

How much (joint) Resummation do we need?

[with Gillian Lusermans &
Wouter Waalewijn]



UNIVERSITY OF AMSTERDAM

Andreas Papaefstathiou

[Parton Showers, Event Generators and Resummation (PSR19),
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Why go exclusive?

Precise understanding of Jets properties & substructure:

→ confirm SM @ higher accuracy \leftrightarrow identify deviations.

→ **multivariate** analyses,

→ **correlations.**

Why go exclusive?

Monte Carlos

→ “Fully Multi-differential” predictions of observables.



EXCLUSIVE



Analytic Resummation

→ Possible to systematically improve.



Why go exclusive?

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EXCLUSIVE



Difficult to systematically improve.

VS

Analytic Resummation

→ Possible to systematically improve.



Why go exclusive?

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EXCLUSIVE

X Difficult to systematically improve.

VS

Analytic Resummation

→ Possible to systematically improve.



X Difficult to make more differential.

Why go exclusive?

Monte Carlos

→ “Fully Multi-differential” predictions of observables.



EXCLUSIVE

X Difficult to systematically improve.

VS

Analytic Resummation

→ Possible to systematically improve.



EXCLUSIVE [?]

X Difficult to make more differential.

How exclusive is enough?

Analytic Resummation is **hard** to make more differential. [e.g. Procura, Waalewijn, Zeune, 1806.10622]

→ How much effort should we put into trying?

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2 \dots d\mathcal{O}_N} \Bigg|_{\text{Resummed}} \rightarrow N_{\text{optimal}} = ?$$

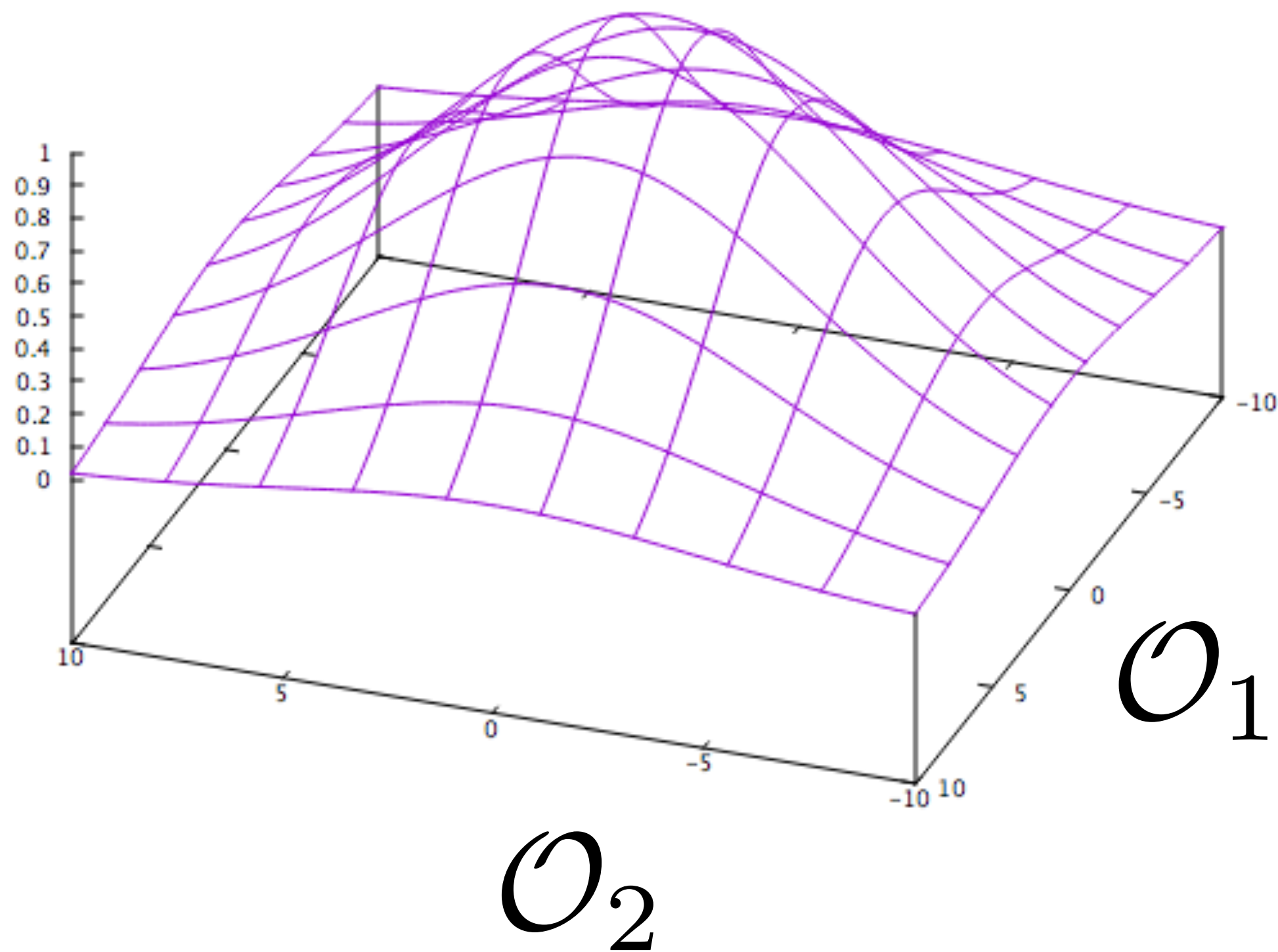
How exclusive is enough?

Answer in case study:

1. Start: “Flat” phase space MC: N observables \mathcal{O} . [**PhSp**]
2. Reweigh by calculation for N observables \mathcal{O} [**MC** or Resummation].
3. Look at “other” (not Reweighed) observables ,
4. Compare Reweighed [**RW**] prediction \leftrightarrow to **MC** or Resummation.
- [5. Profit!]

Reweighting

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$



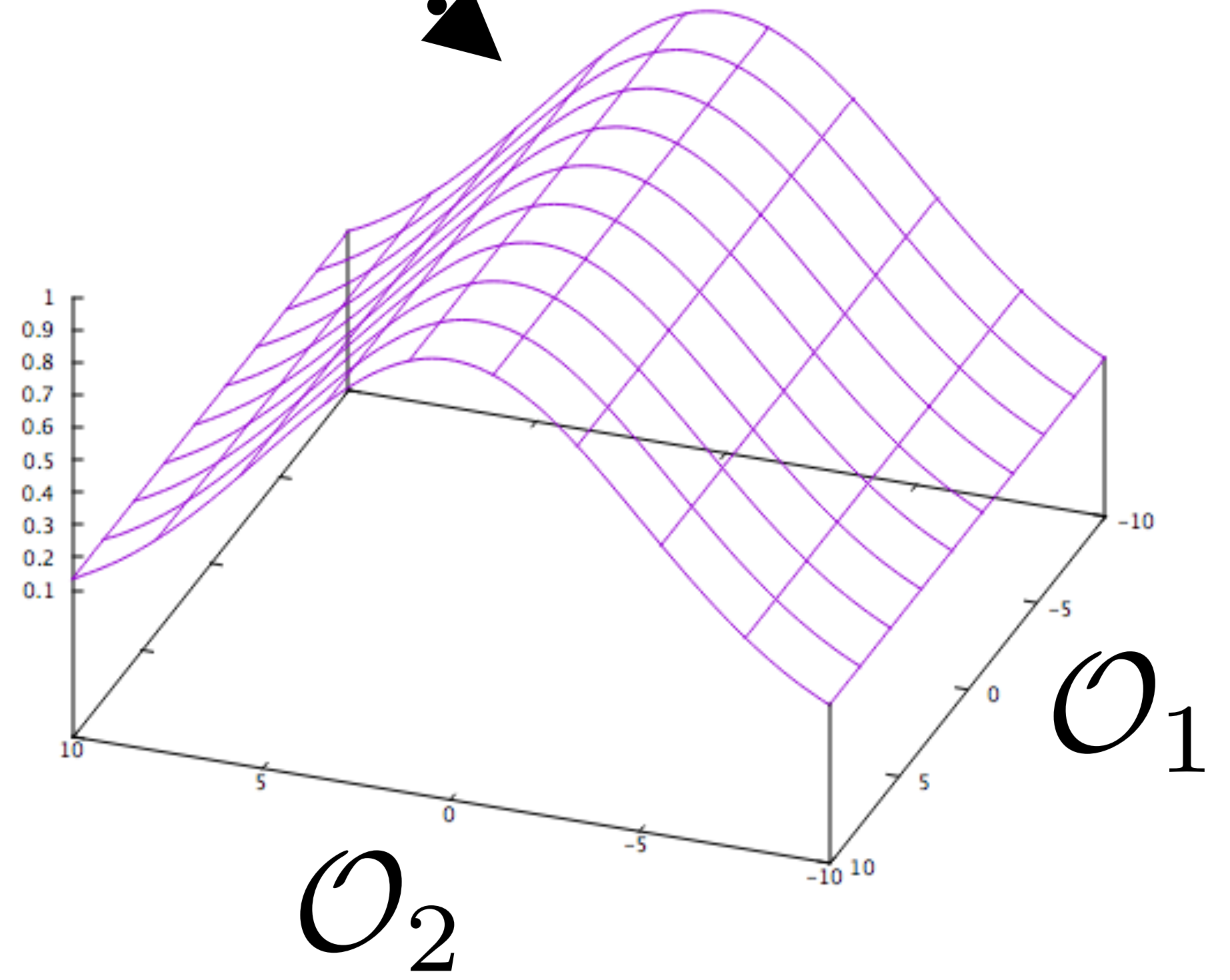
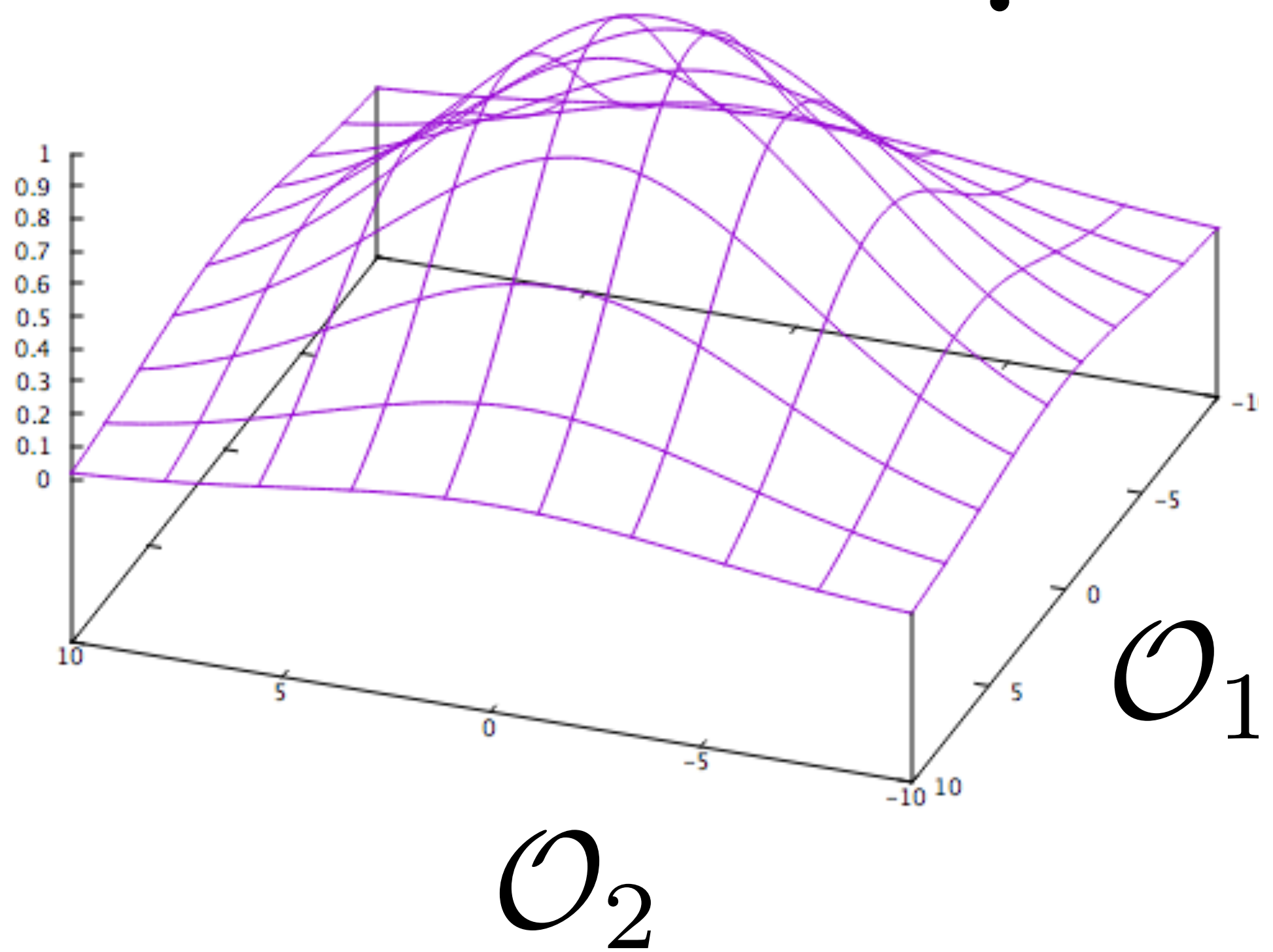
Reweighting

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

Make "Flat" in
dimension \mathcal{O}_1 :

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

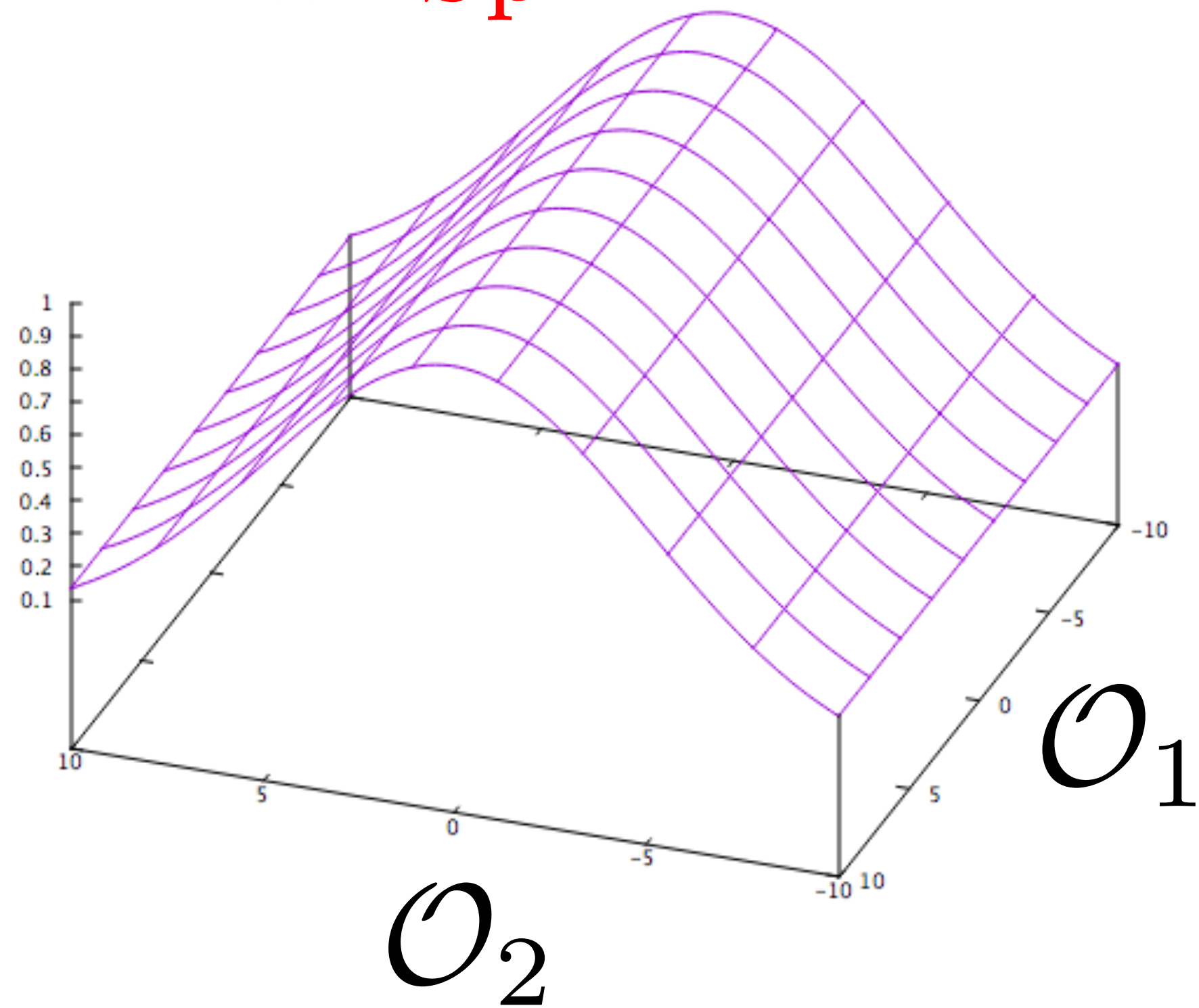
$$\frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{PhSp}}$$



