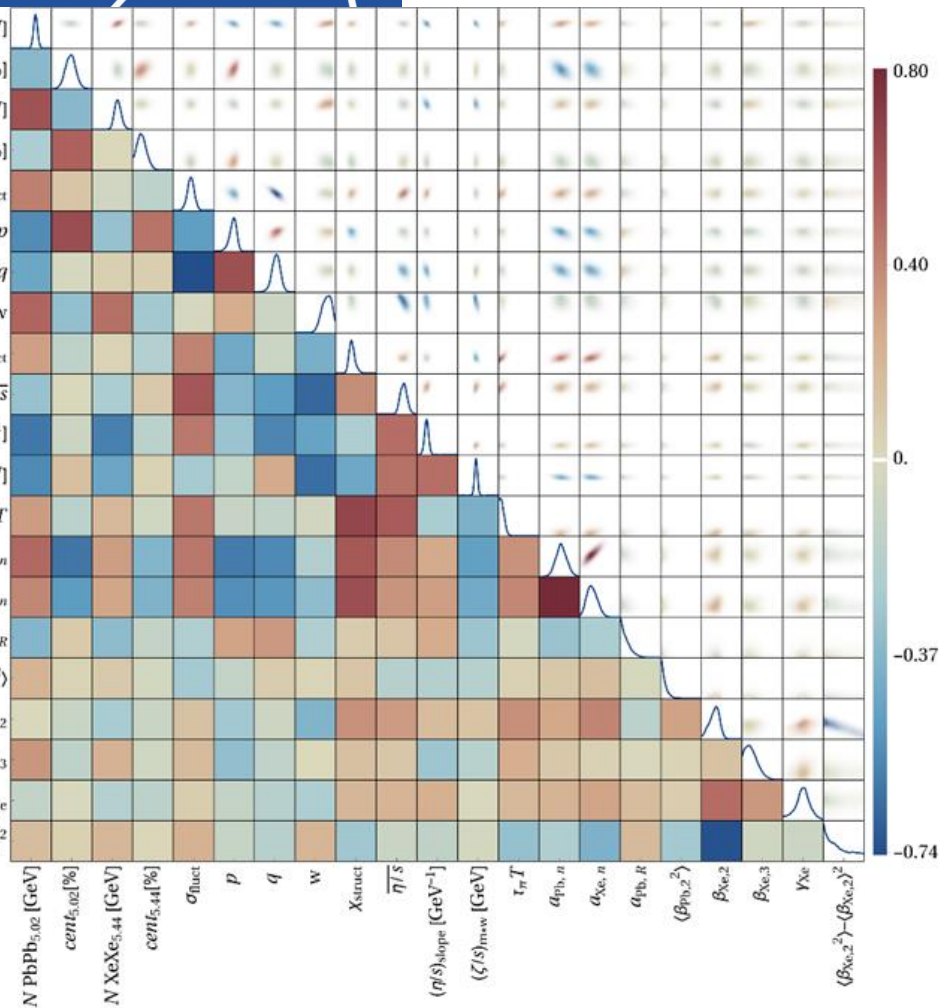
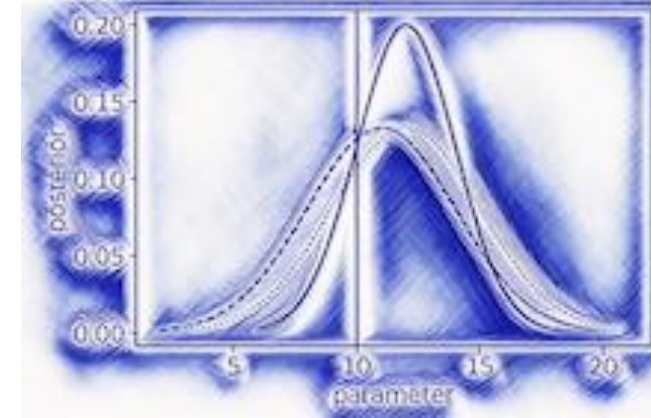
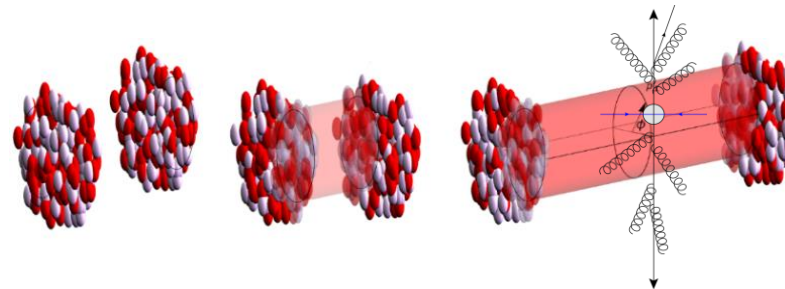




