number of measured particles per event.



Our naïve expectation is not fulfilled.
It is not just counting pears and apples!

Charm with dijets track multiplicity

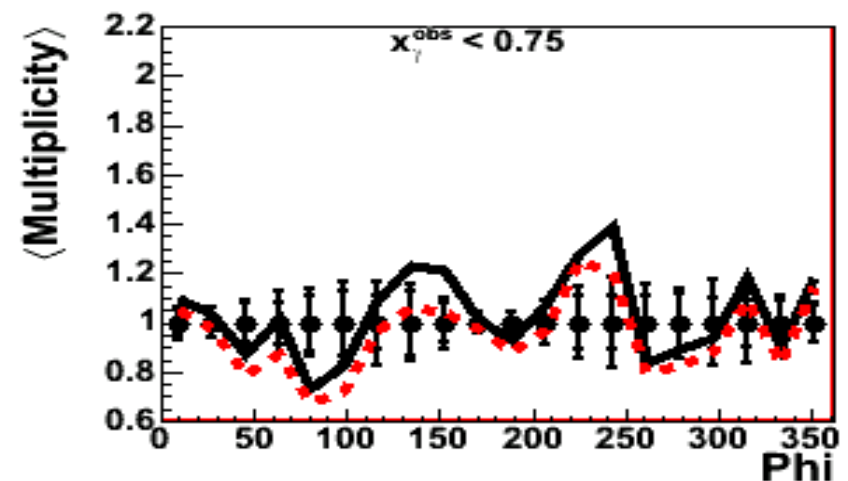
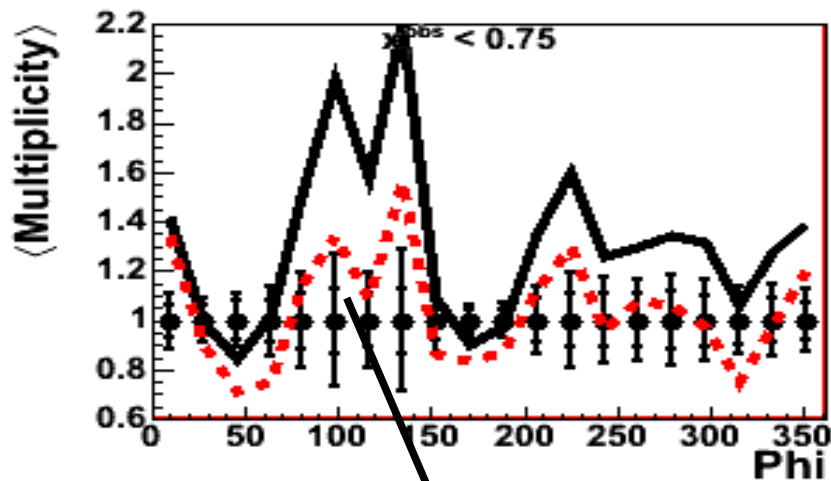
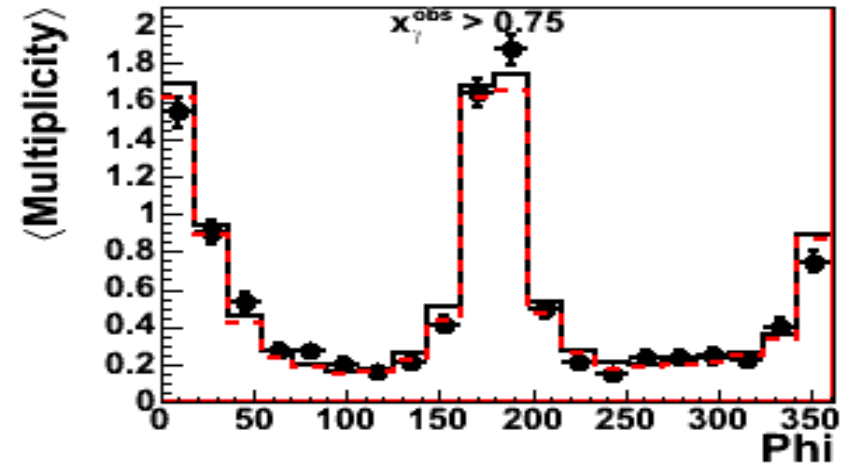
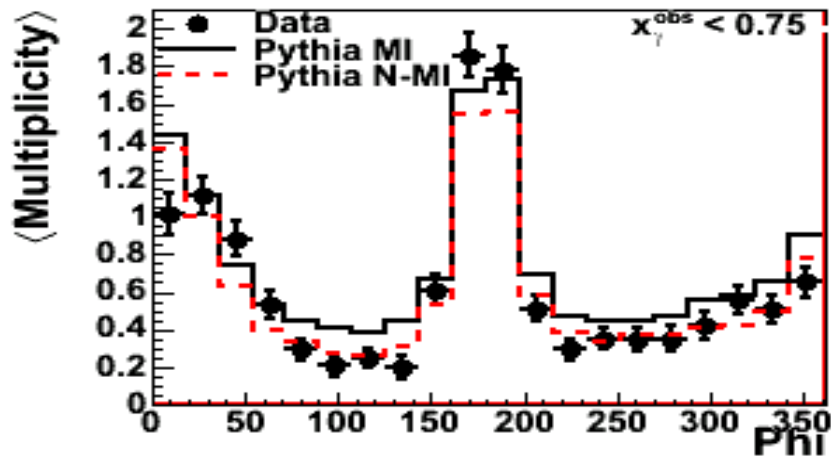
Particle multiplicity: Average number of measured particles per event.