Reweighting

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

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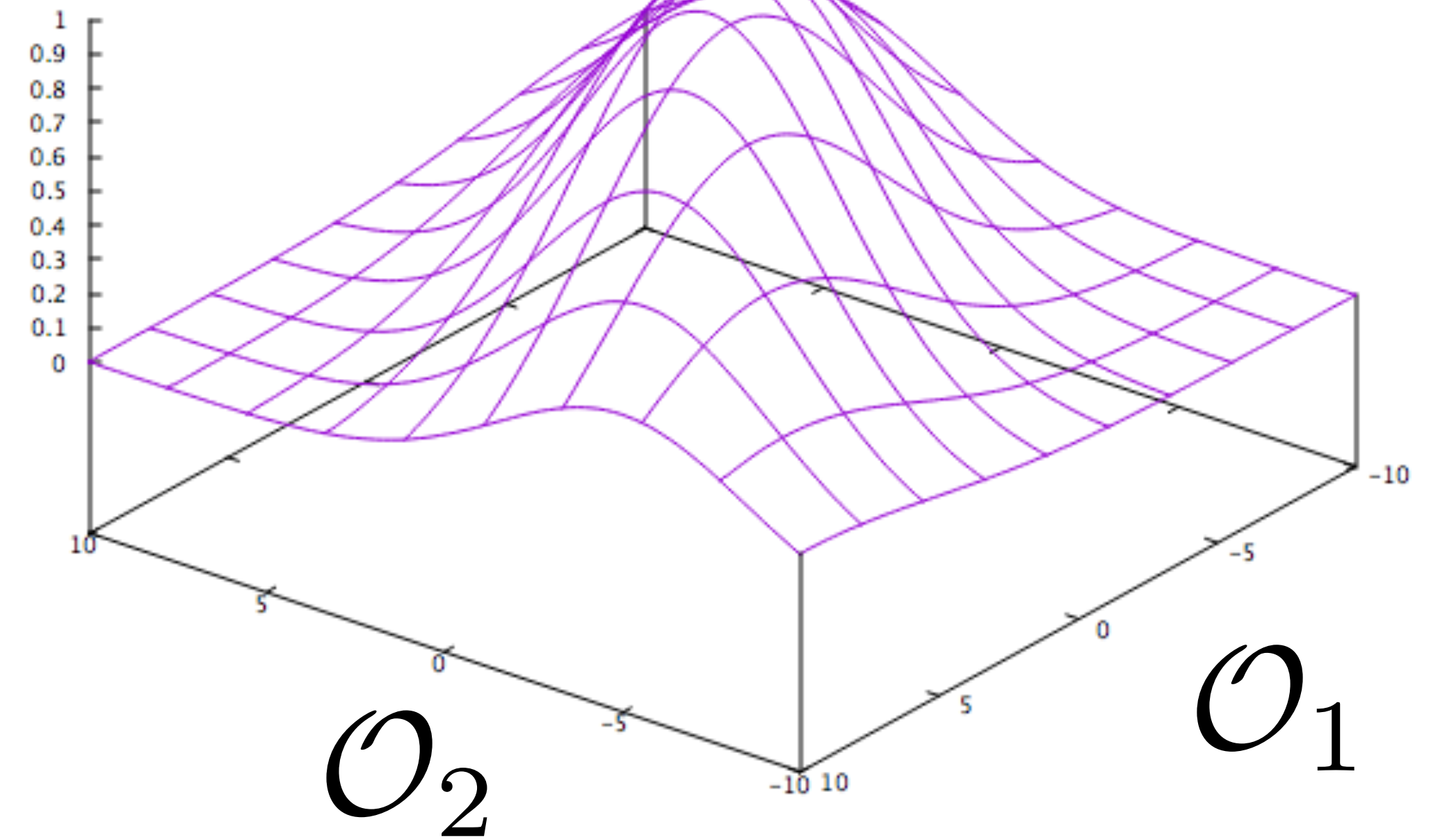
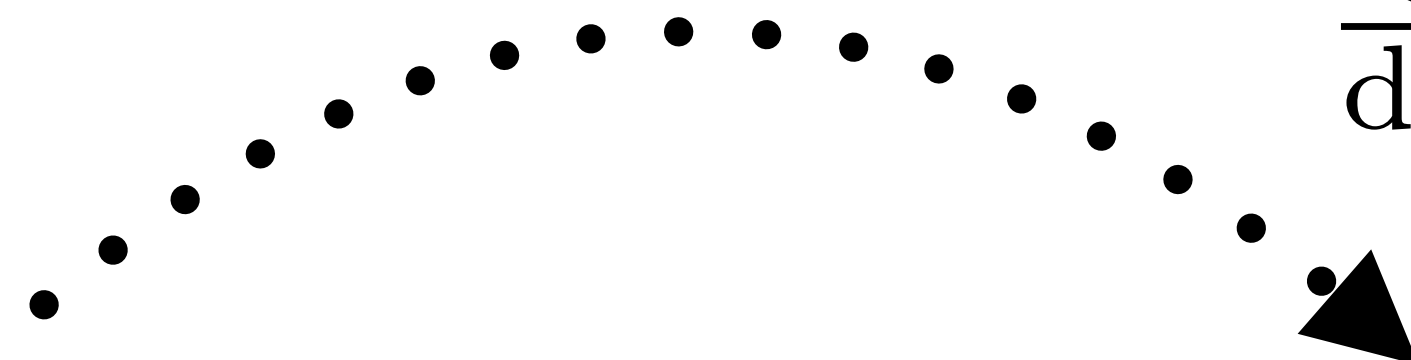
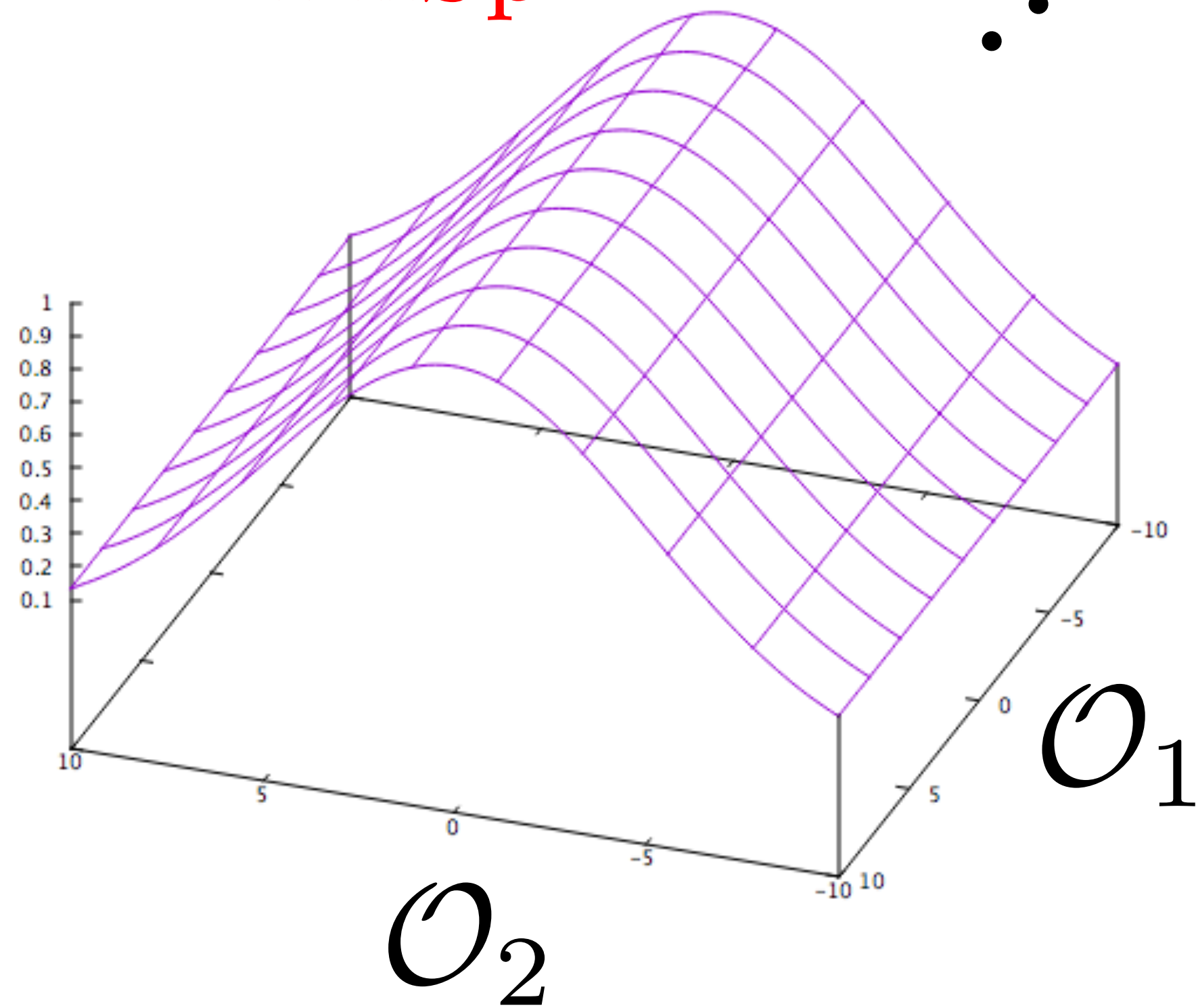


Reweighting

$$\frac{\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}}{\frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{PhSp}}}$$

Reweigh by
MC \mathcal{O}_1 distribution

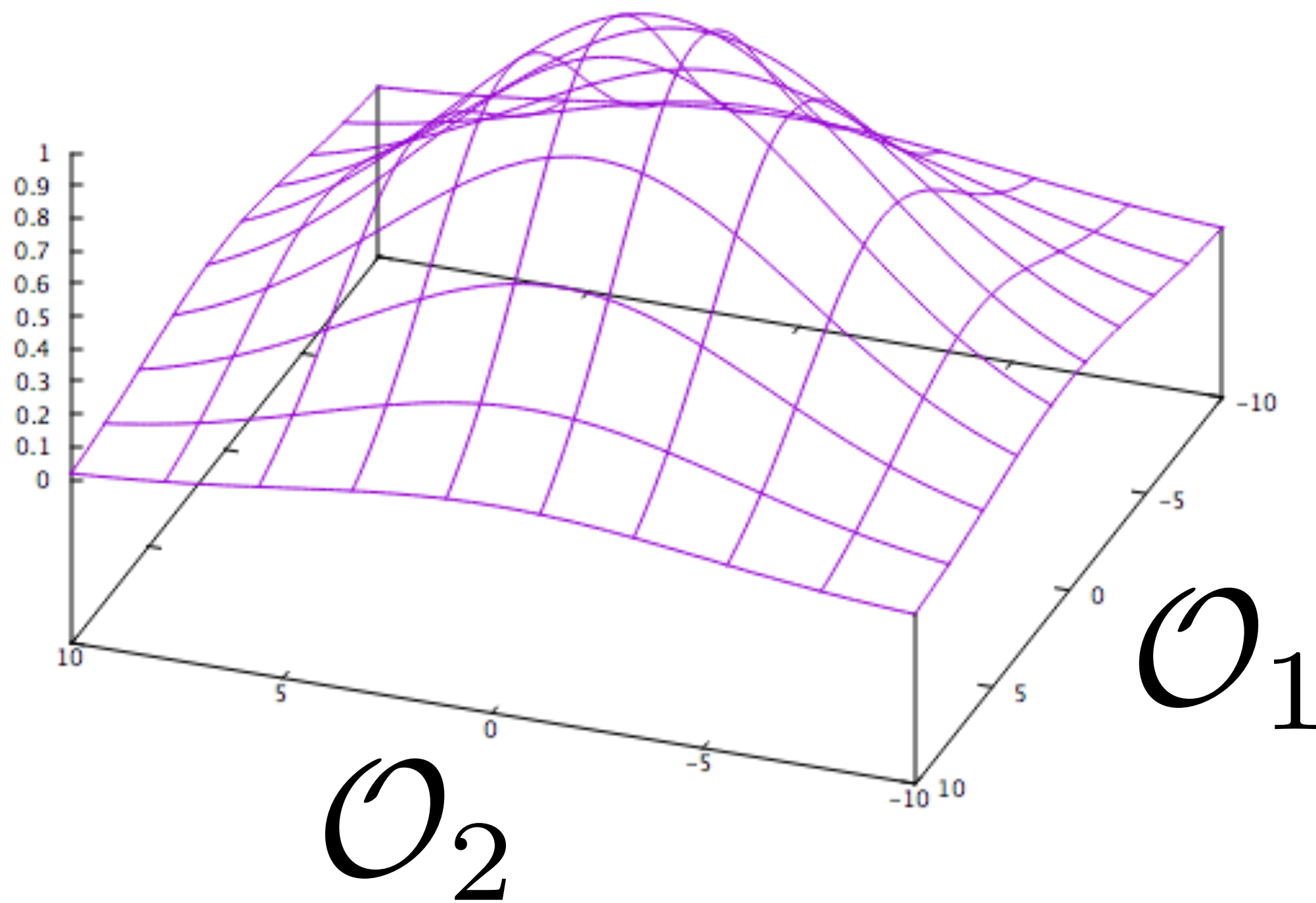
$$\frac{\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}}{\frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{PhSp}}} \times \frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{MC}}$$



WARNING

Drawn for illustrative purposes!

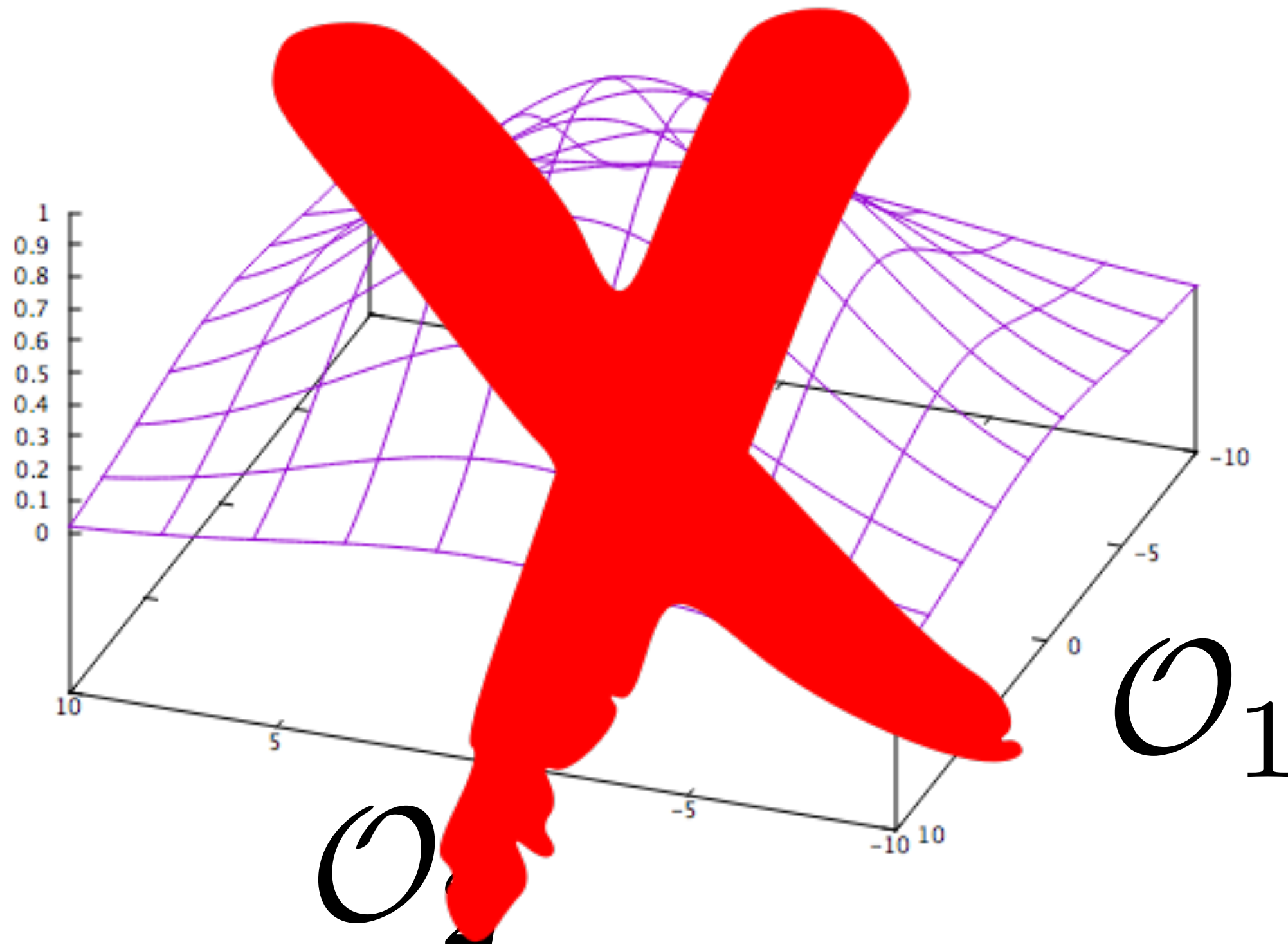
In general, \mathcal{O}_1 - \mathcal{O}_2 are correlated!



WARNING

Drawn for illustrative purposes!