Universiteit
Utrecht



Bayesian inference in heavy ion collisions



Wilke van der Schee

Precision Frontier of QCD Matter: Inference and Uncertainty Quantification

Wuhan, China, 2 September 2025

Bayes theorem is about uncertainties

A mathematical theorem with three basic ingredients:

1. A (physics) model with experimental data
Parameter posterior not realistic if model is not realistic
2. Prior probability
Often more relevant than one would wish
3. Uncertainties:
 1. Statistical and emulator uncertainties are relatively easy
 2. Correlated observables already significantly harder
 3. Systematic (model) uncertainty often impossible to estimate
Becomes more relevant if emulator is more accurate

What happens if we have many parameters? Too few parameters? Too many data points? Too few datapoints?

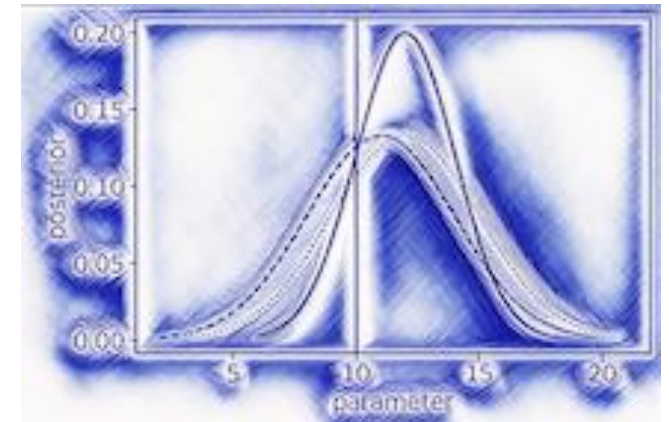
Can we trust our posterior?

Only with a realistic model (for data considered) and realistic uncertainty

Bayes theorem:

$$\mathcal{P}(\mathbf{x}|\mathbf{y}_{\text{exp}}) = \frac{e^{-\Delta^2/2}}{\sqrt{(2\pi)^n \det(\Sigma(\mathbf{x}))}} \mathcal{P}(\mathbf{x})$$

$$\text{with } \Delta^2 = (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}}) \cdot \Sigma(\mathbf{x})^{-1} \cdot (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}})$$