Now Pythia MI does not describe data in the low X_γ bin!

Charm with dijets track multiplicity

Particle multiplicity: Average number of measured particles per event.



Statistical uncertainties are larger than before! ($\sim 15\%$ in transverse region)

Summary and outlook

Summary:

Dijet sample:

- ✓ PYTHIA MI describes multiplicity in both X_γ bins and all regions (especially the transverse).
- ✓ The difference in multiplicity between PYTHIA w/o MI and Data is $\sim 6\sigma$ in the transverse region.

Charm with dijet:

- ✓ Only the high X_γ region is well described.
- ✓ In the low X_γ the best description is done with Pythia w/o MI.

It is clear we still have a lot to understand about MI!

Backup

Pythia MI Model.

Parameter values used in Pythia:

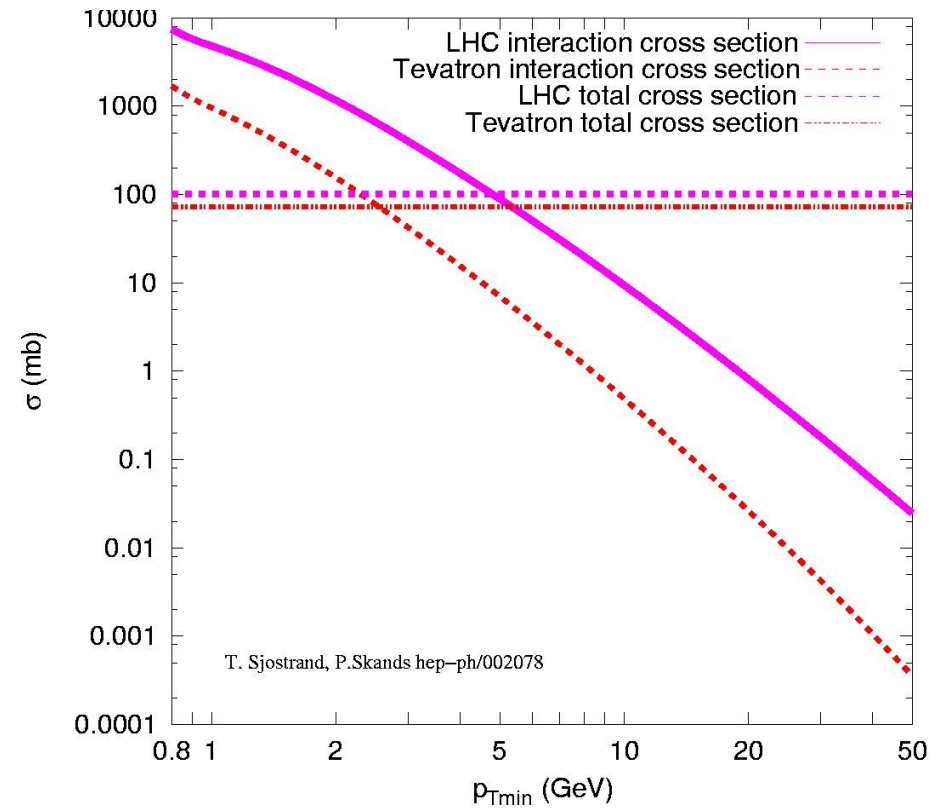
- ✗ PARP 67: set to 4 (default 1). Scale factor that governs the amount of initial-state radiation
- ✗ MSTP 82: set to 1. Same MI probability in all events. Abrupt P_{tmin} cut-off.
- ✗ PARP 82: set to 1.20GeV (default 1.55GeV). Regularization scale. Cut-off for MI.
- ✗ MSTP 93: set to 5 (default 1). Primordial K_t distribution in photon $dK_t^2 / (K_{t0}^2 + K_t^2)$
- ✗ PARP 99: set to 0.6GeV (default 0.4GeV). K_{t0}^2 value in MSTP93
- ✗ PARP 100: set to 5GeV (default 2GeV). Upper cut-off for primordial K_t distribution.

Pythia MI Model.

Some observations and hypothesis....

Fact: The $\sigma_{int}(p_t)$ diverges and is larger than the measured σ_{nd} for $p_{\perp} \rightarrow 0$.

A numerical illustration:



hep-ph/0402078

Pythia MI Model.

Some observations and hypothesis....

Fact: The $\sigma_{int}(p_t)$ diverges and is larger than the measured σ_{nd} for $p_{\perp} \rightarrow 0$.