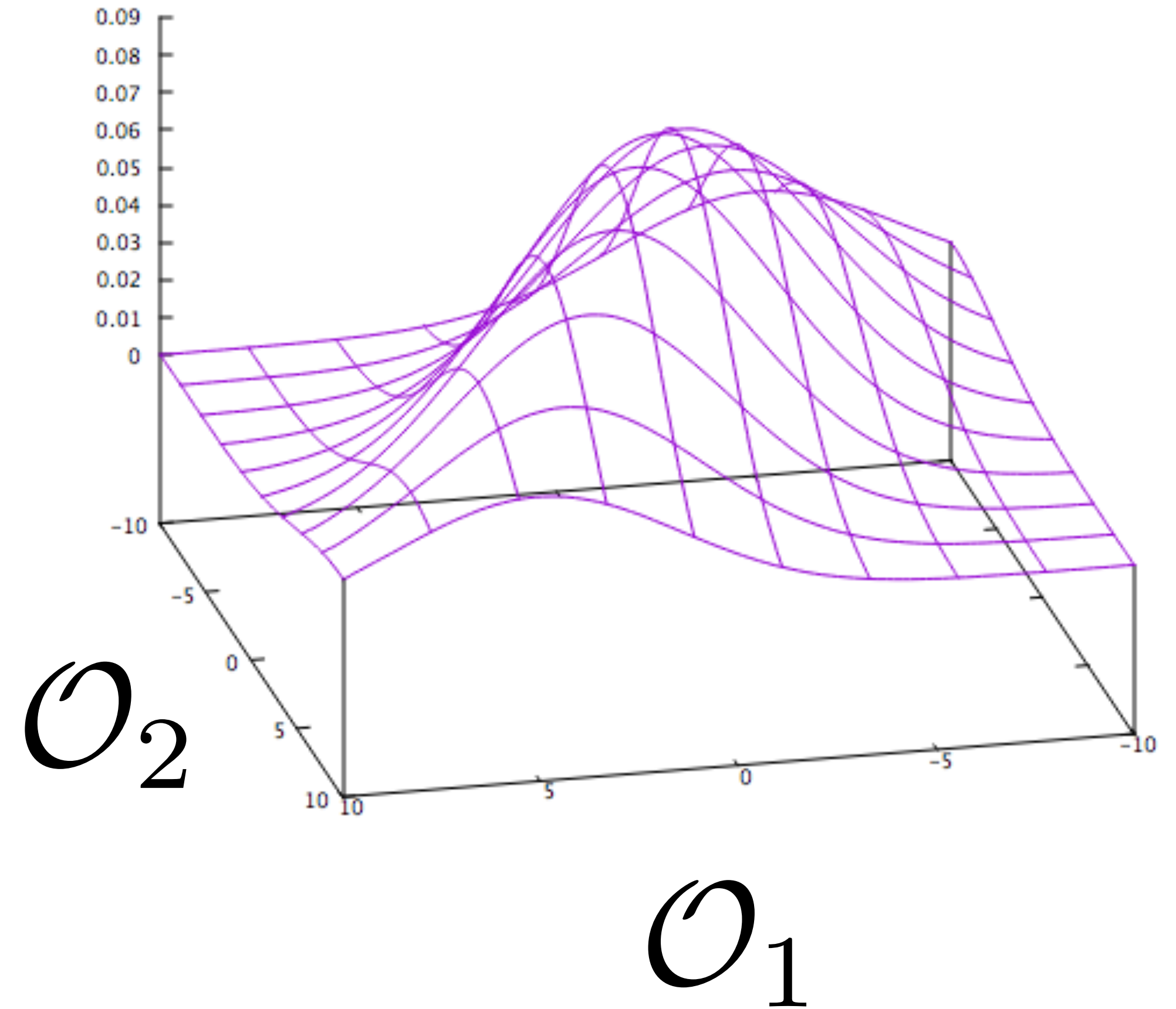
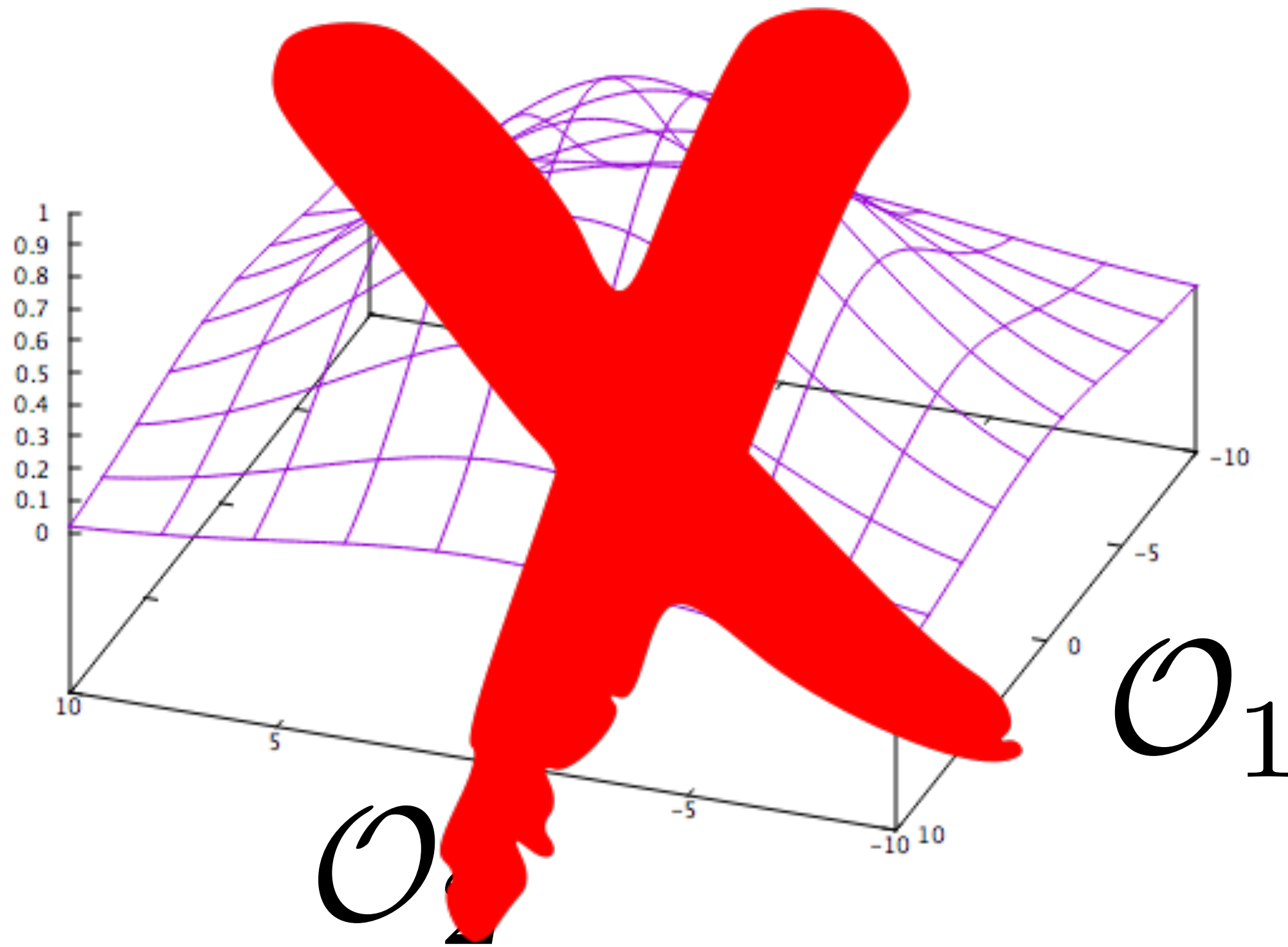
In general, \mathcal{O}_1 - \mathcal{O}_2 are correlated!



WARNING

Drawn for illustrative purposes!

In general, \mathcal{O}_1 - \mathcal{O}_2 are correlated!



Projecting down

$$\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{RW} = \int d\mathcal{O}_1 \frac{\left. \frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \right|_{PhSp}}{\left. \frac{d\sigma}{d\mathcal{O}_1} \right|_{PhSp}} \times \left. \frac{d\sigma}{d\mathcal{O}_1} \right|_{MC}$$

Compare “other”
observable:

$$\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{RW}$$



$$\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{MC}$$

“prediction”

“truth”

Projecting down

Compare “other”
observable:

$$\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{RW}$$



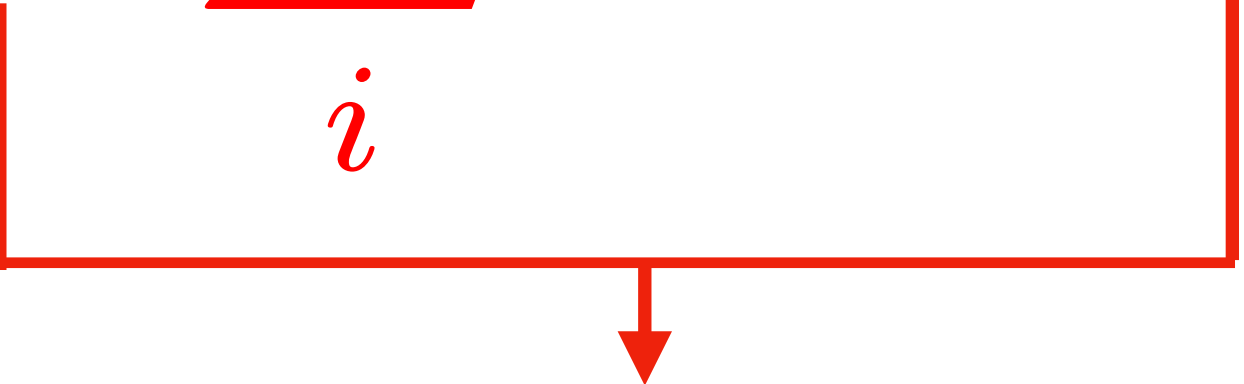
$$\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{MC}$$

via:

$$\chi^2(\mathcal{O}_2) = \sum_{\text{bin } i} \left[\left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{RW,i} - \left. \frac{d\sigma}{d\mathcal{O}_2} \right|_{MC,i} \right]^2$$

“goodness of fit”

Flat Phase Space Monte Carlo

$$d\phi_n(p_i|Q) = \delta\left(\sum_i p_i - Q\right) \prod_i d^4 p_i \delta(p_i^2) \theta(p_i^0)$$


Total momentum conservation

Flat Phase Space Monte Carlo

$$d\phi_n(p_i|Q) = \delta\left(\sum_i p_i - Q\right) \prod_i d^4 p_i \delta(p_i^2) \theta(p_i^0)$$



Total momentum conservation

→ Use Simon's "RAMBO on diet" [Thanks!].

[Plätzer, 1308.2922]

Flat Phase Space Monte Carlo

$$d\phi_n(p_i|Q) = \delta\left(\sum_i p_i - Q\right) \prod_i d^4 p_i \delta(p_i^2) \theta(p_i^0)$$



Total momentum conservation

→ Use Simon's "RAMBO on diet" [Thanks!]. [Plätzer, 1308.2922]

→ with a "twist": **importance sampling** for our observables .

The setup

Process: $e^+ e^- \rightarrow jj$ at fixed Q . [MCs: HERWIG 7, Pythia 8]

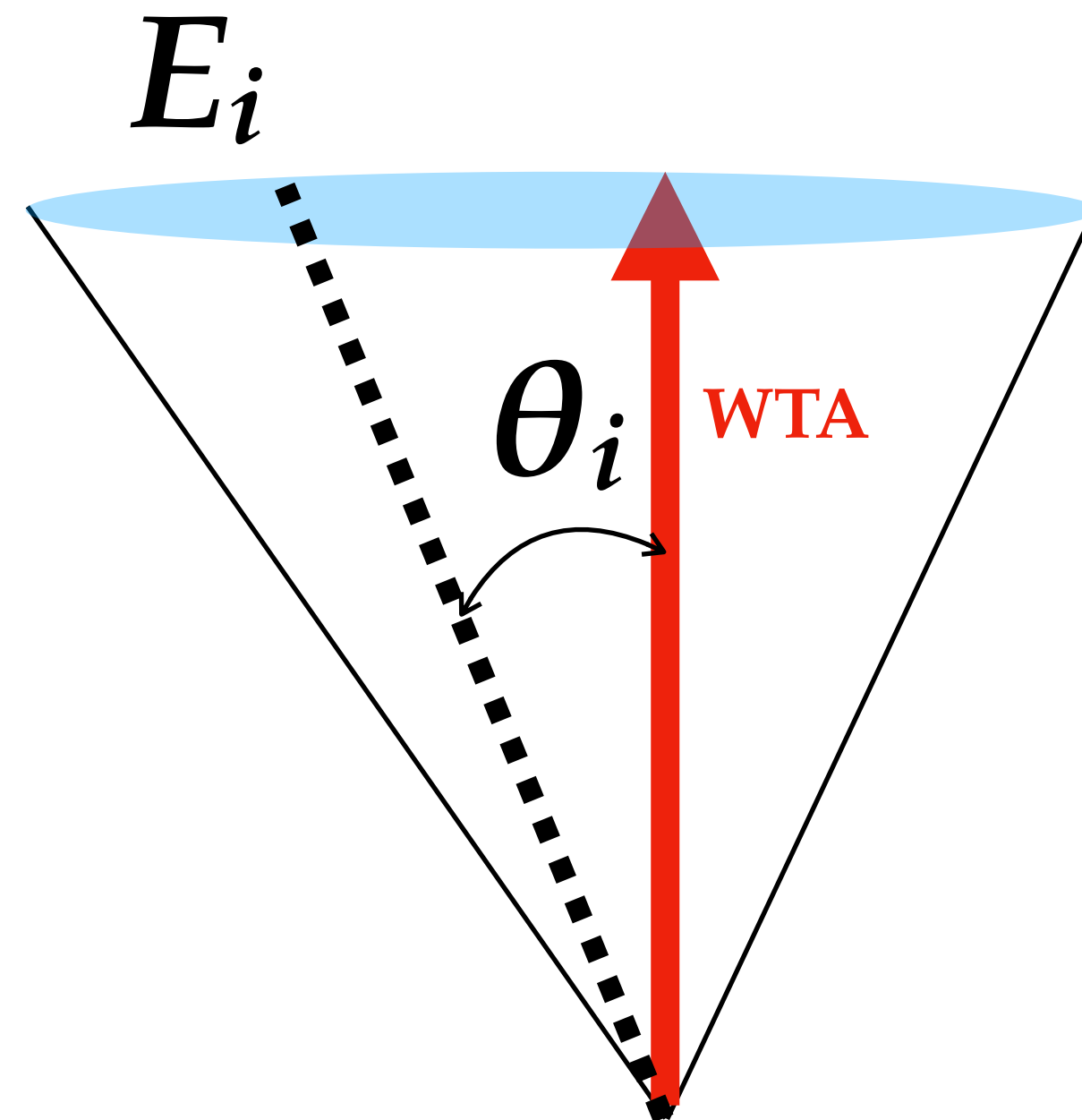
[no hadronisation / non-perturbative effects.]

Analysis: Exclusive k_T algorithm \rightarrow 2 jets,
“Winner-Takes-All” recombination:

$A + B \rightarrow (AB) : \hat{\vec{p}}_{AB} = \text{hardest of } (\hat{\vec{p}}_A, \hat{\vec{p}}_B) \quad [\rightarrow \text{Recoil-insensitive}]$

The observables

Jet Angularities: $e_\alpha \sim \frac{1}{Q} \sum_i E_i \theta_i^\alpha$ [over both jets]