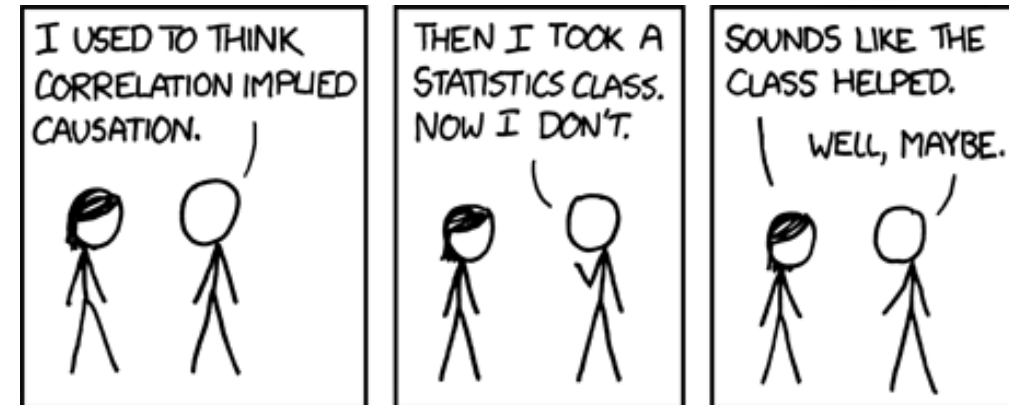
Two messages --- precision era of HIC

Wealth of data poses a significant challenge

1. Necessary to fully utilise experimental knowledge
2. Correlation matrix often important
(Unknown) systematic uncertainty often important
Rarely data follows standard χ^2 distribution ...

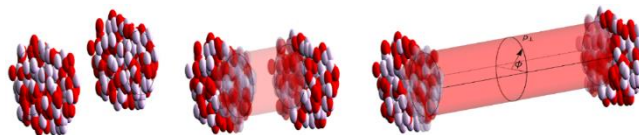
Often necessary to 'interpret' results:

1. Some observables may be more reliable than others
2. Some parameters can be trusted better than others



Correlation doesn't imply causation, but it does waggle its eyebrows suggestively and gesture furtively while mouthing 'look over there'.

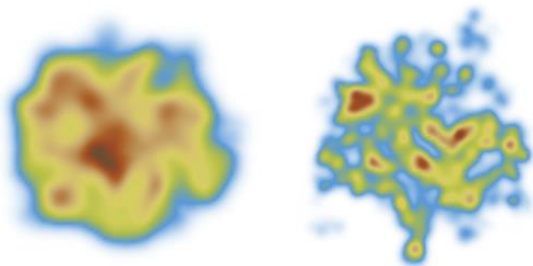
Standard model of heavy ion collisions



(# parameters)

Initial stage (13)

Subnucleonic structure? (8)



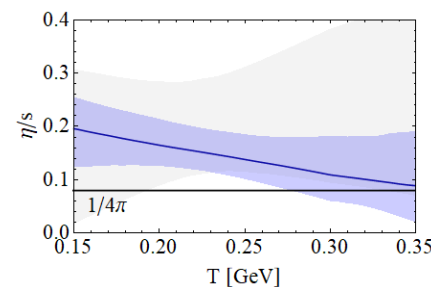
Non-thermal flow? (2)
with hydrodynamised initial stage

Fluctuations? (1)

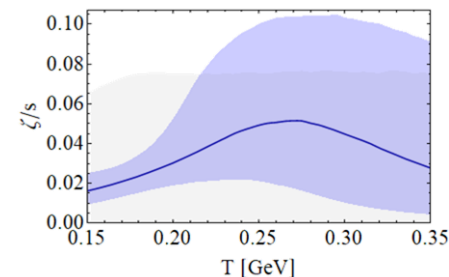
Shape (2)

Viscous hydrodynamics (10)

Shear viscosity (4)



Bulk viscosity (3)