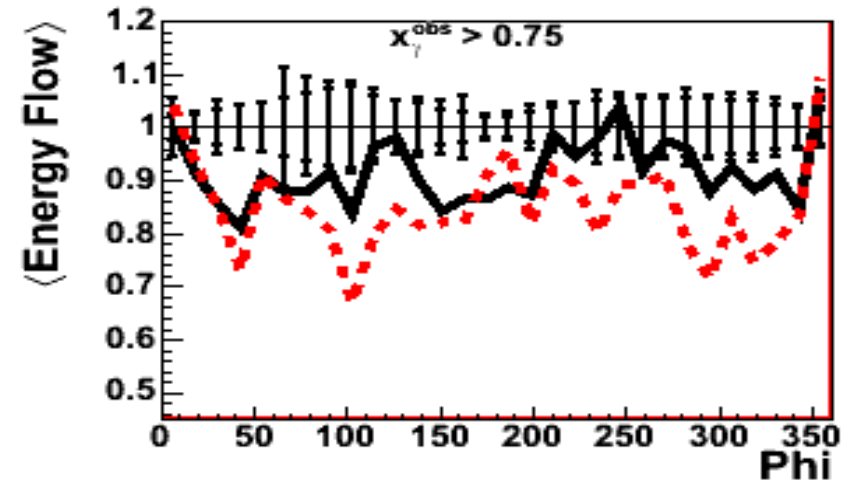
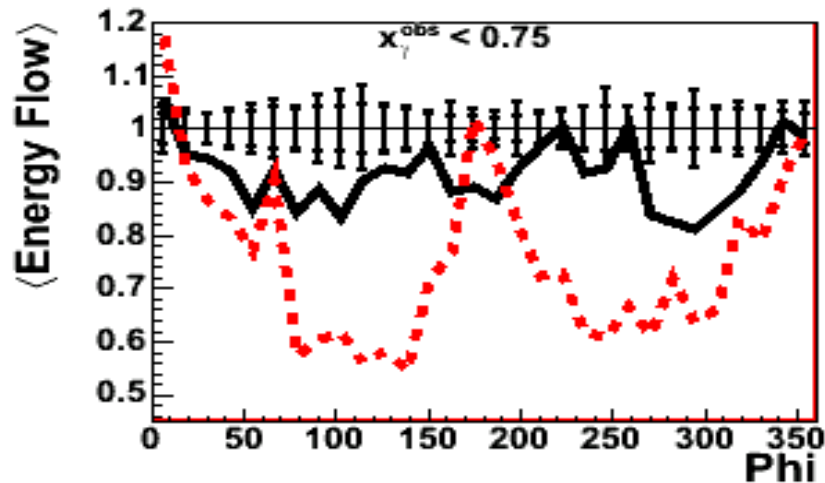
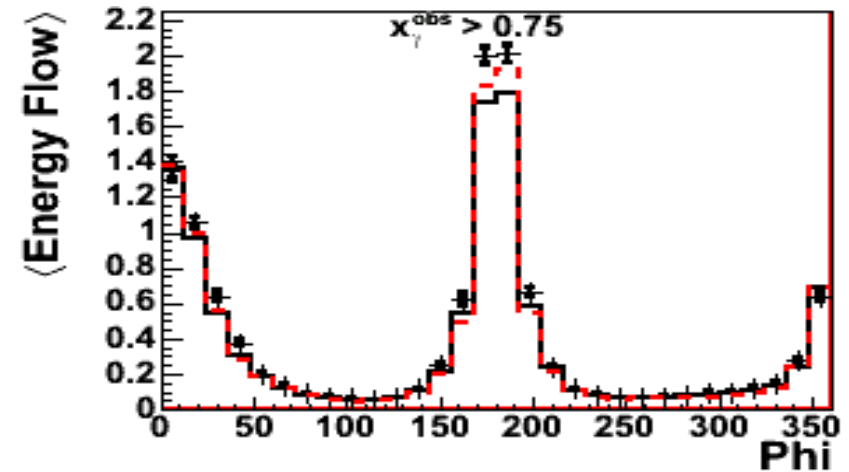
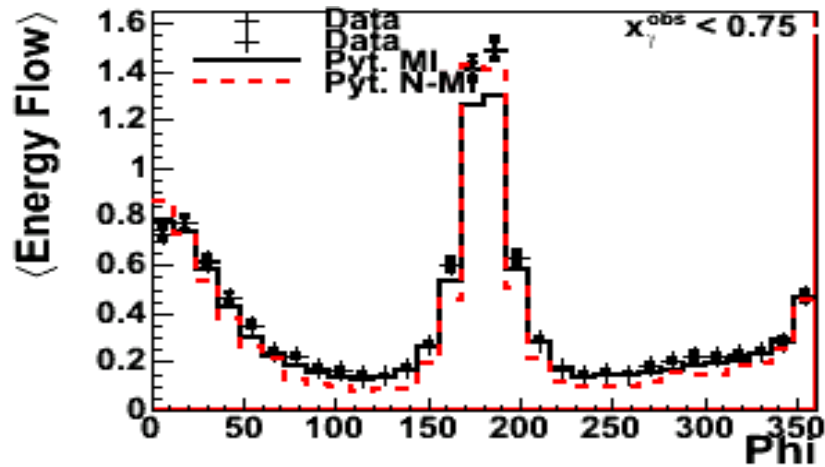
We interpret this as MPI: if we have two interactions this is counted twice in $\sigma_{int}(p_t)$ but only once in σ_{nd} .