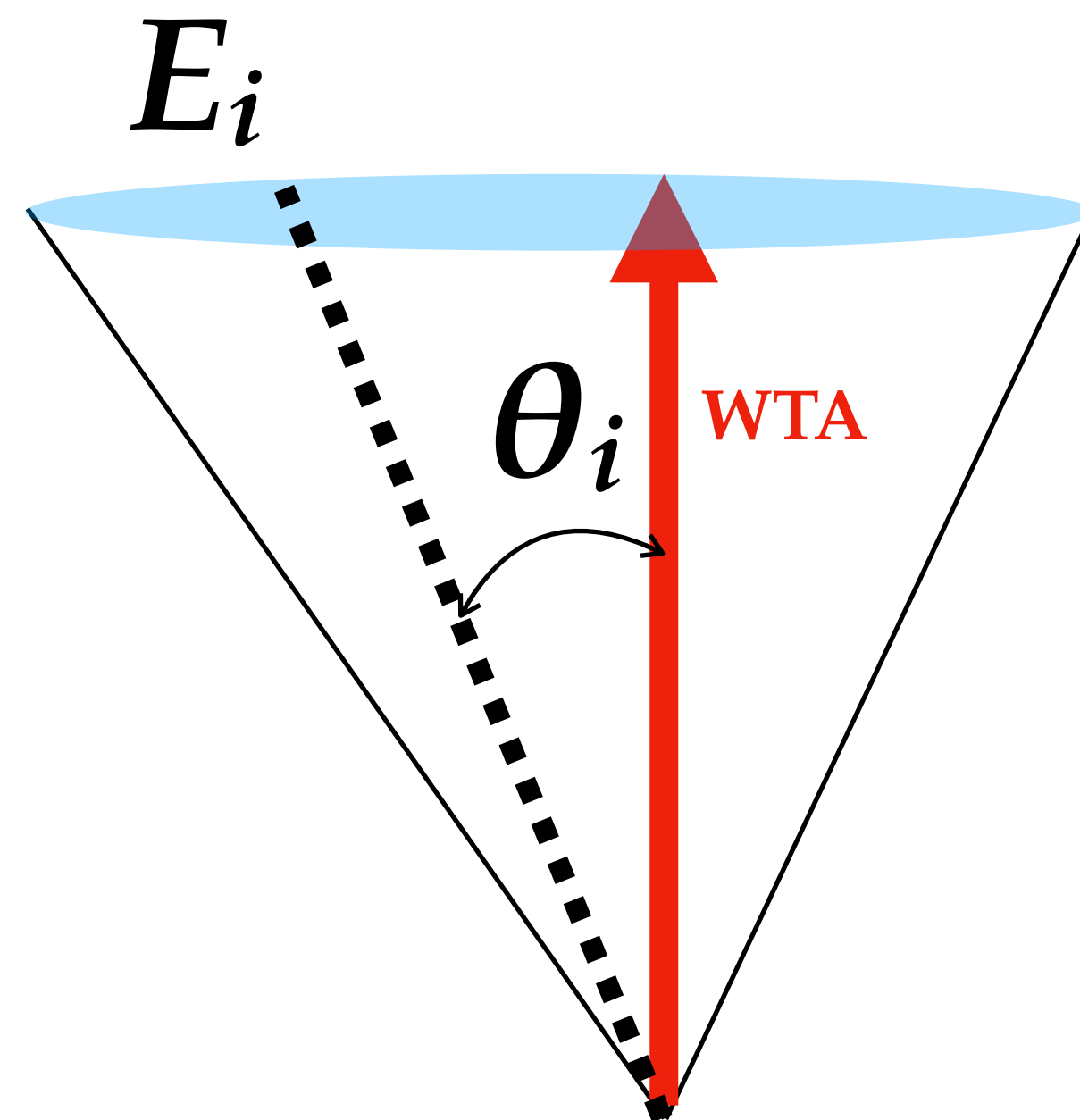


$\alpha = 2 \sim$ Thrust

$\alpha = 1 \sim$ Total broadening

The observables

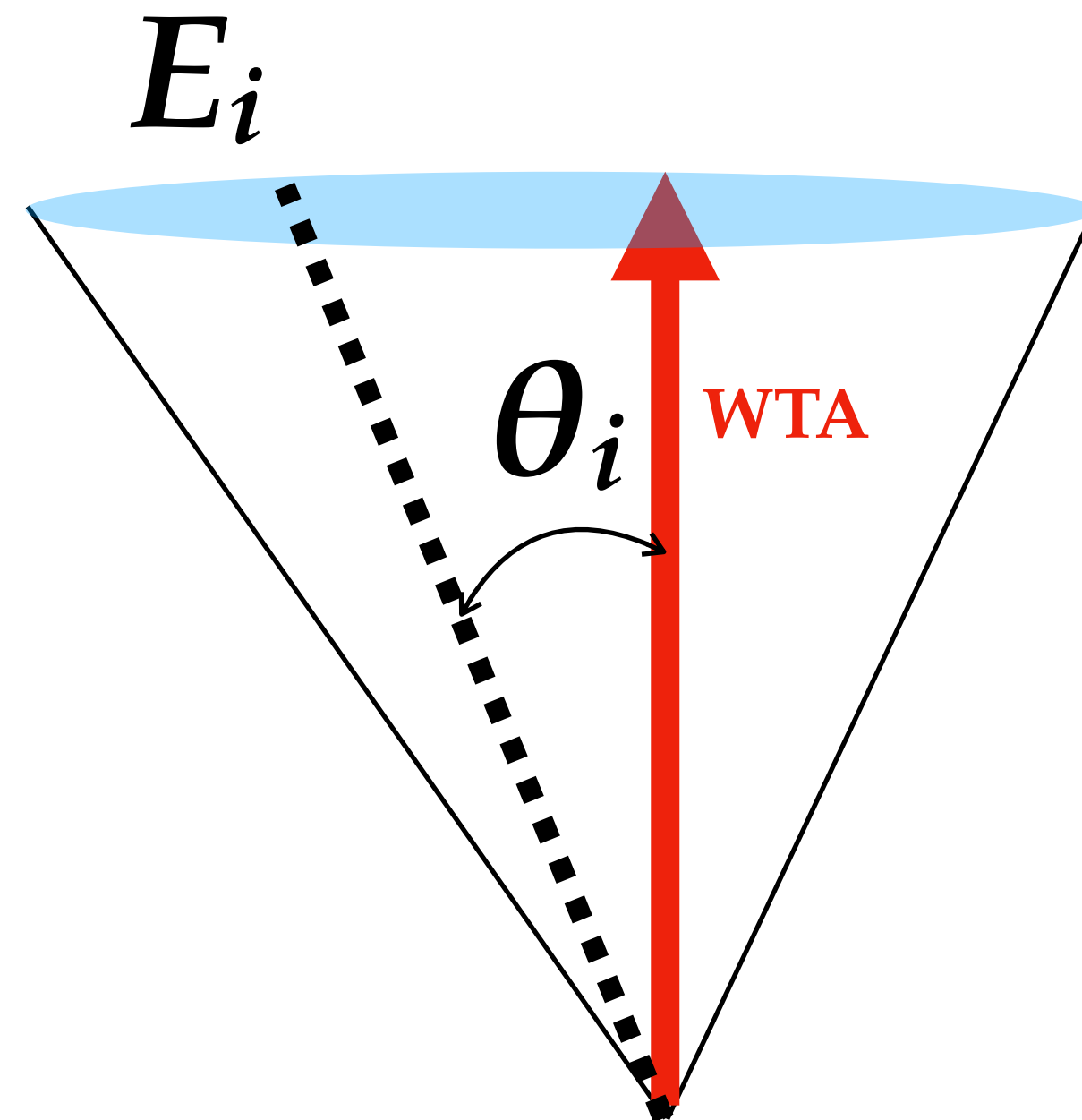
Jet Angularities: $e_\alpha = \frac{2}{Q} \sum_i E_i \left[\sin \left(\frac{\theta_i}{2} \right) \right]^\alpha$ [over both jets]



Actually look at

The observables

Jet Angularities: $e_\alpha = \frac{2}{Q} \sum_i E_i \left[\sin \left(\frac{\theta_i}{2} \right) \right]^\alpha$ [over both jets]



Actually look at

$$\log_{10} e_\alpha$$

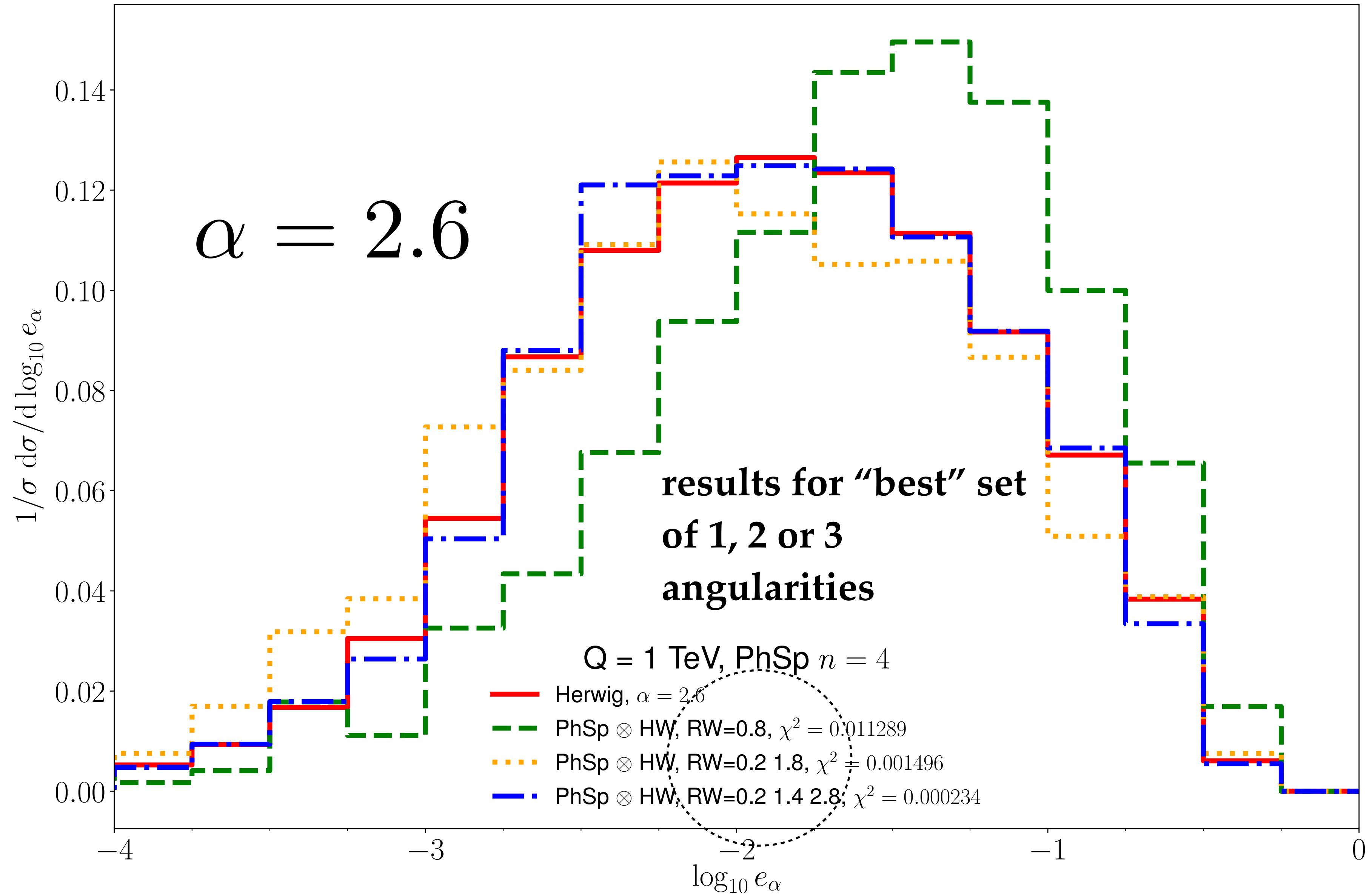
The “best” Angularity set

→ Reweigh by: [1, 2, 3, 4, 5] angularities. $\left\{ \begin{array}{l} \alpha \in [0.2, 3.0] \\ \text{in steps of 0.2} \end{array} \right.$

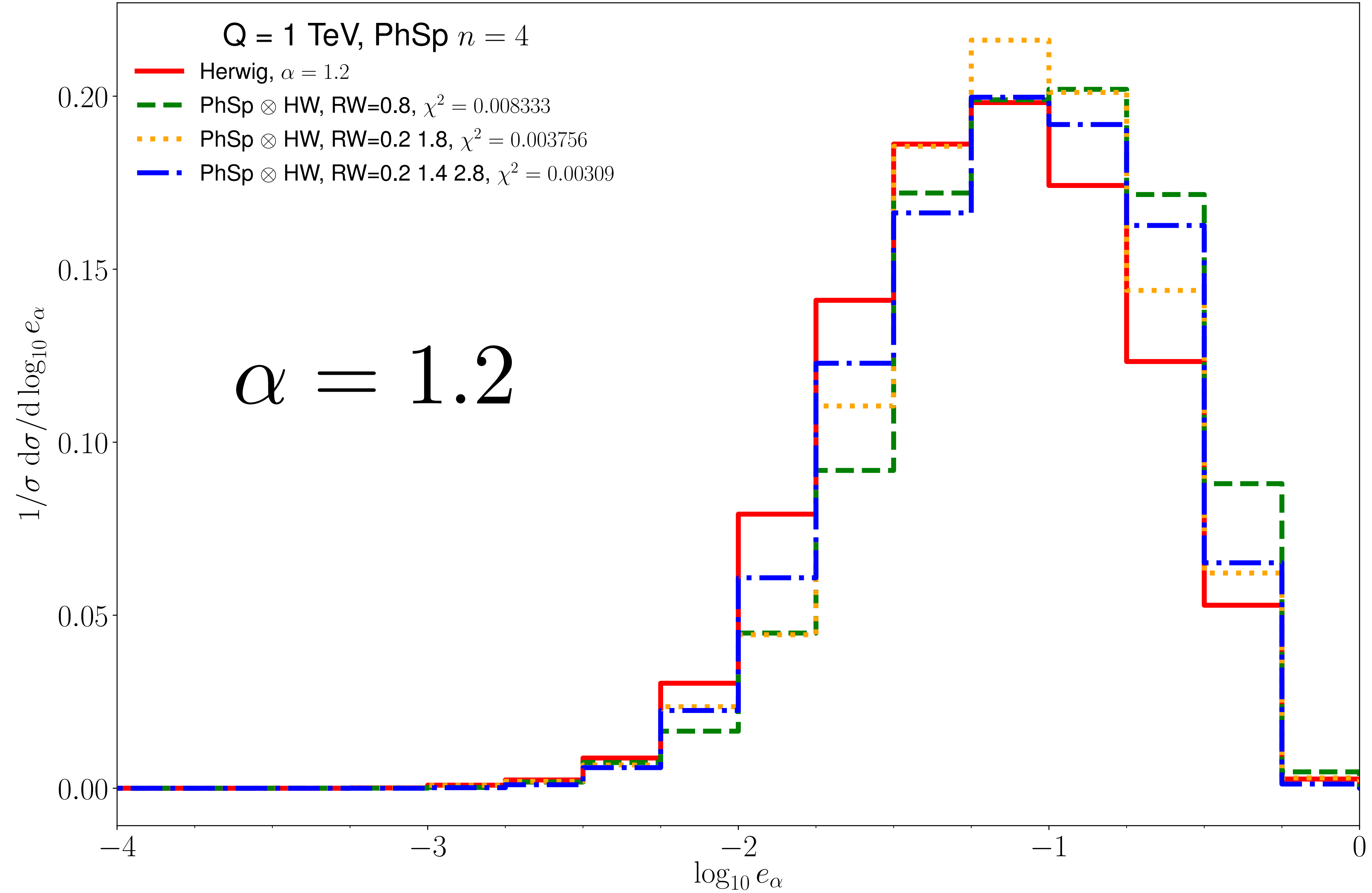
→ Get “best” RW angularity set → by minimising sum of the χ^2 over all “other” angularities:

$$\min \sum_{\alpha} \chi^2(\alpha)$$

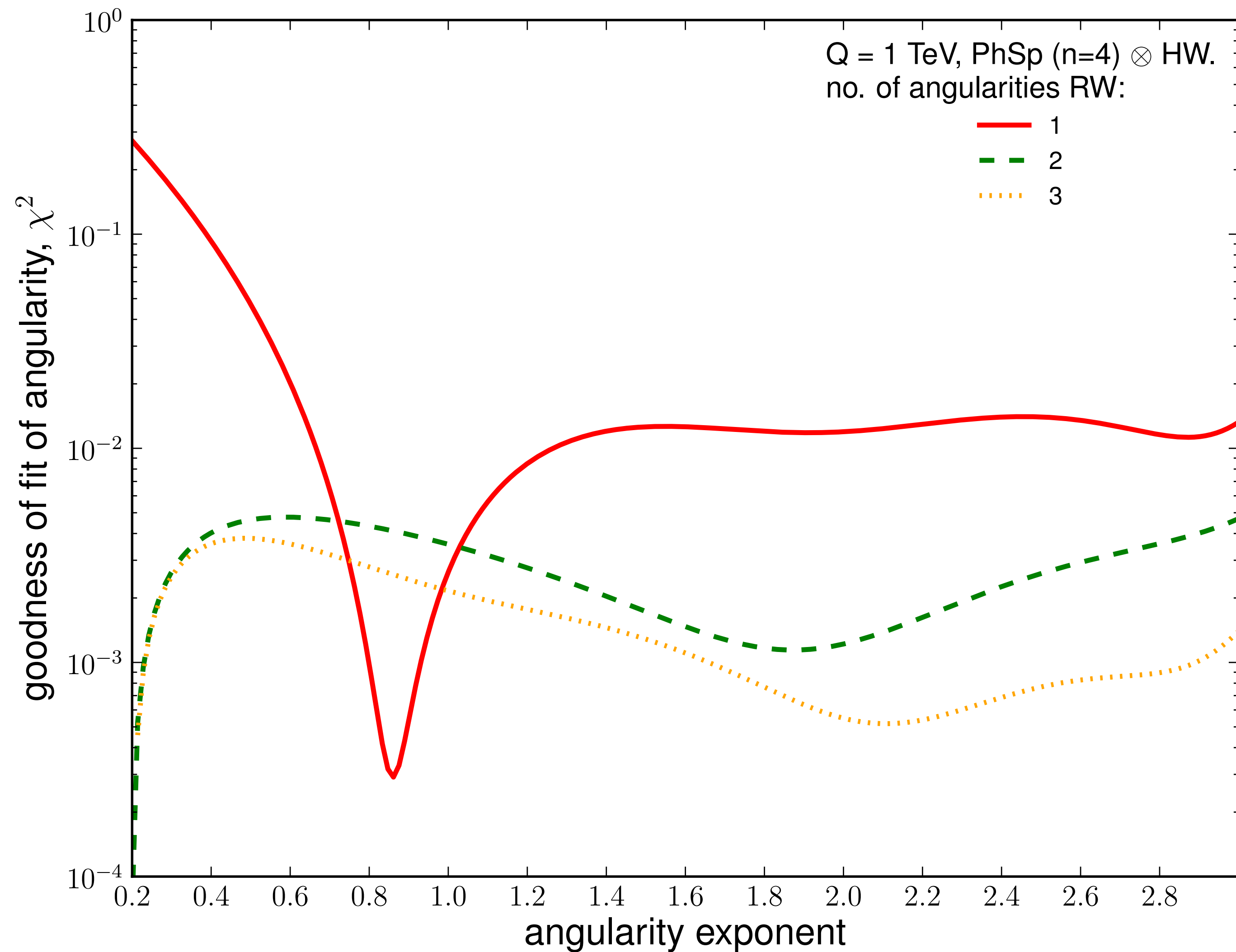
$$Q = 1 \text{ TeV} \quad n = 4$$



$$Q = 1 \text{ TeV} \quad n = 4$$

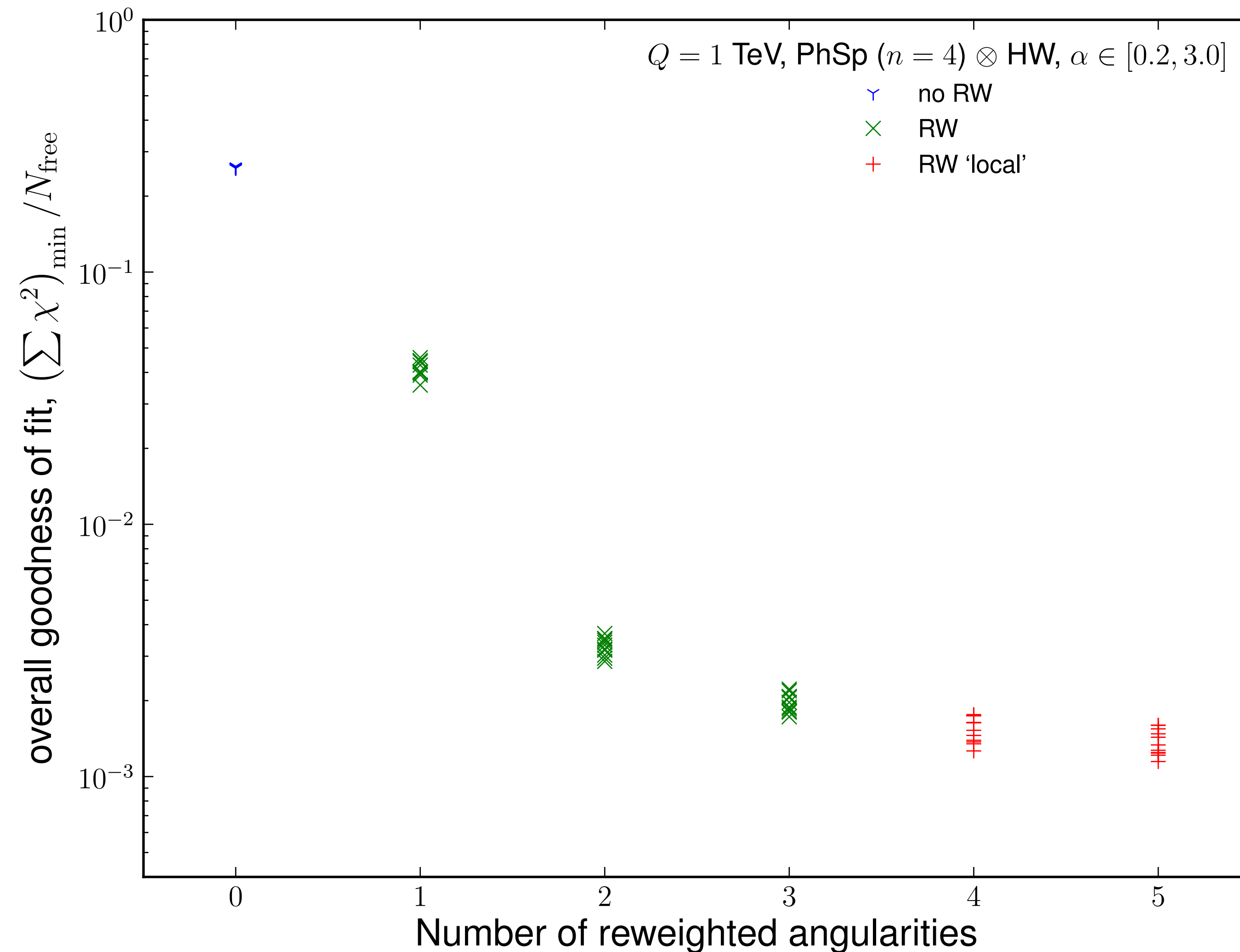


How good is “best” set do for angularity α ?