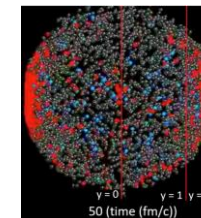


Second order transports: 2

EOS: 1

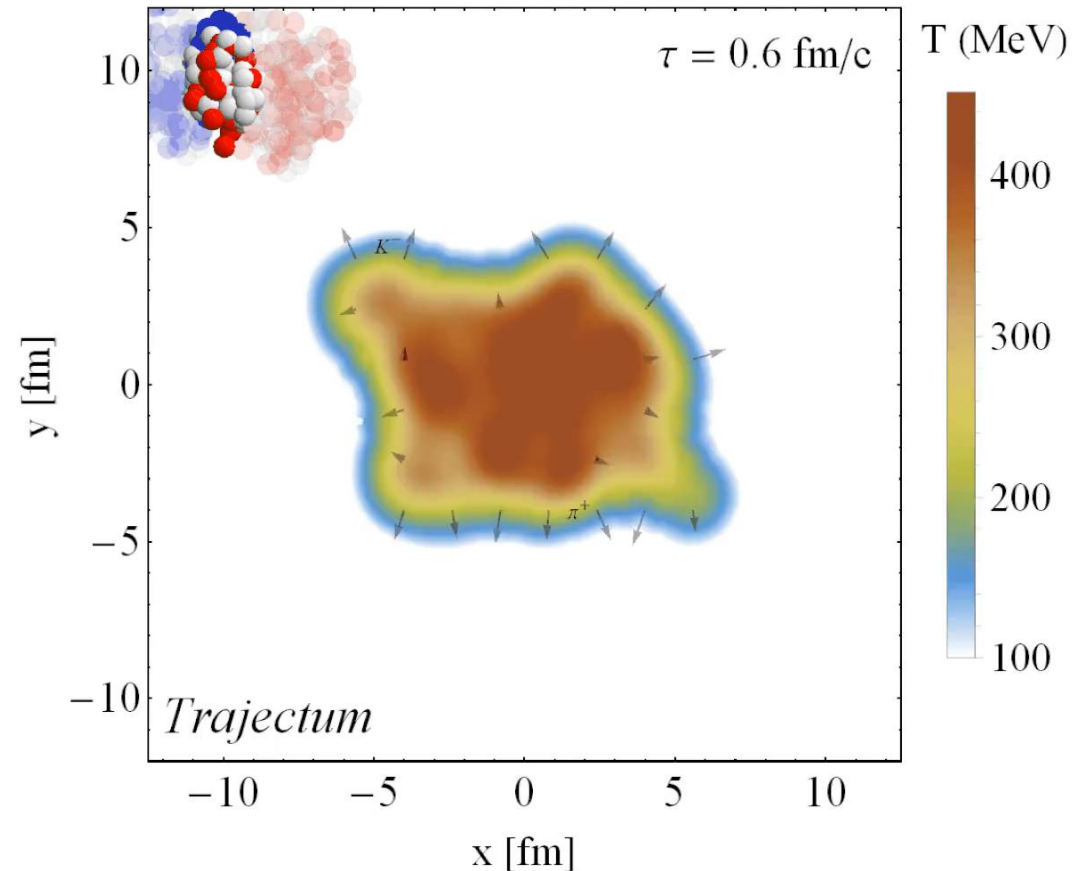
Cascade of hadrons (2)

SMASH



A visualisation

1. Depends on parameters (see param file, right)
2. In this case *Trajectum* (many other models available)
3. Compute observables as close to experiment as possible



```

general{
  output=out
  format=smash
  f0500=false
  numevents=1
  seed=7398984.747399307
  debugoutput=true
  numthreads=2
}

entropyacceptanceprobability{
  0:0.0
  24:0.0
  24.5:0.05
  25.5:0.05
  26:0.0
  100:0.0
}

trentosubstructurePbPb{
  dmin=0.63933
  w=0.701919
  sigmann=70.0
  sigmafluct=0.73579
  p=0.14388
  q=1.0
  Eref=0.2
  norm=23.507
  freestreamingreferencetime=1.1708
  freestreamingvelocity=0.62672
  weaktostrong=0.0
  nref=20
  alpha=0
  nc=3.2747
  voverw=0.4892041602706295
}

secondorderhydro{
  numlatticesites=166.0
  latticesize=33.2
}

musclsolverktminmodfastmidpoint{
  cflconstant=0.08
}

LatticeE0StempdepDuke{
  shearhg=0.0895066
  shearmin=0.0895066
  shearslope=0.43252
  shearcrv=0.231195
  shearrelaxationtime=6.318855
  bulkmax=0.0030138
  bulkT0=0.21471
  bulkwidth=0.10906
  bulkrelaxationtime=0.0687
  deltapiiovertaui=1.3333333333333
  phi7overpressure=0.128571
  taupiovertaui=1.61033
  lambdaPiPiovertaui=1.2
  deltaPiPiovertaui=0.6666666666666
  lambdaPiPiovertaui=1.6
  phi1overpressure=0
  phi3overpressure=0
  phi6overpressure=0
}

cooperfryehadronizer{
  freezeouttemp=153.456
  rapidityrange=0.1
}
  