Thus:

$$\langle n \rangle_{p_{\perp min}} = \frac{\sigma_{int}(p_{\perp min})}{\sigma_{nd}}$$

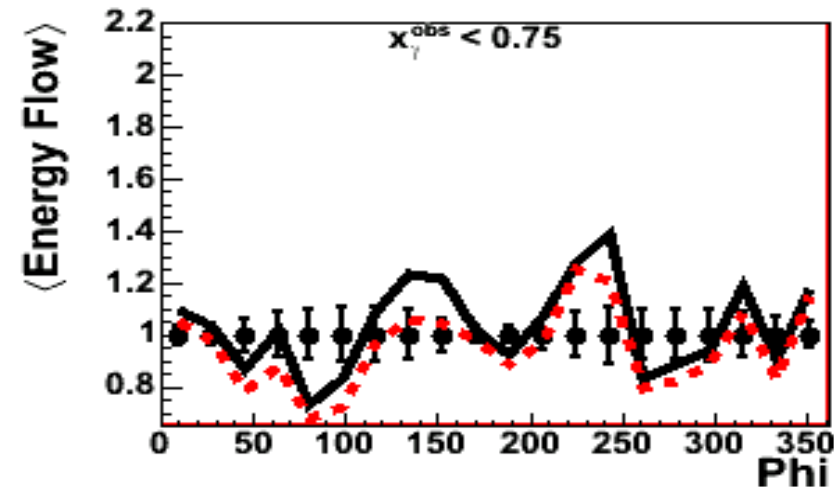
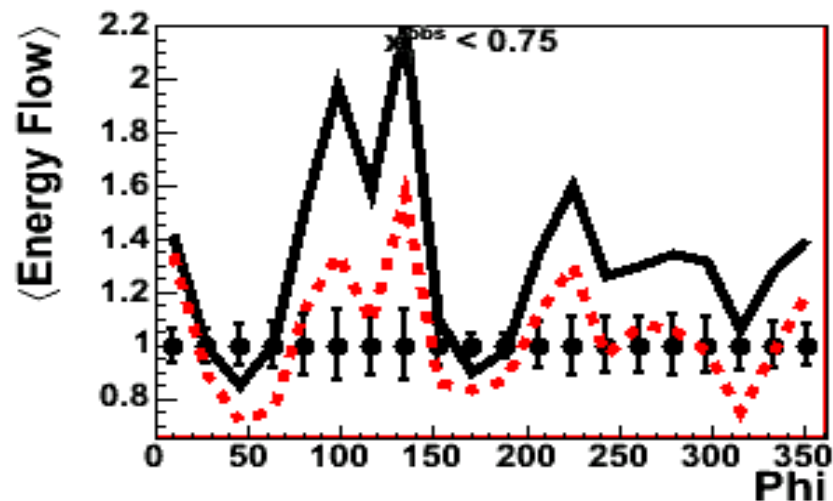
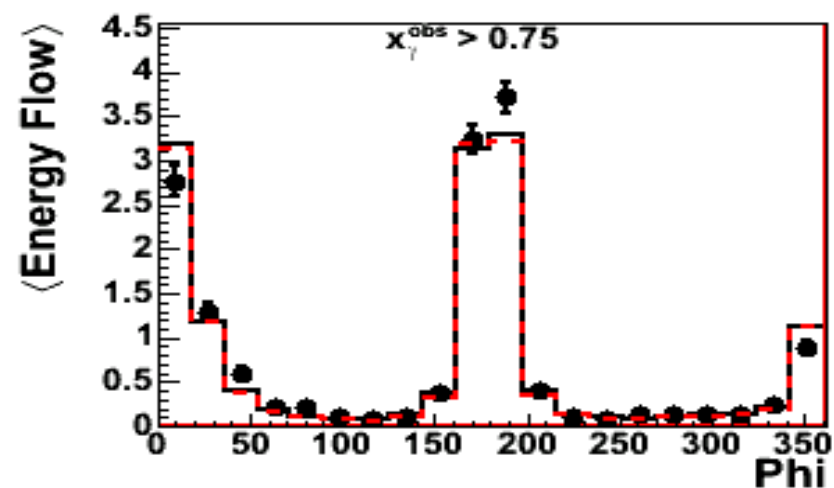
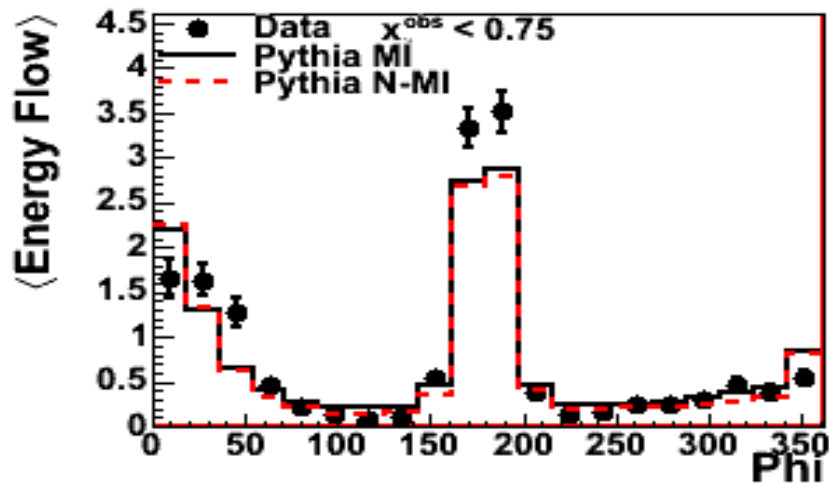
Dijet sample energy flow