Lines: the median [over 11 “pseudo-experiments”].

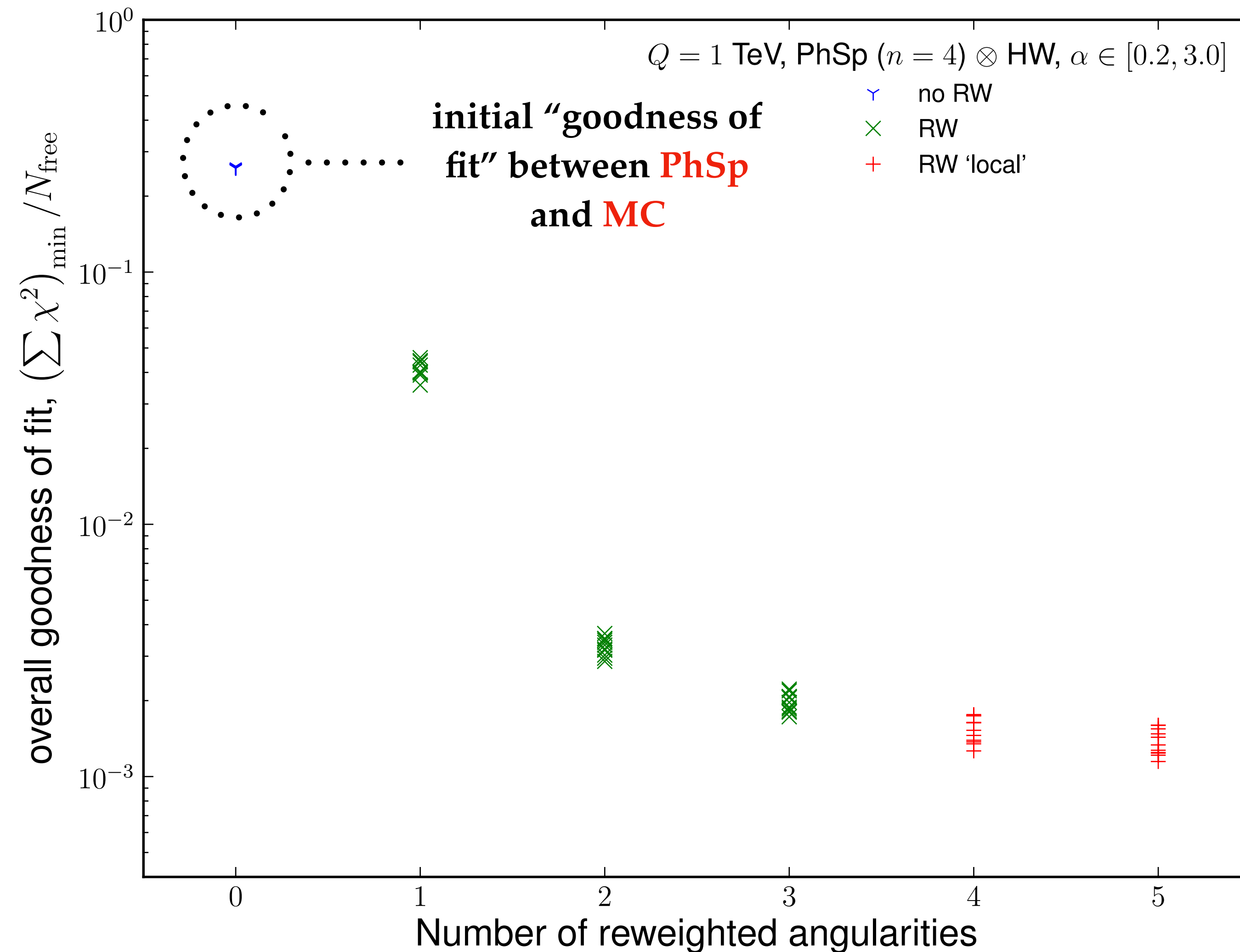
What is the overall improvement?



**Estimate
uncertainty:**

**11 “pseudo-
experiments” →
Get Median &
uncertainty.**

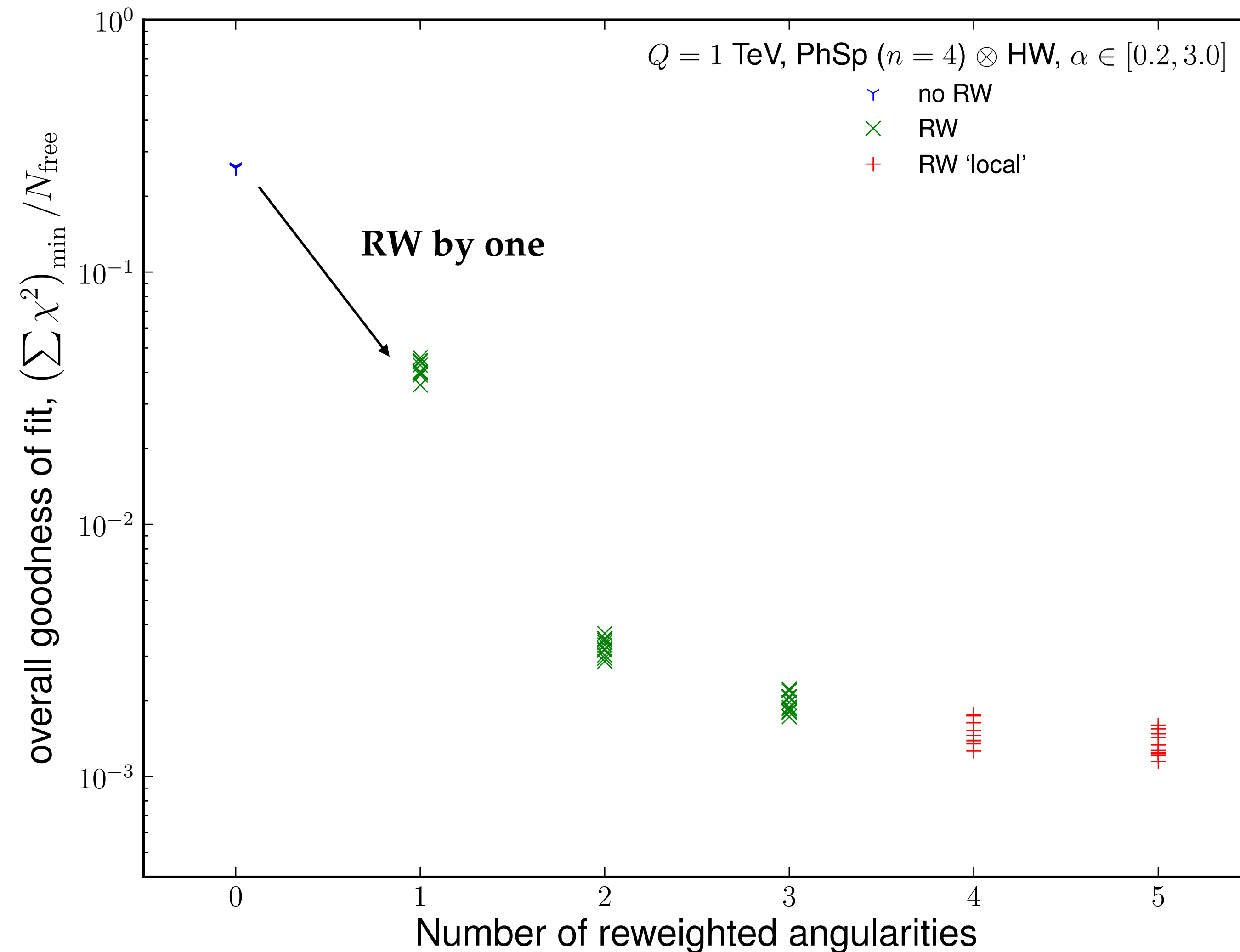
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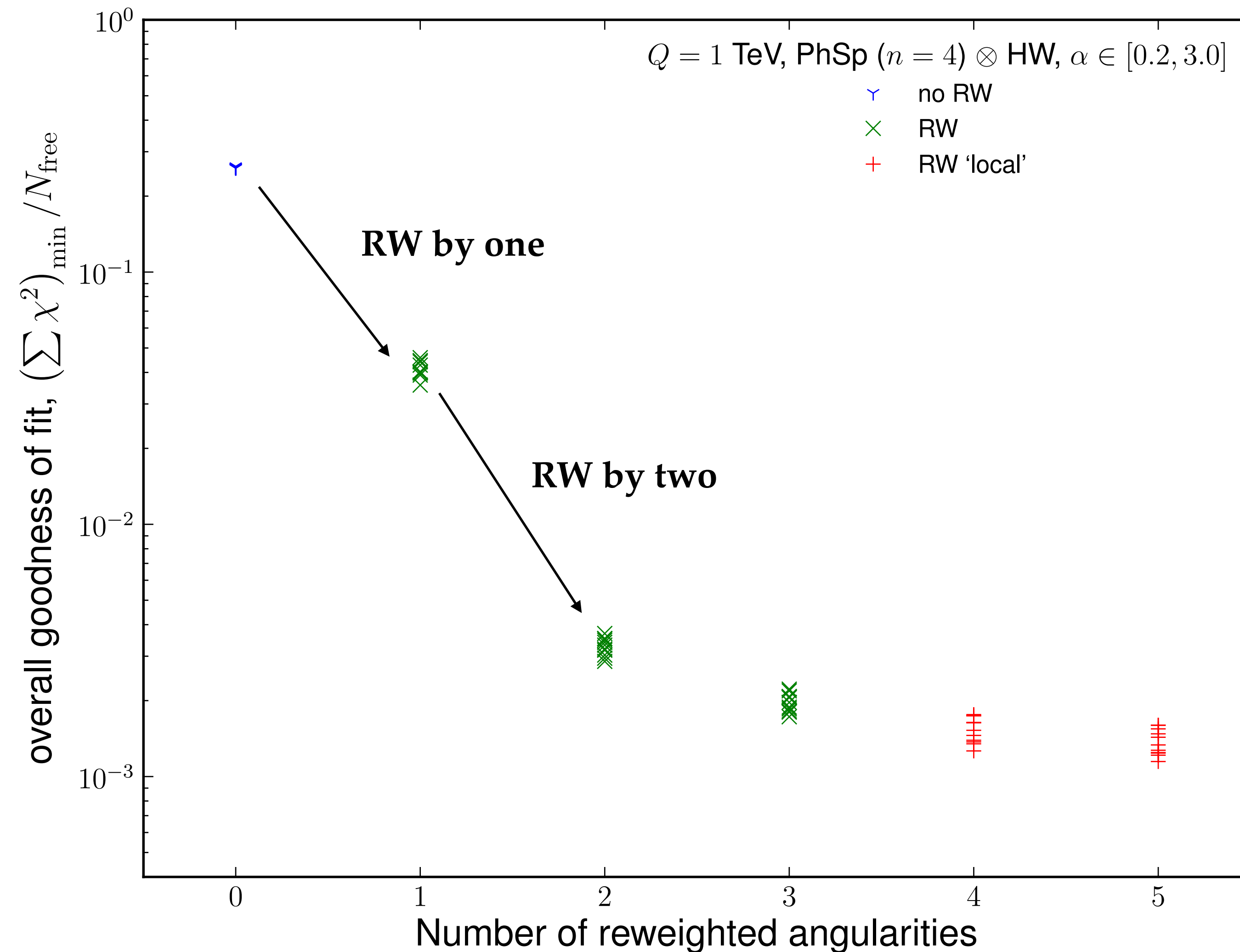
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**Estimate
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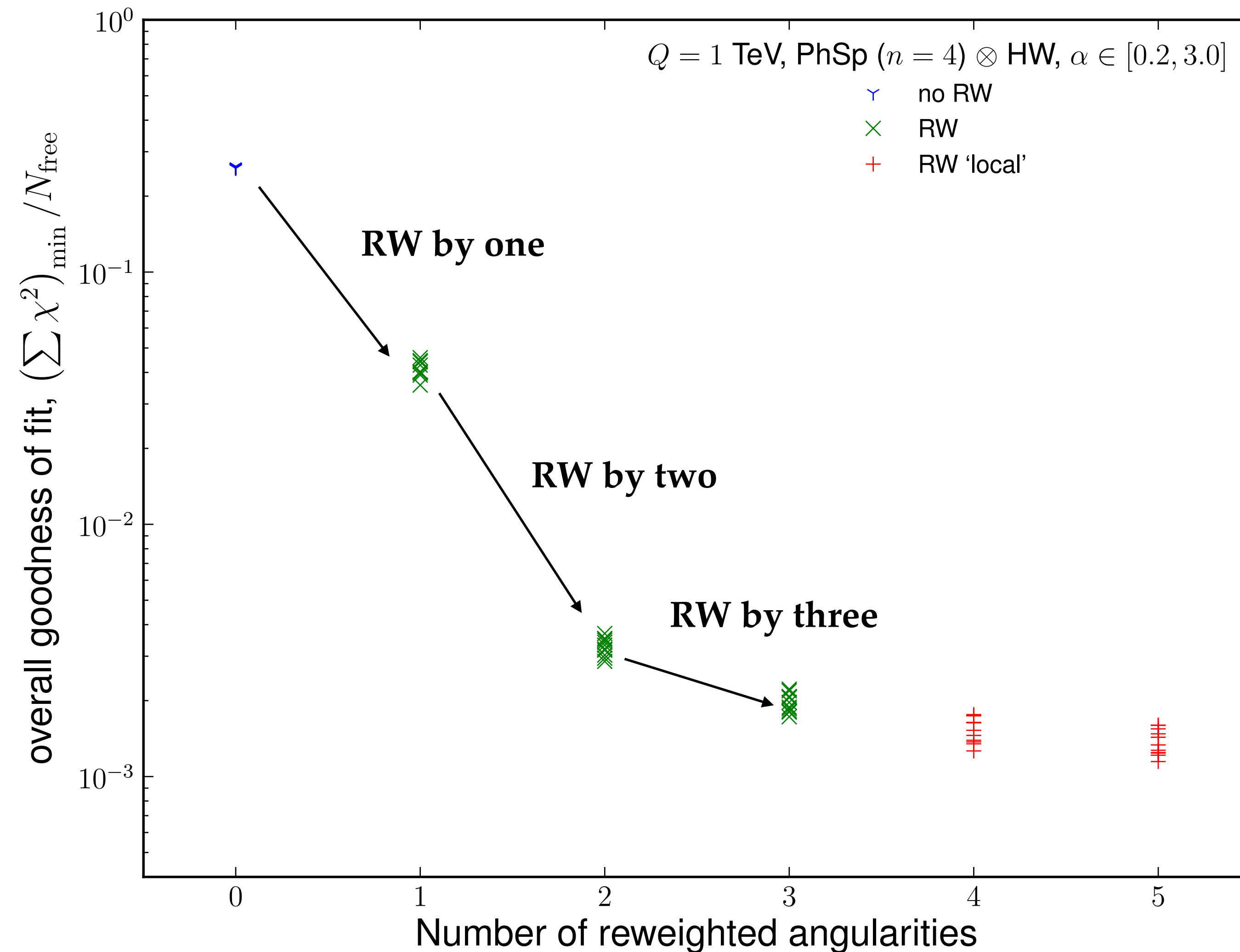
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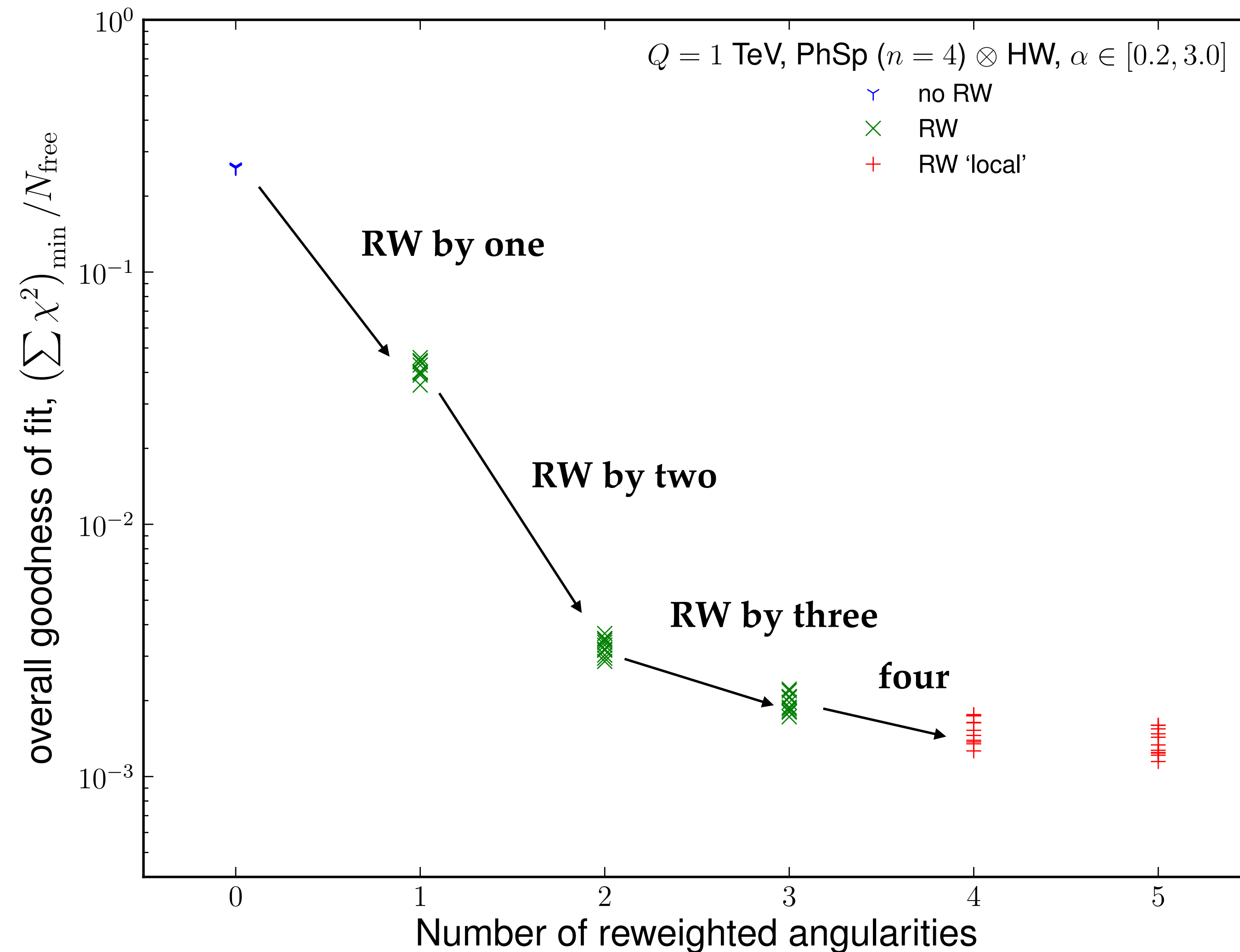
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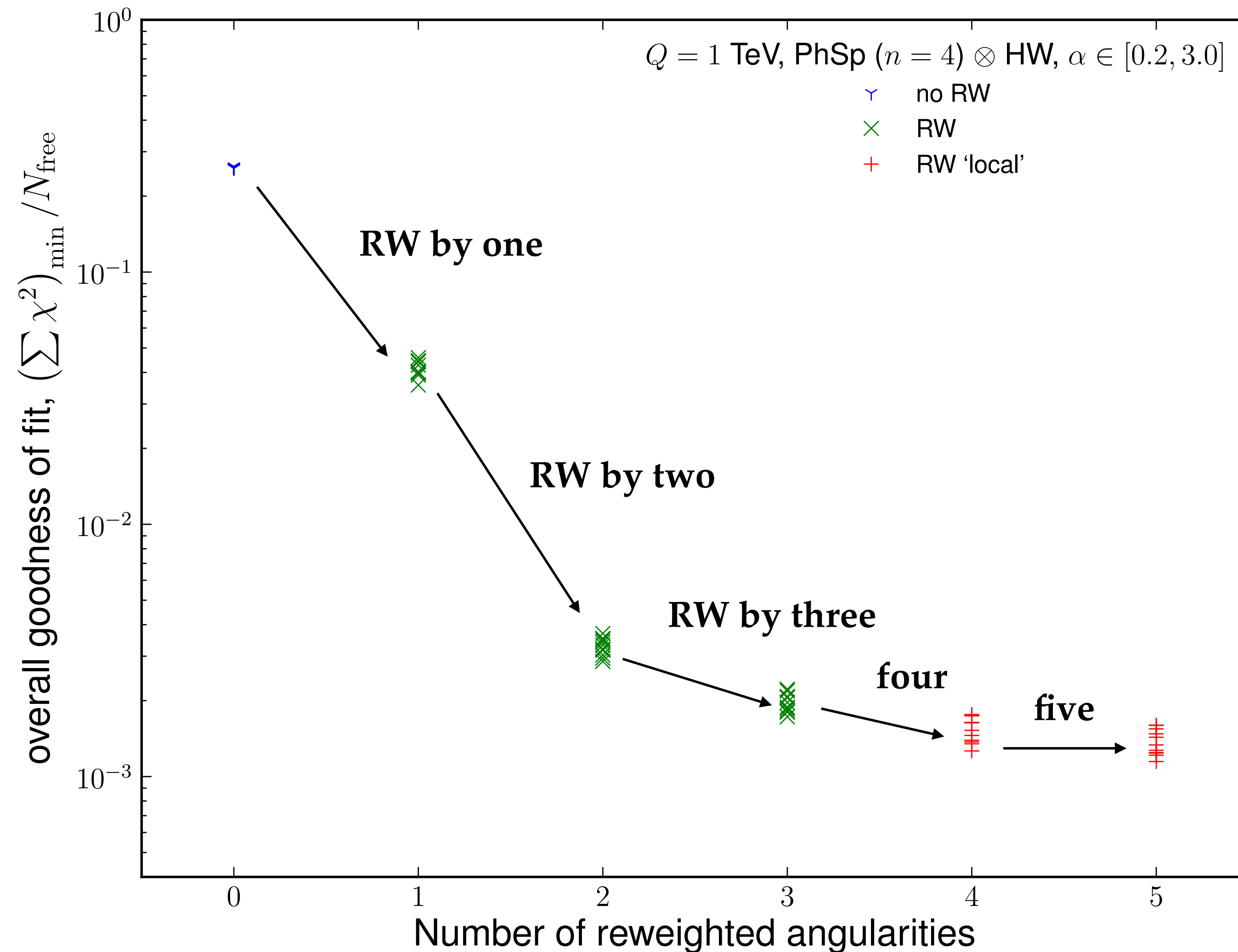
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What is the overall improvement?