```

Performing a global analysis

Model depends on parameters non-linearly

- Run model on 3000 'design' points
- Use an emulator for any point in parameter space (**GP**)

Markov Chain Monte Carlo

- 653 data points
- Obtain posterior probability density of parameters

Compare posterior with data

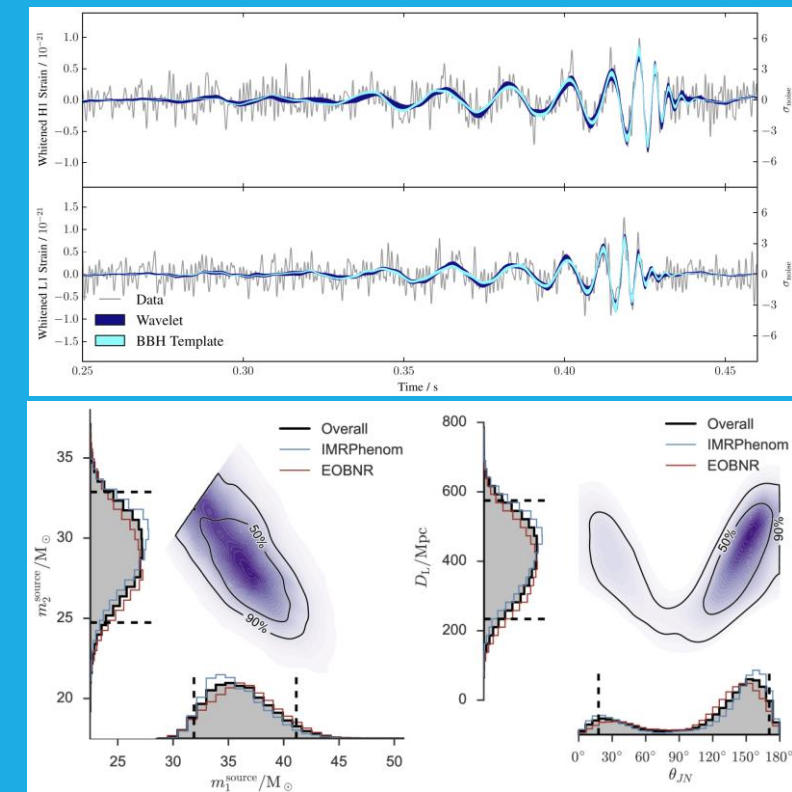
- Can include high statistics run

Bayes theorem:

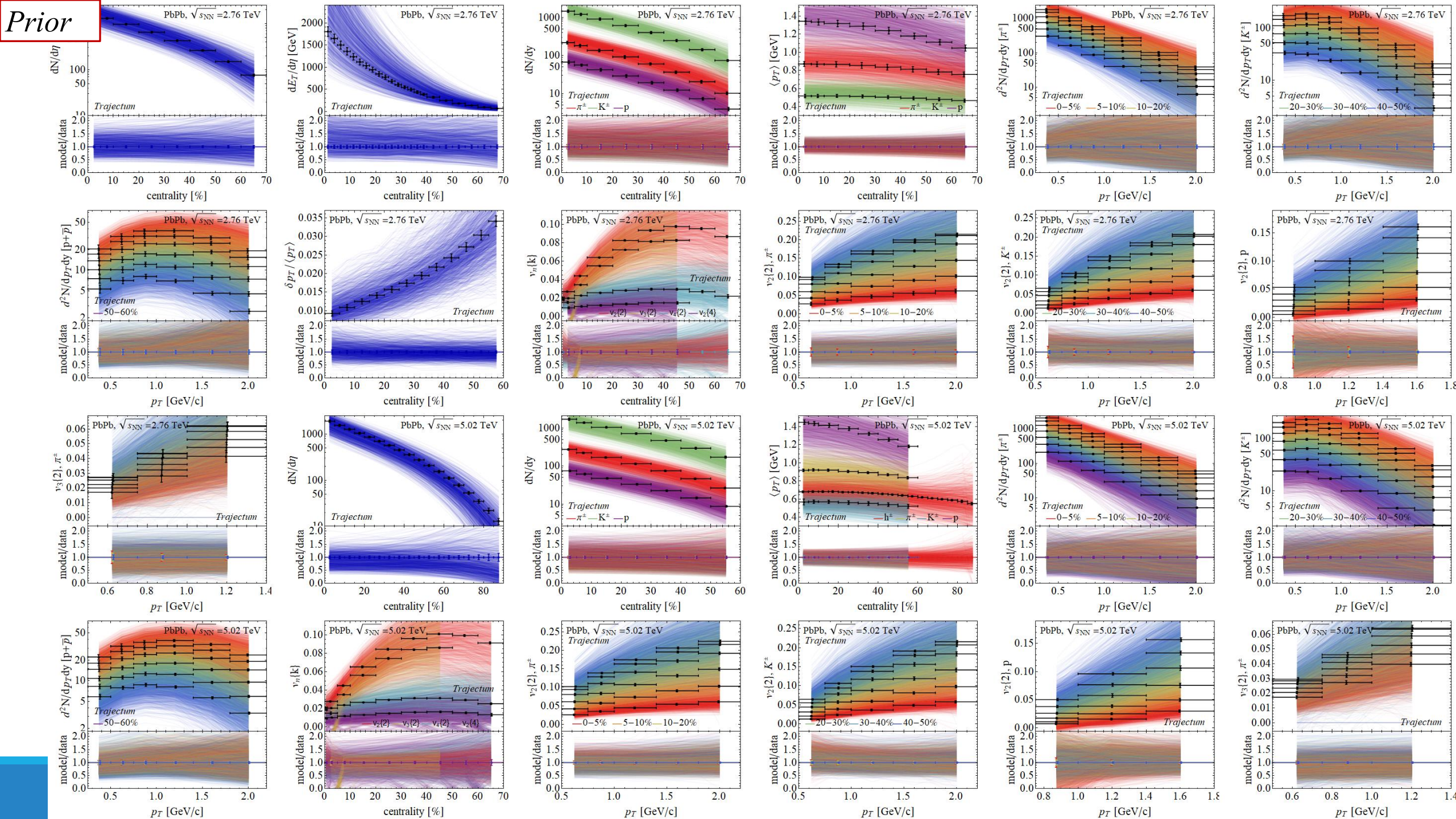
$$\mathcal{P}(\mathbf{x}|\mathbf{y}_{\text{exp}}) = \frac{e^{-\Delta^2/2}}{\sqrt{(2\pi)^n \det(\Sigma(\mathbf{x}))}} \mathcal{P}(\mathbf{x})$$

$$\text{with } \Delta^2 = (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}}) \cdot \Sigma(\mathbf{x})^{-1} \cdot (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}})$$