Energy flow: average scalar Pt sum per event as a function of φ .

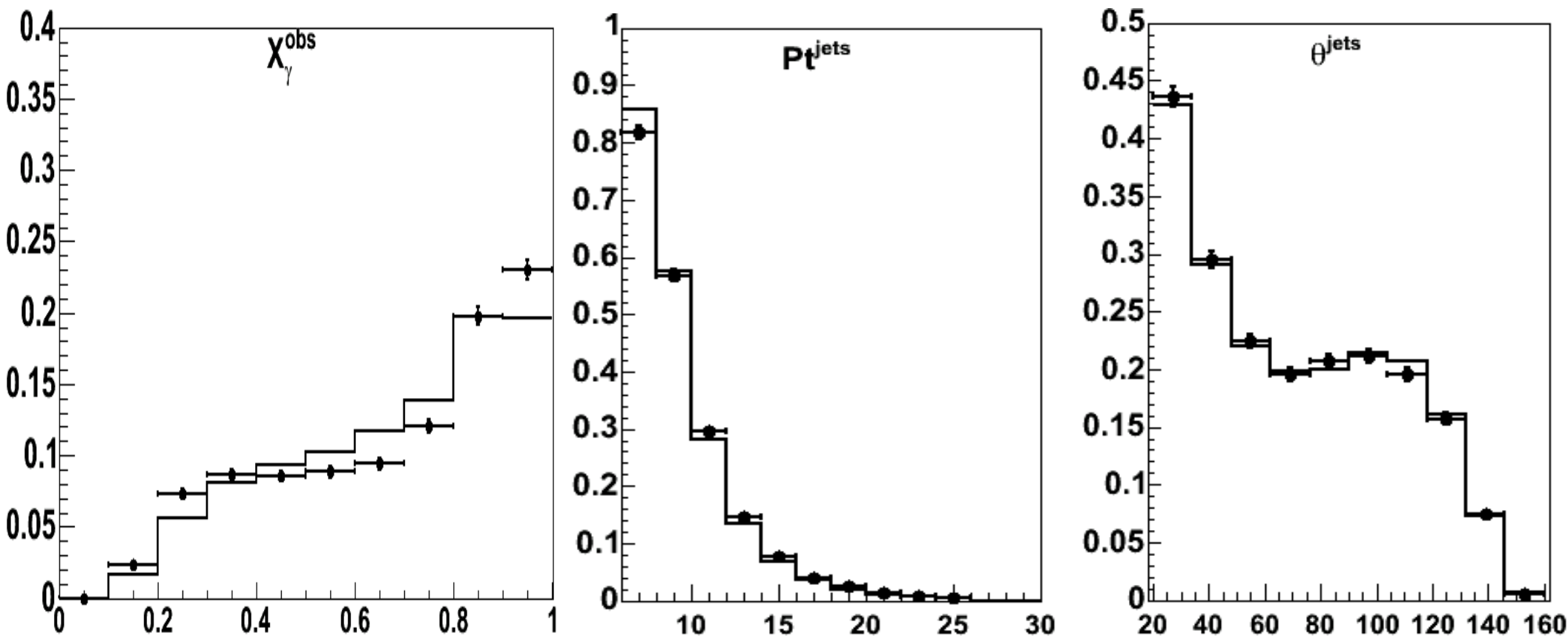


Charm with dijet energy flow

Energy flow: average scalar Pt sum per event as a function of φ .