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uncertainty:**

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uncertainty.**

Conclusions & Outlook

- Reweighting flat phase space by MC → Get predictions of other angularities.
- Investigate “How much angularity resummation required”:

$$2 > 1$$

$$3 \gtrsim 2$$

$$4 \sim 3$$

[...]

Conclusions & Outlook

- Conclusions **robust** under:
 - ▶ Number of phase space particles,
 - ▶ MC dependence [HERWIG 7, Pythia 8],
 - ▶ COM Energy Q ,
 - ▶ Binning,
 - ▶ Set of angularities.
- Multi-differential resummation: **any number of angularities at NLL**, via **SCET**.

COMING SOON



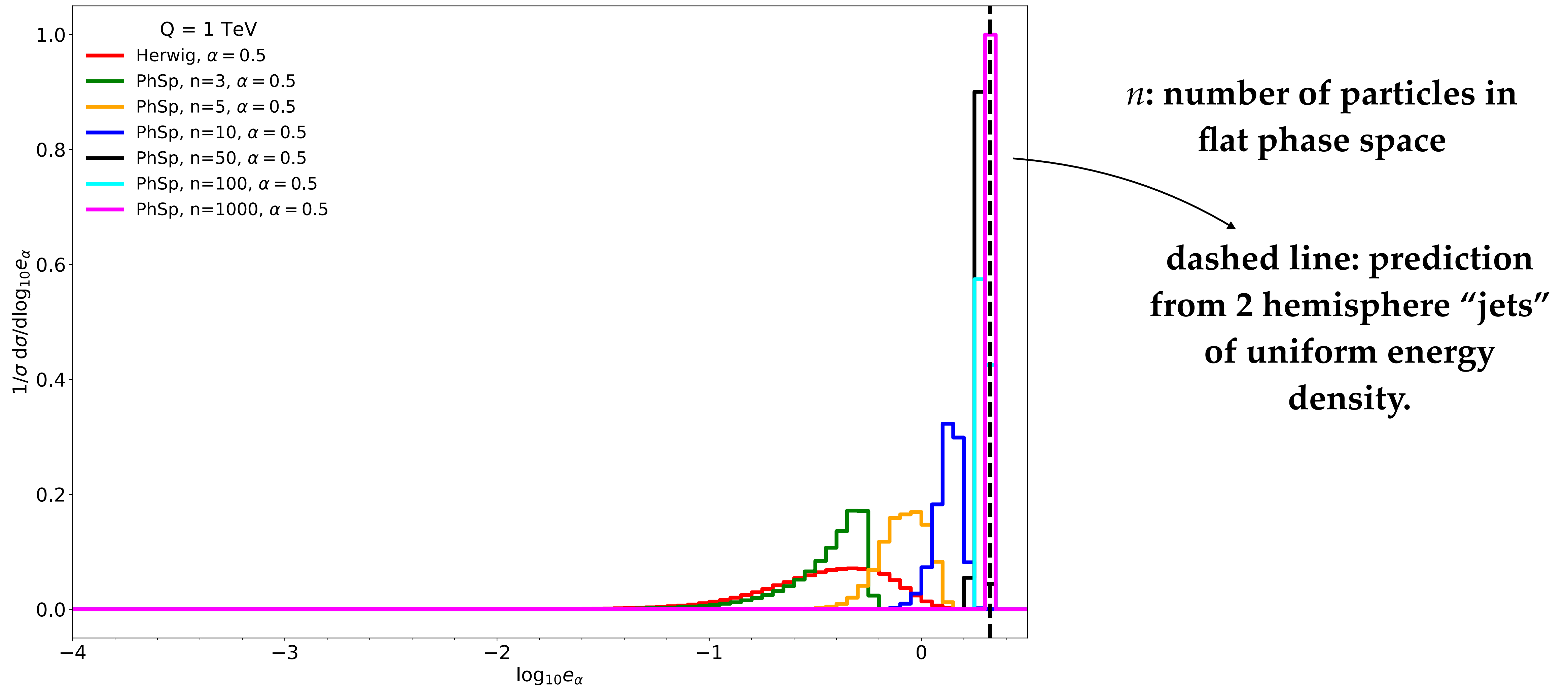
LOADING
PLEASE WAIT...

[with Gillian Lusterms &
Wouter Waalewijn]

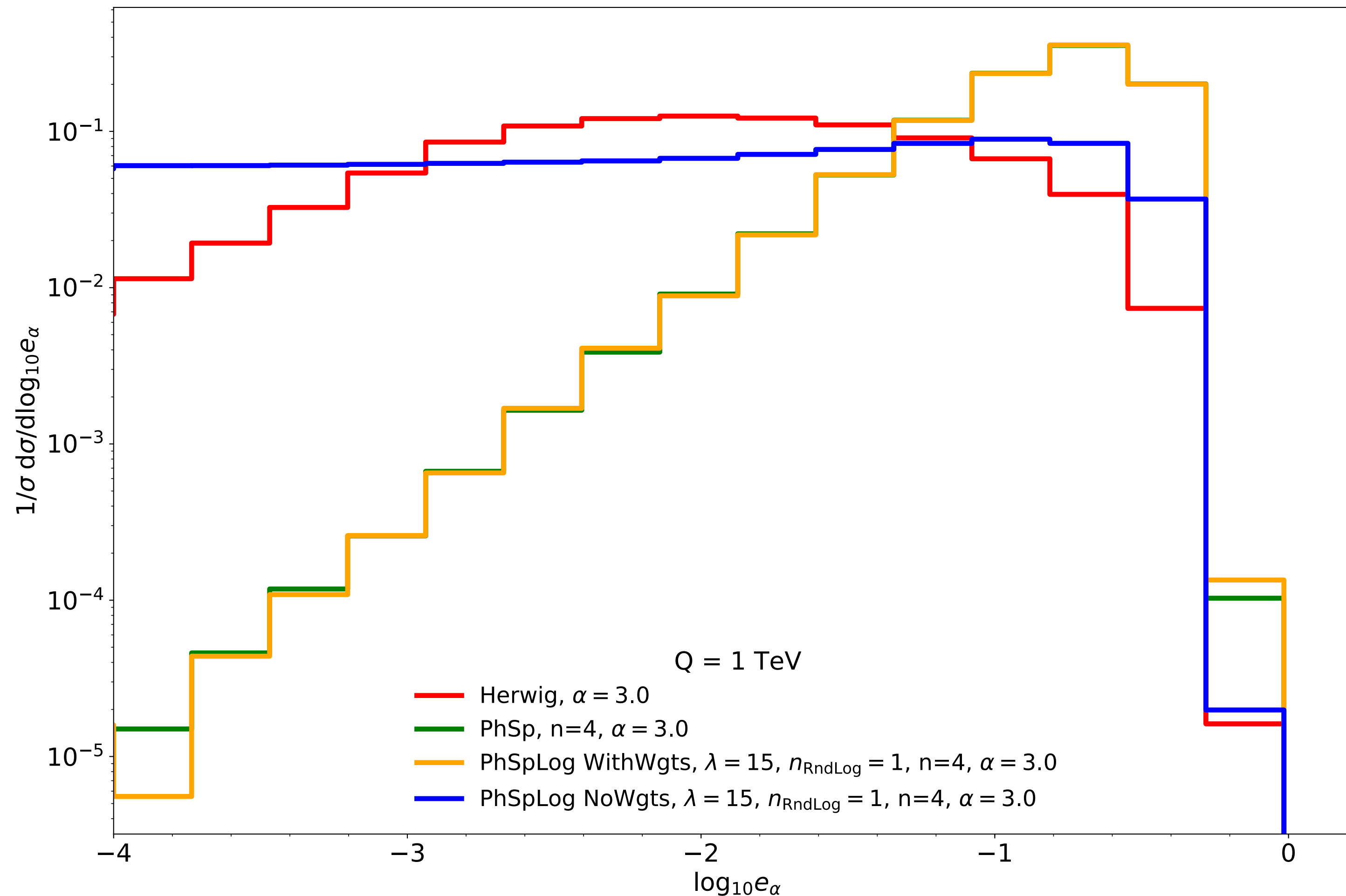
Thanks!

APPENDIX

Flat phase space



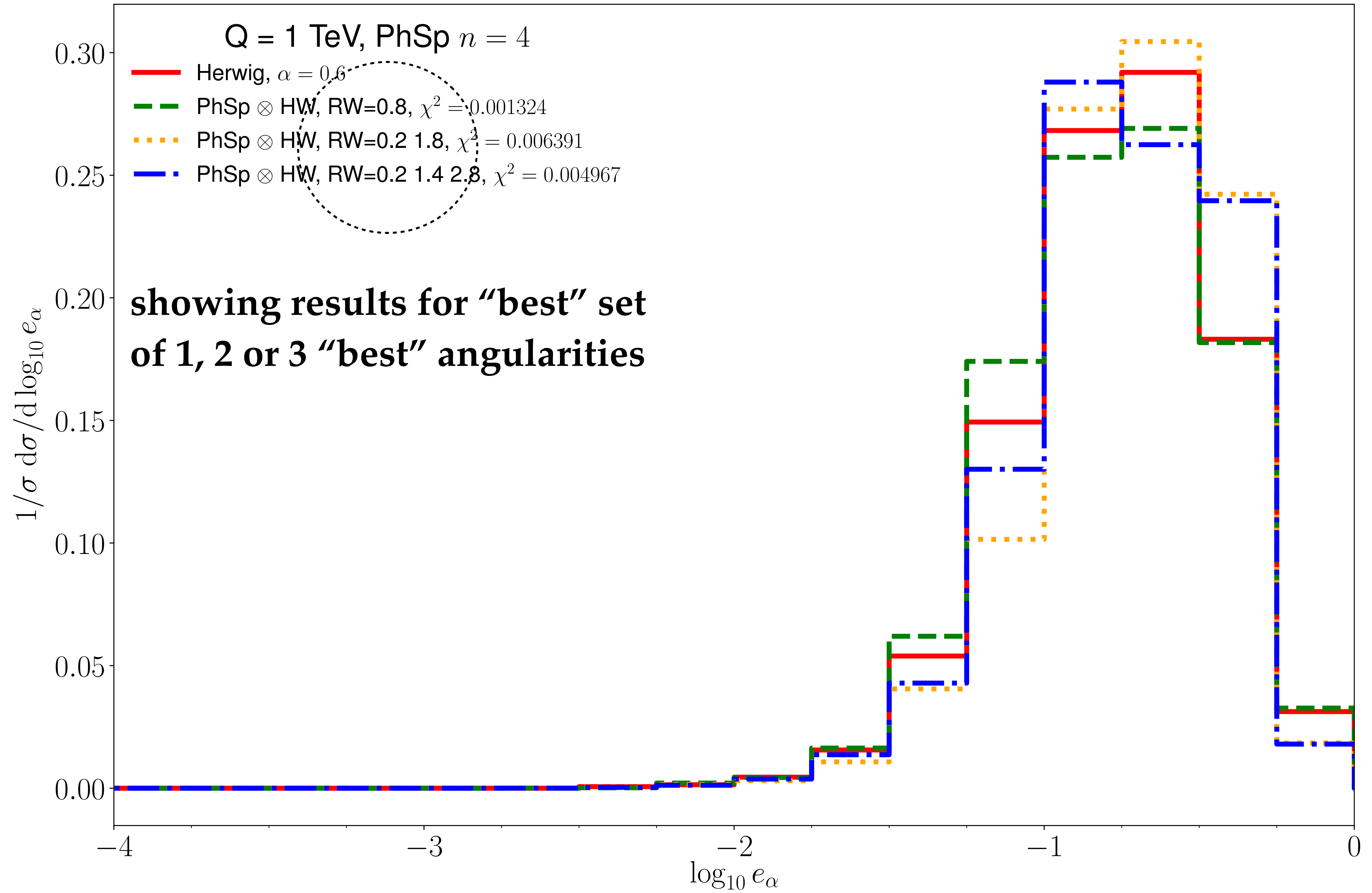
PhSp with “importance sampling”



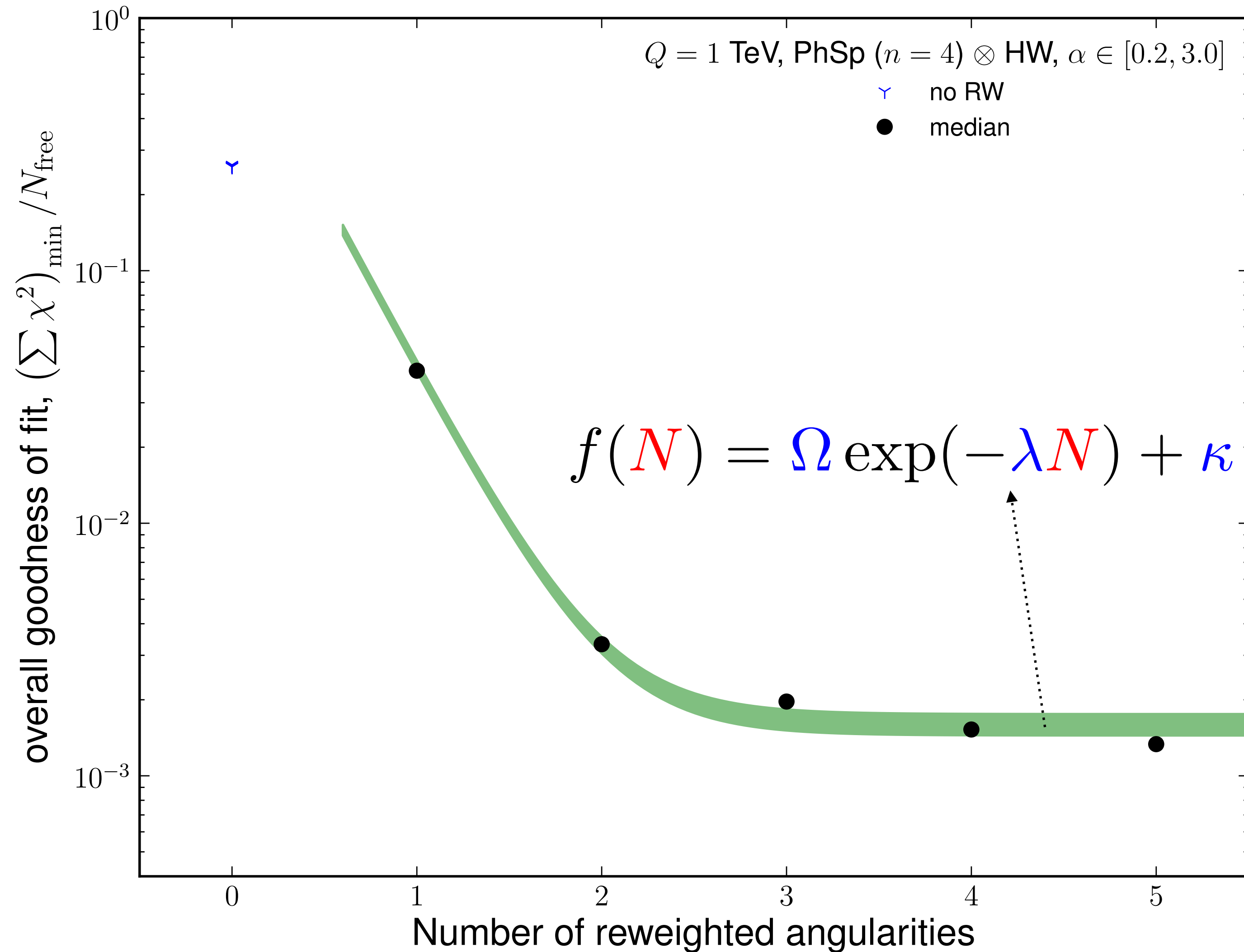
n : number of particles in flat phase space

λ : parameter that determines range of distribution of random numbers in RAMBO-on-diet method.

n_{RndLog} : number of random numbers distributed logarithmically.



What is the overall improvement?