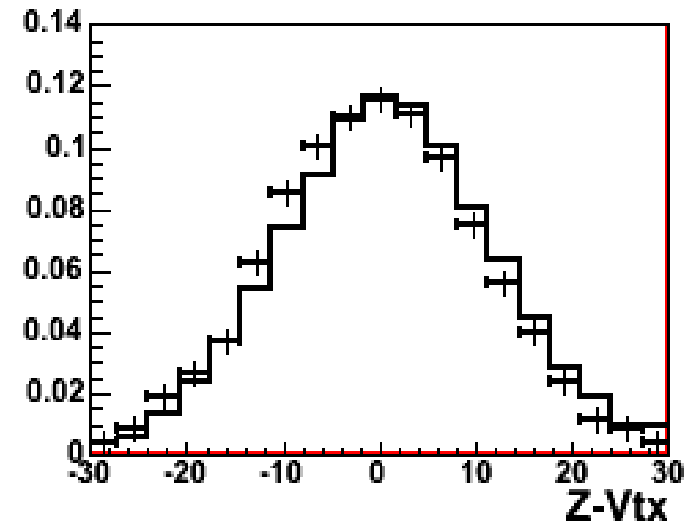
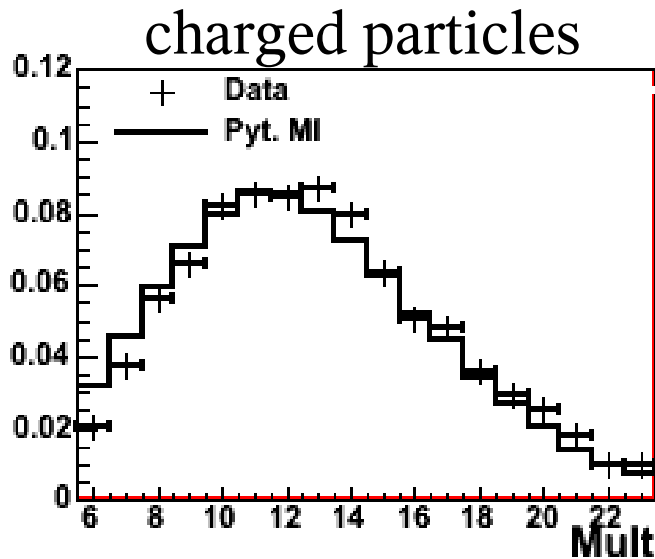
Dijet sample: some control plots



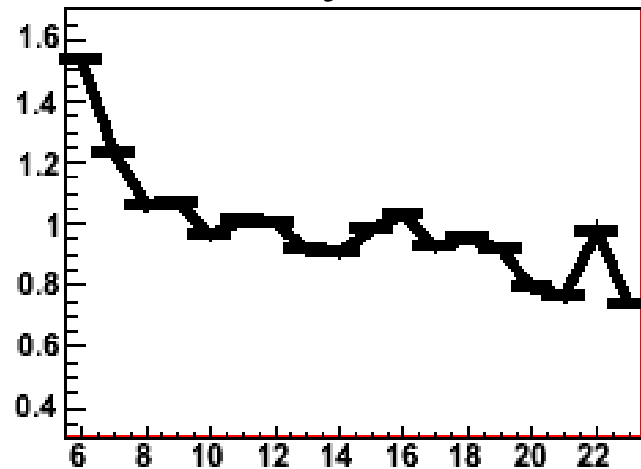
(shape normalized)

PYTHIA MI (def.) describes data reasonably well.

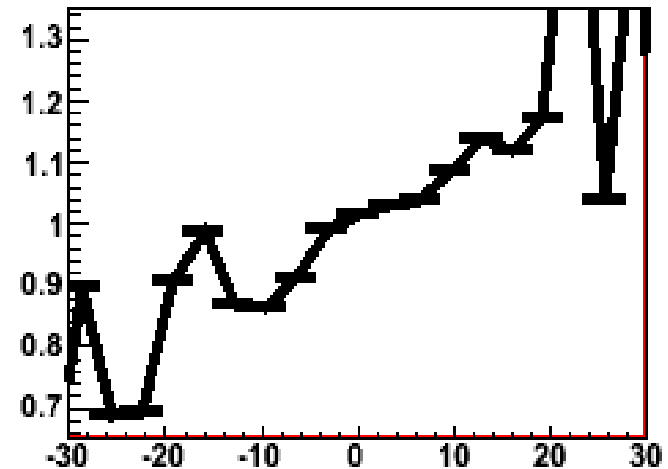
Dijet sample: some control plots



Data/Pythia MI

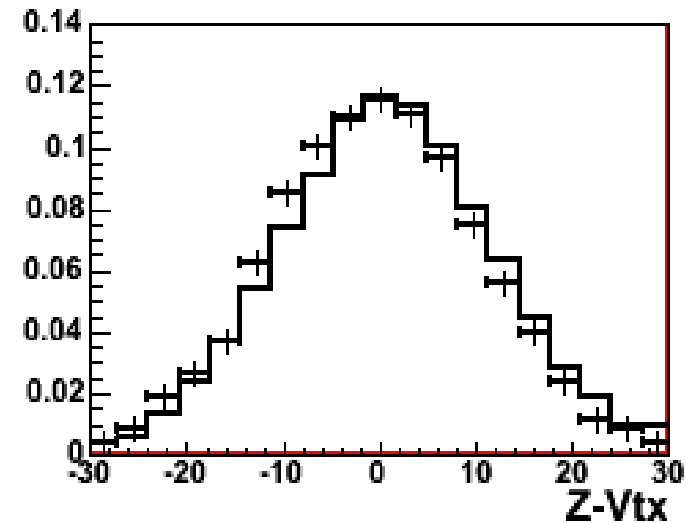
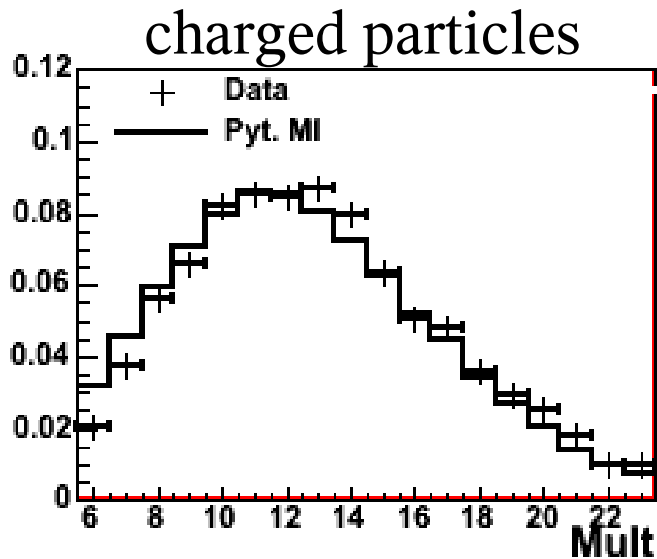


Data/Pythia MI

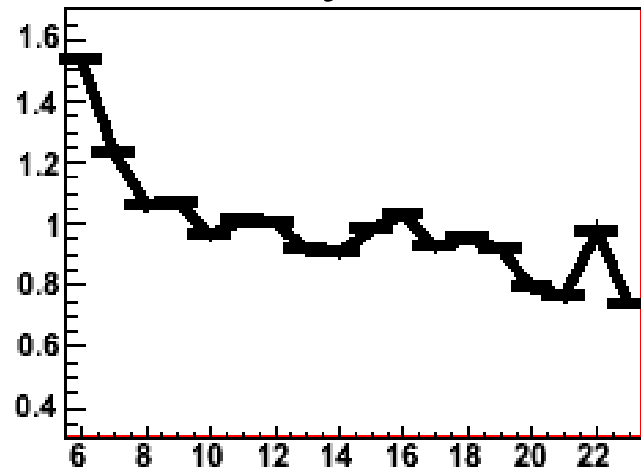


(shape normalized)