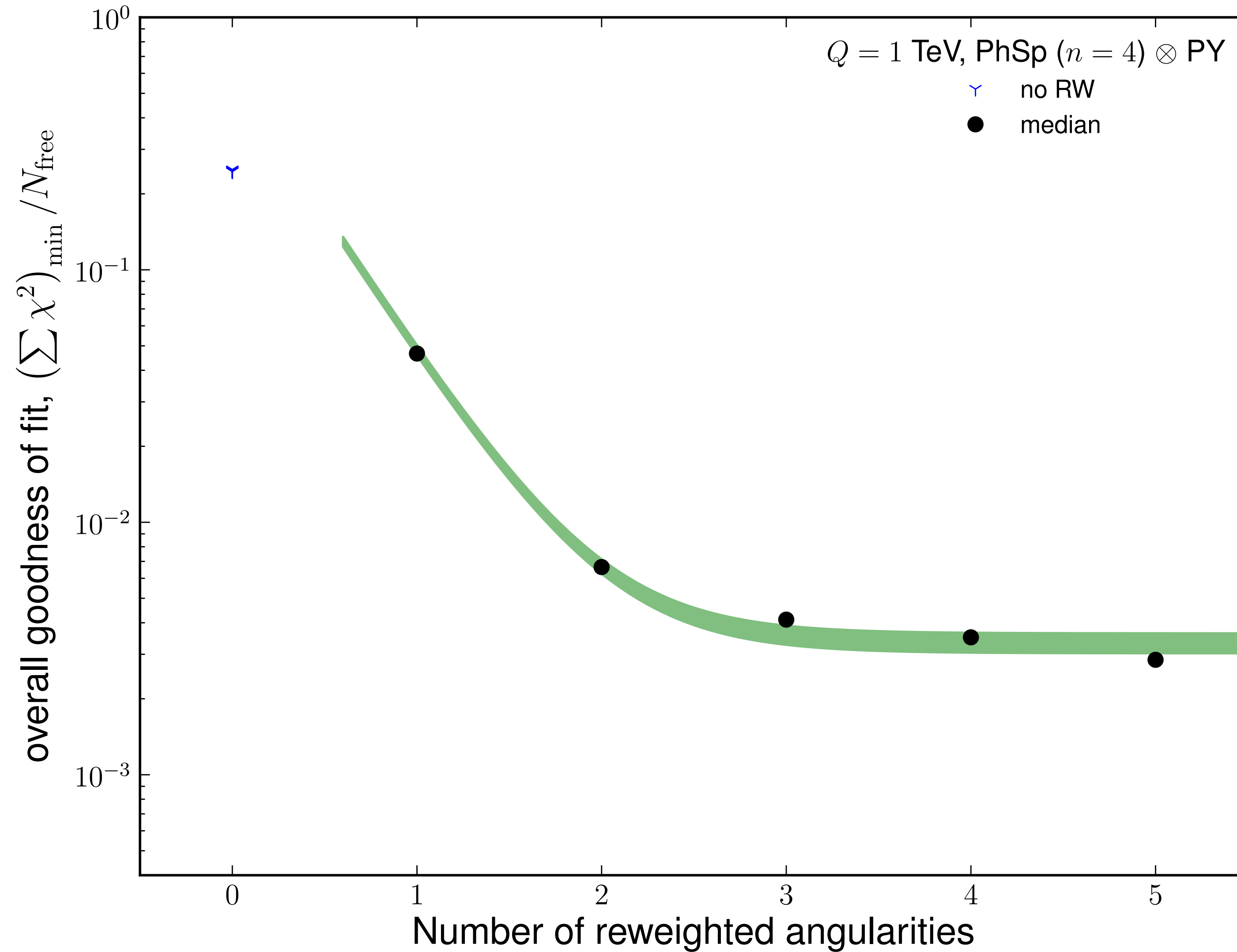


green band: exp. fit of “upper” and “lower” 68%

interval over 11 MC replicas

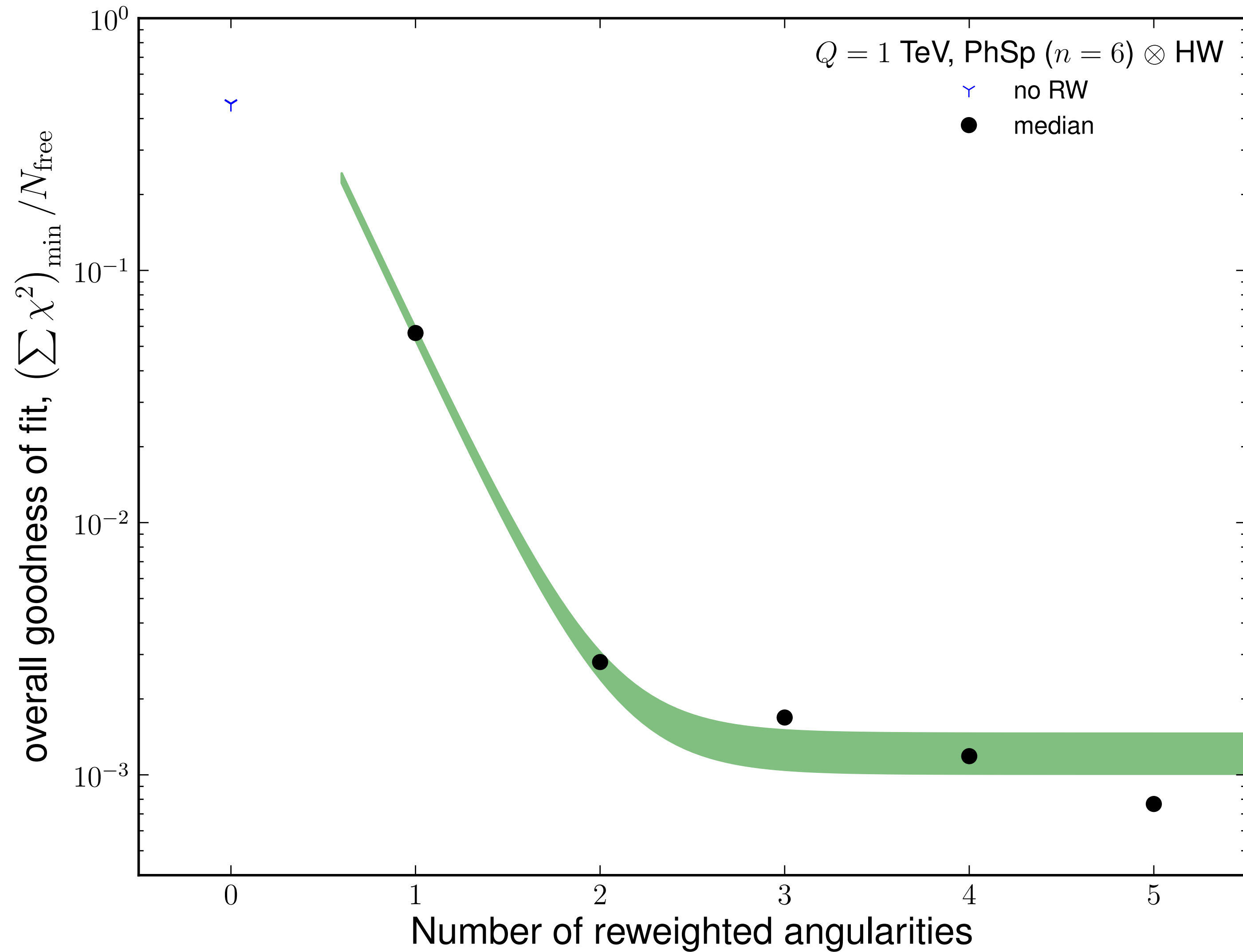
black dots: median

MC dependence



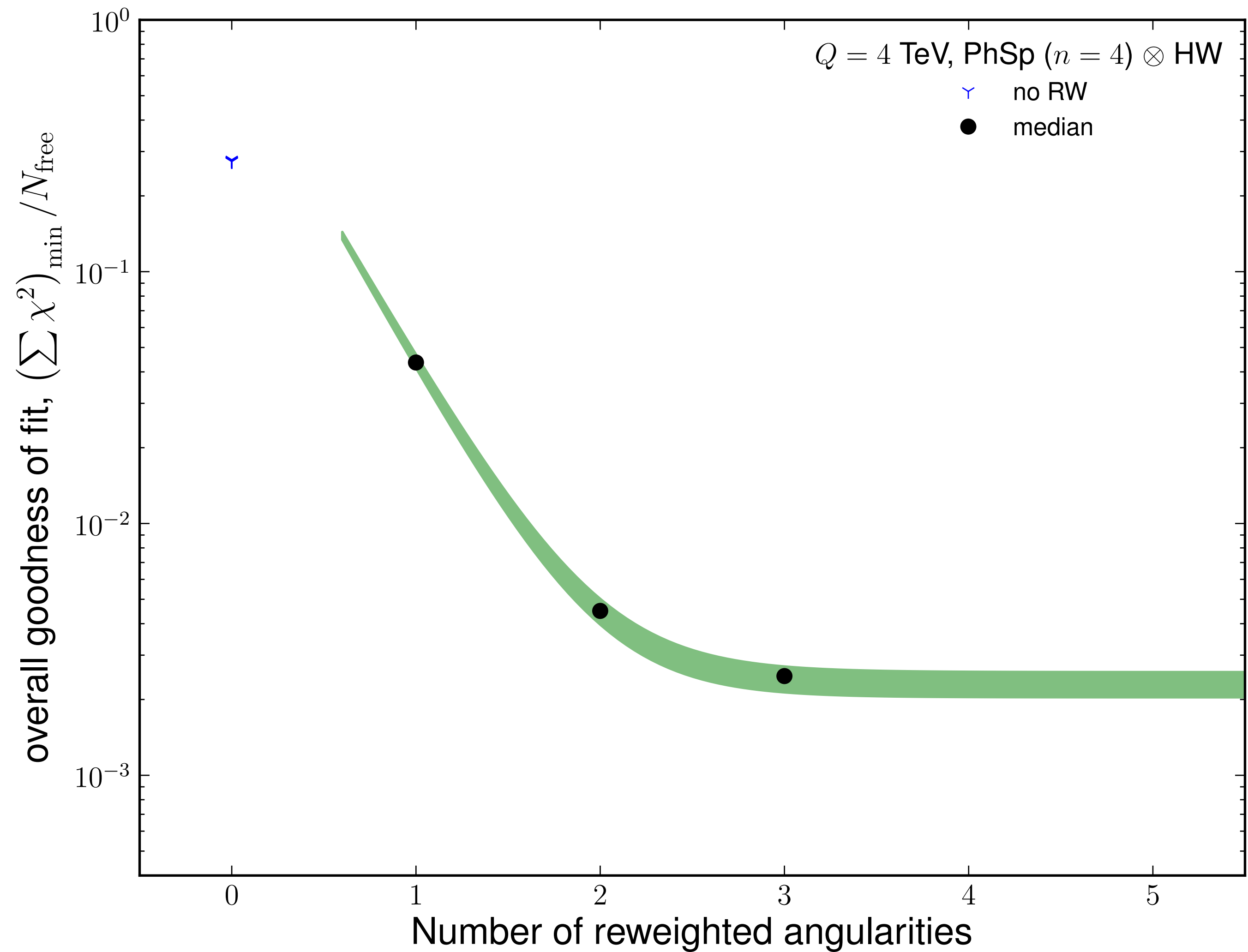
**same as previous
but with Pythia 8**

phase space dependence



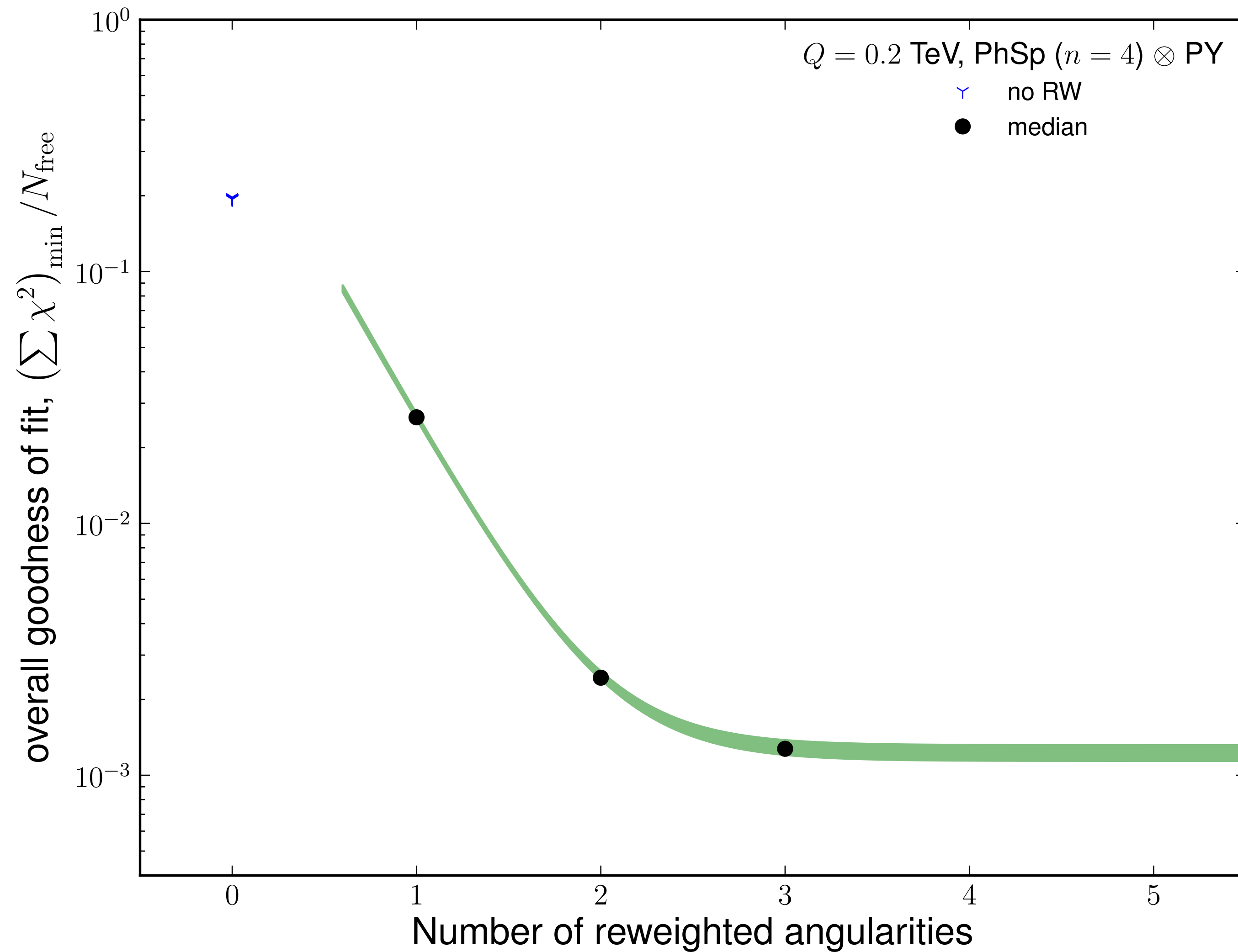
**phase space with
 $n=6$ particles**

COM energy dependence



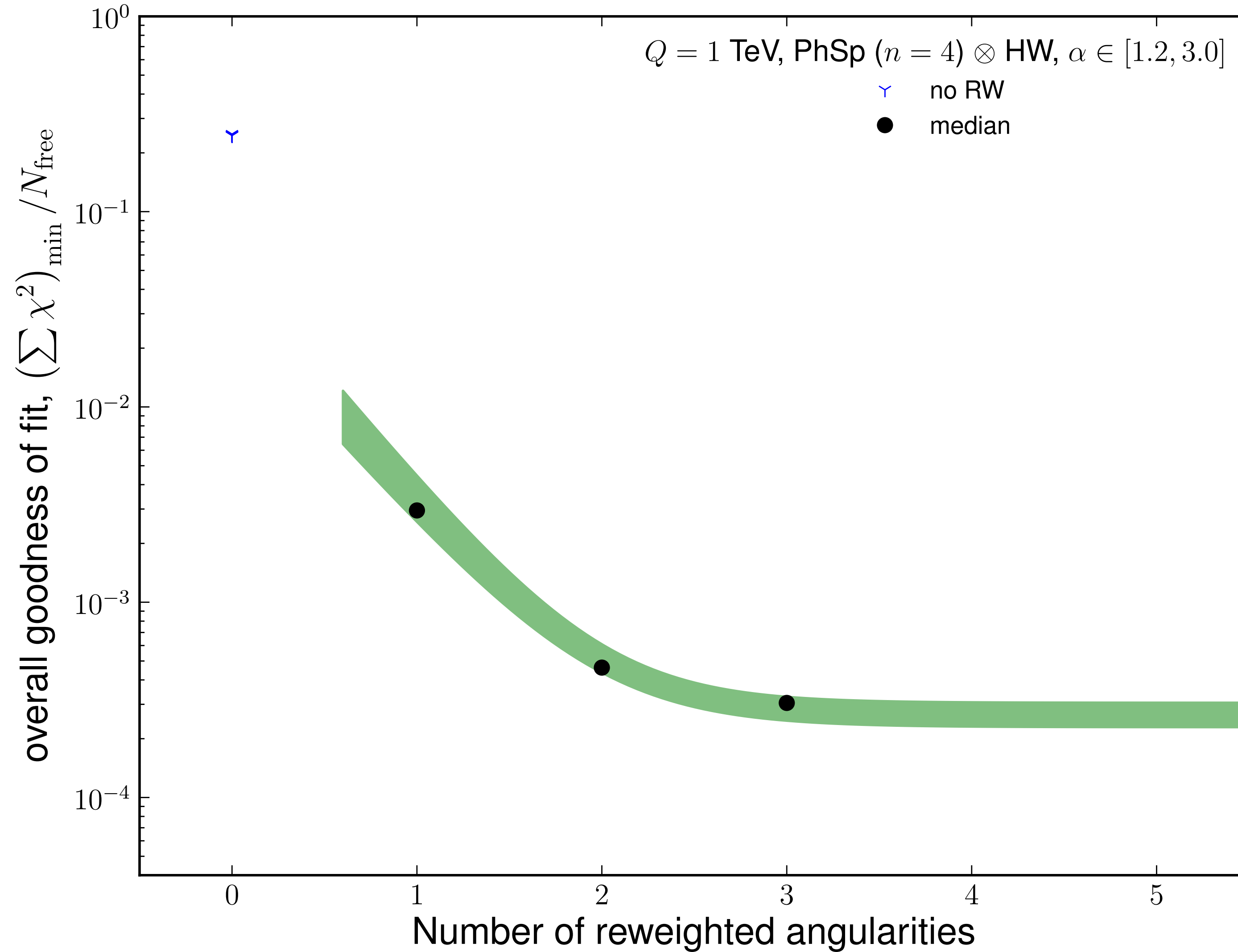
$Q = 4 \text{ TeV}$

COM energy dependence



$Q = 0.2 \text{ TeV}$

set of angularities dependence



$\alpha \in [1.2, 3.0]$

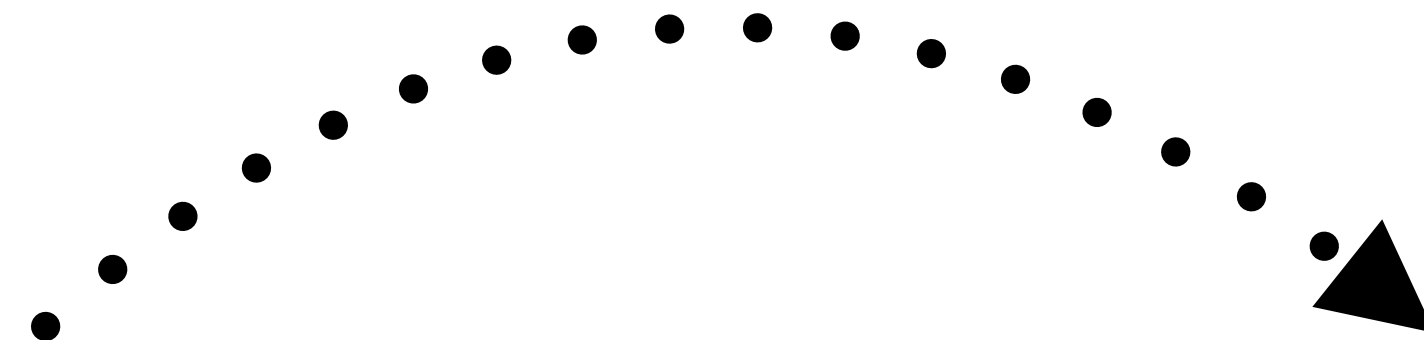
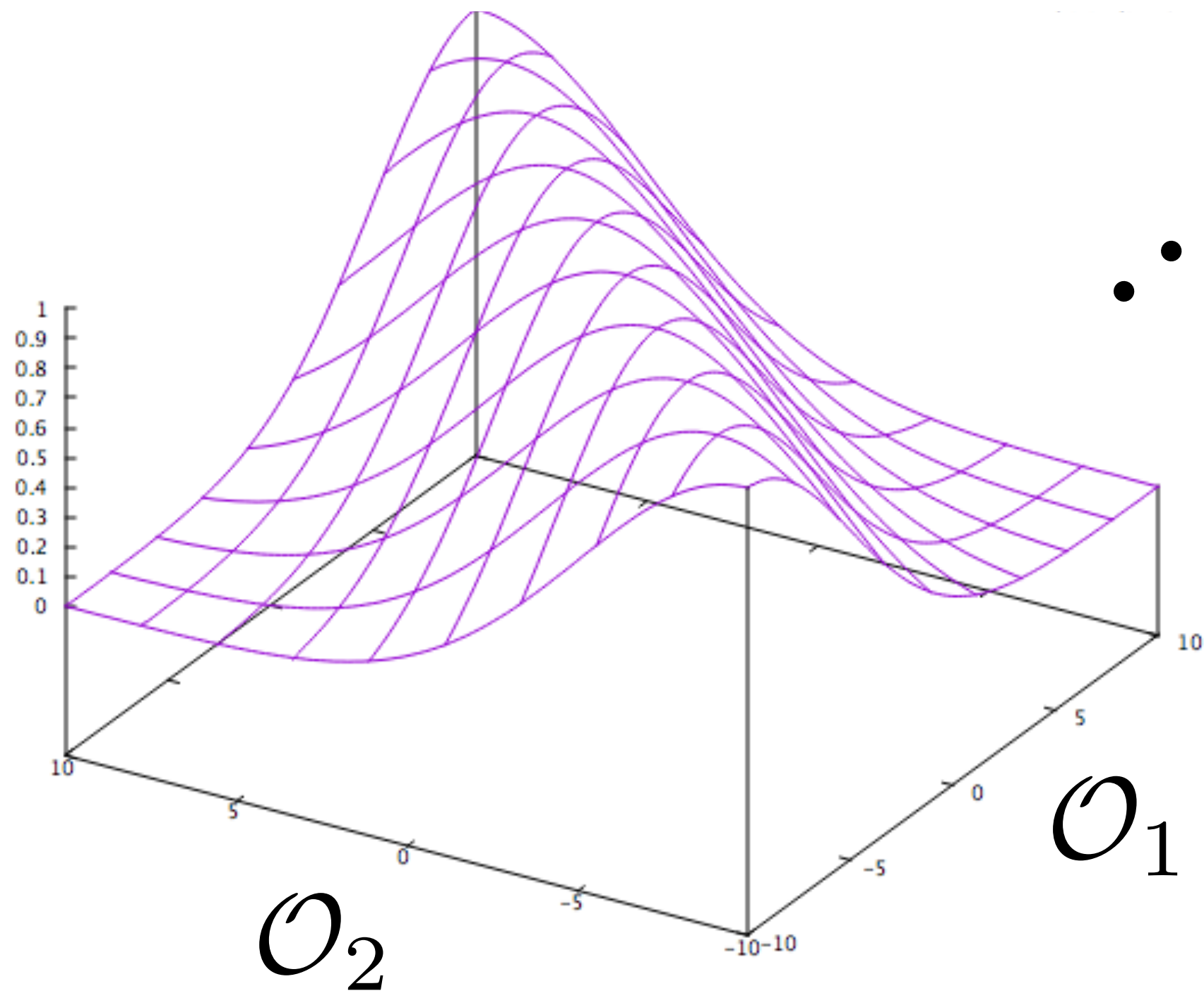
Reweighting

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

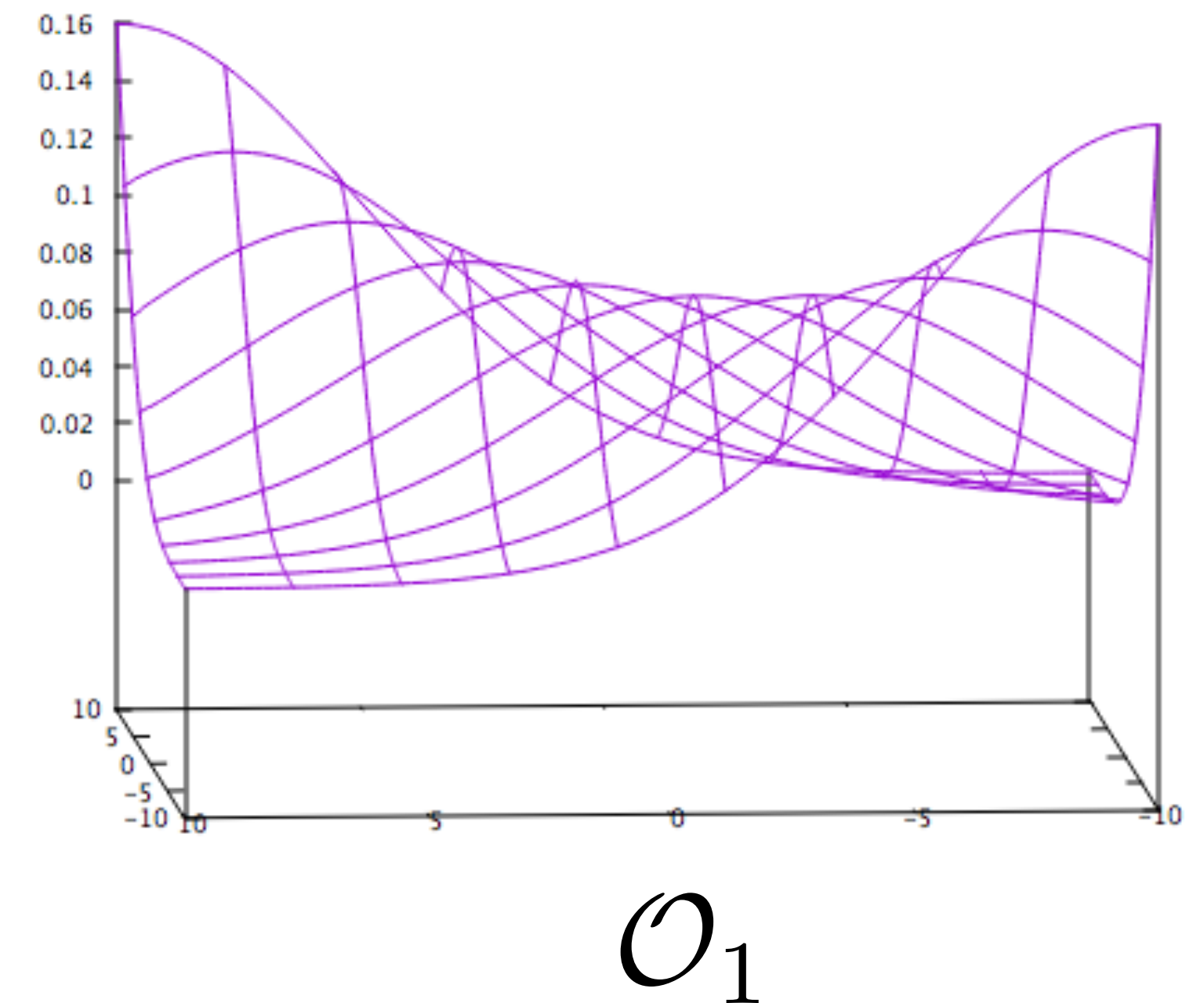
Make "Flat" in
dimension \mathcal{O}_1 :

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

$$\frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{PhSp}}$$



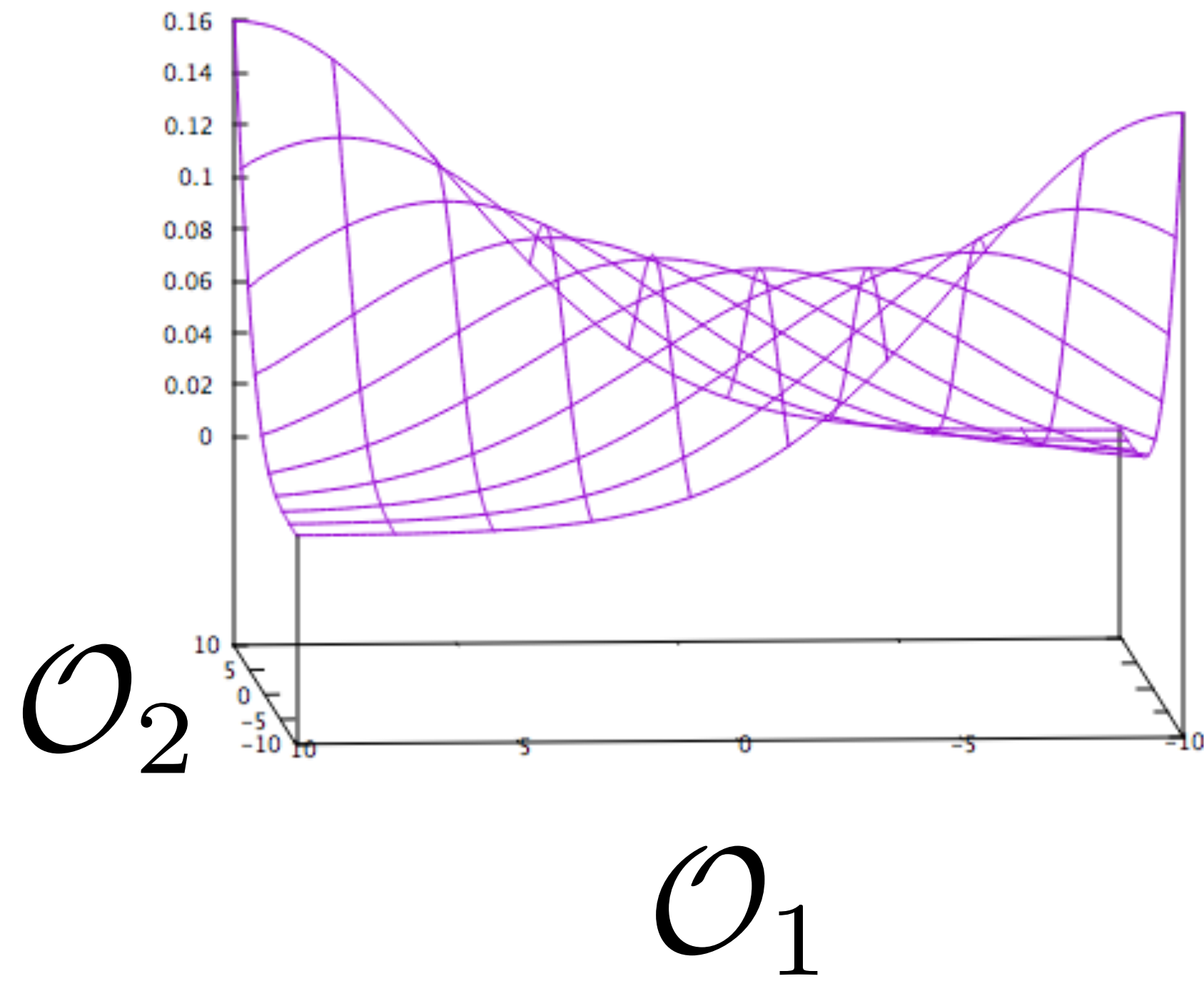
\mathcal{O}_2



Reweighting

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}}$$

$$\frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{PhSp}}$$



Reweigh by
MC \mathcal{O}_1 distribution

$$\frac{d\sigma}{d\mathcal{O}_1 d\mathcal{O}_2} \Big|_{\text{PhSp}} \times \frac{d\sigma}{d\mathcal{O}_1} \Big|_{\text{MC}}$$