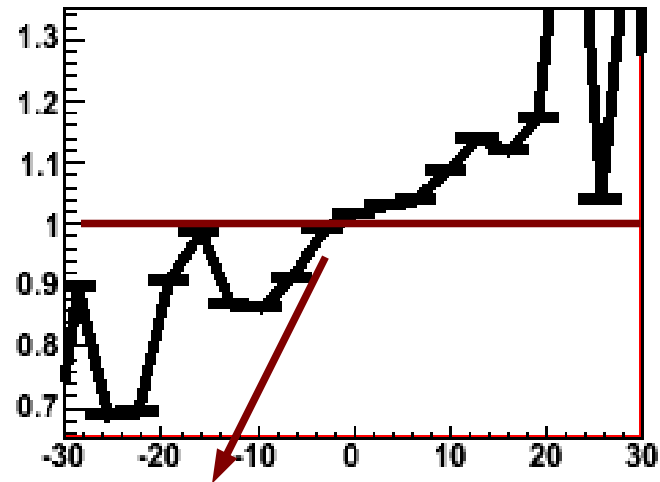
Dijet sample: some control plots



Data/Pythia MI



Data/Pythia MI



weight the events

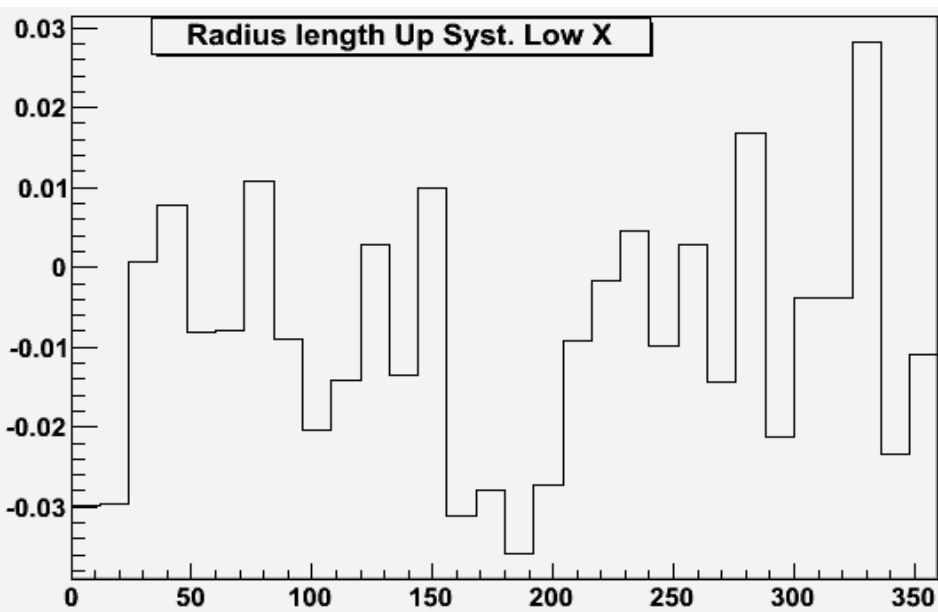
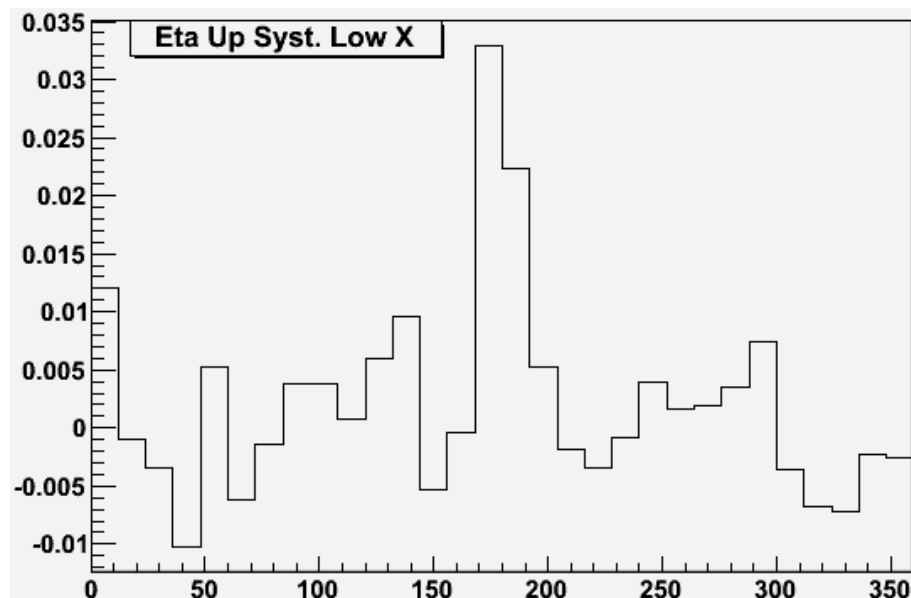
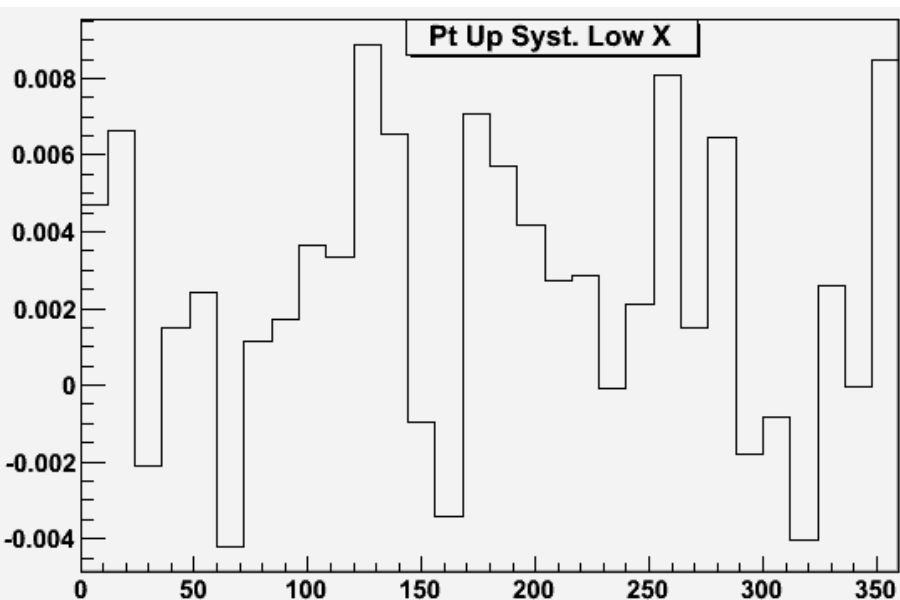
Systematics: calculation and plots

- ✓ We vary the nominal values to get the different contributions as follows:

$$\frac{Data}{MC^{sim/rec}} MC^{Gen} - \frac{Data_{p_t^\pm, \dots}}{MC_{p_t^\pm, \dots}^{sim/rec}} MC^{Gen}$$

Systematics: calculation and plots

- ✓ We vary the nominal values to get the different contributions as follows:



End of Backup