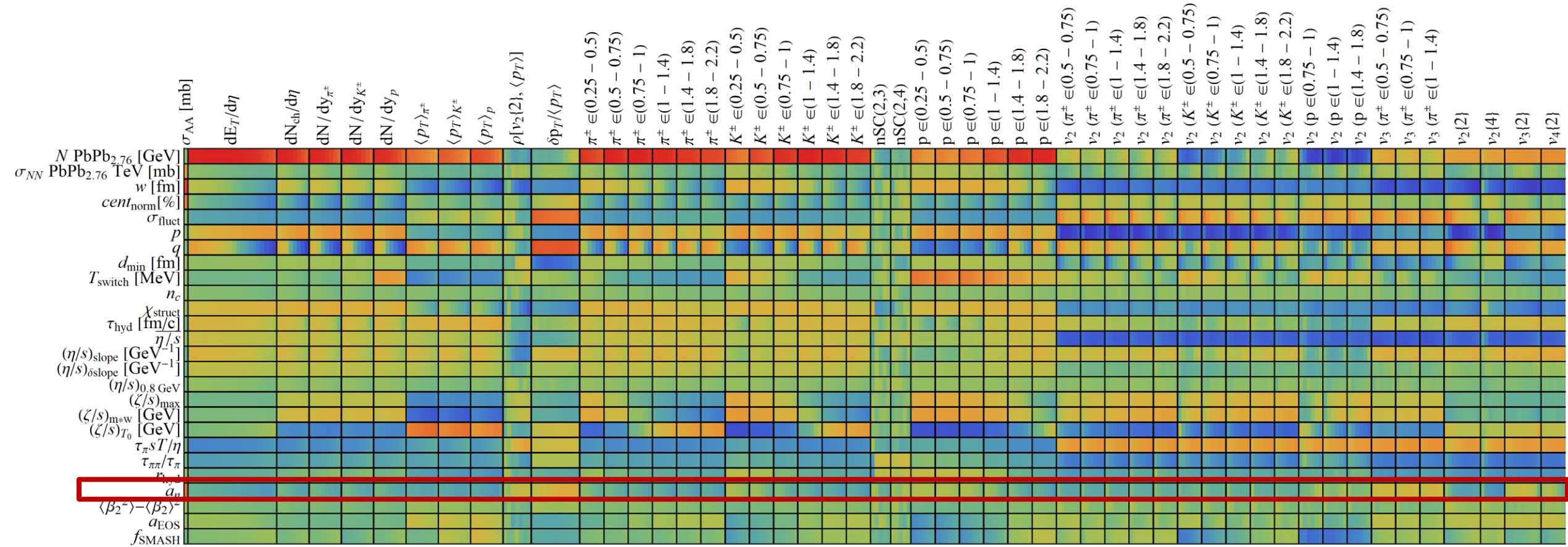
Same technique: gravitational waves



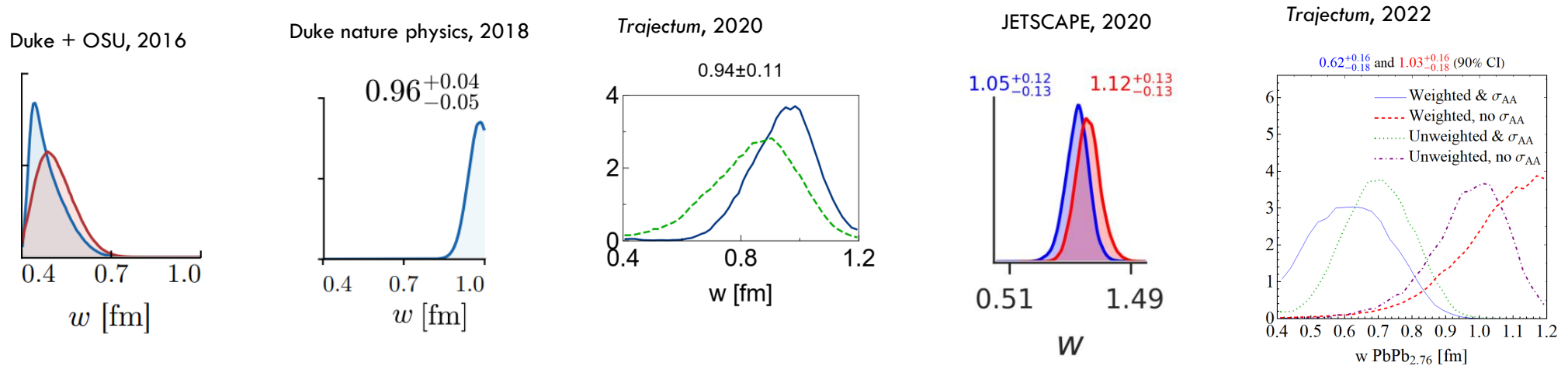
Prior



Design parameter-observable correlations:



Bayes theorem: three words of caution

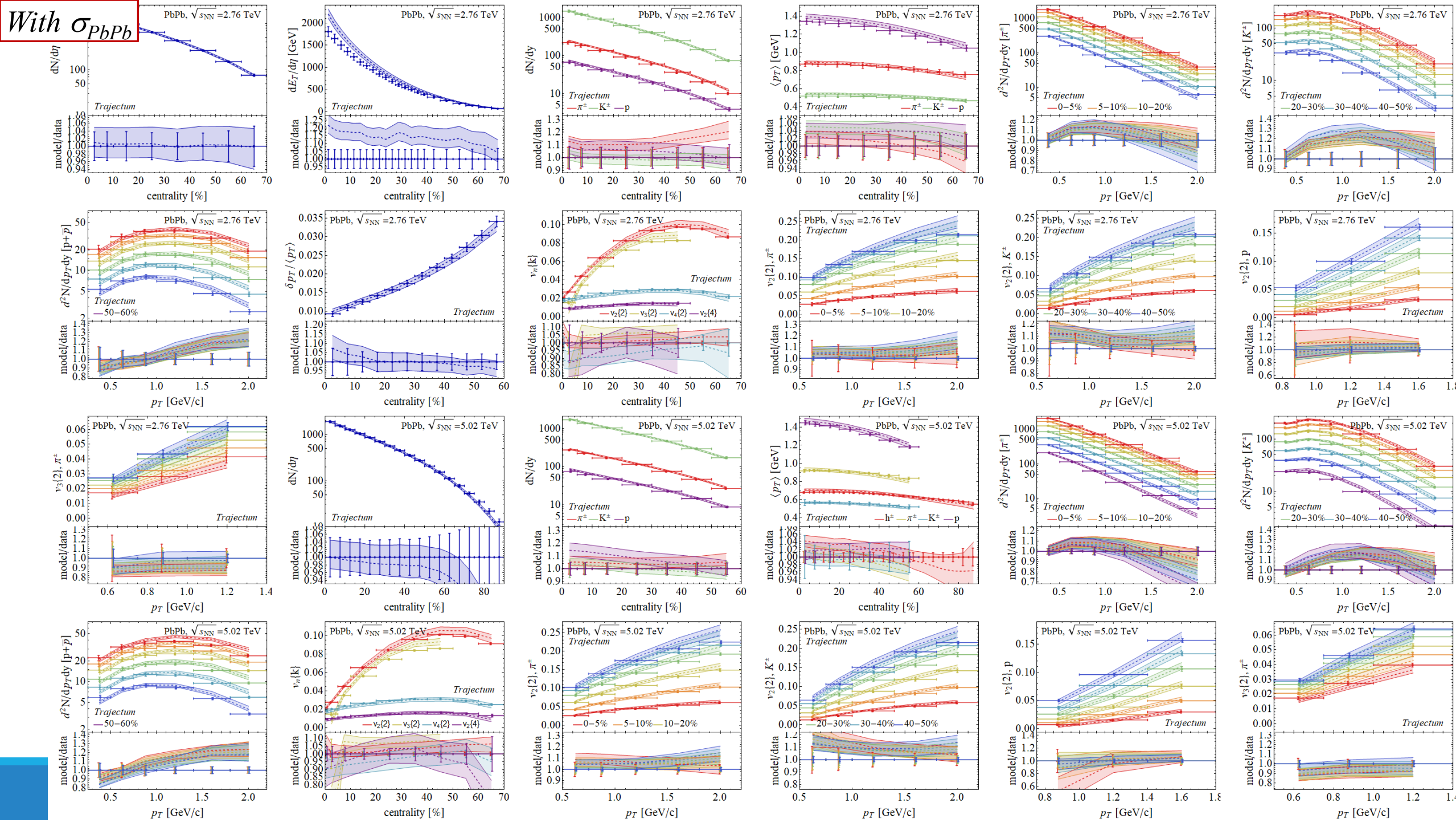


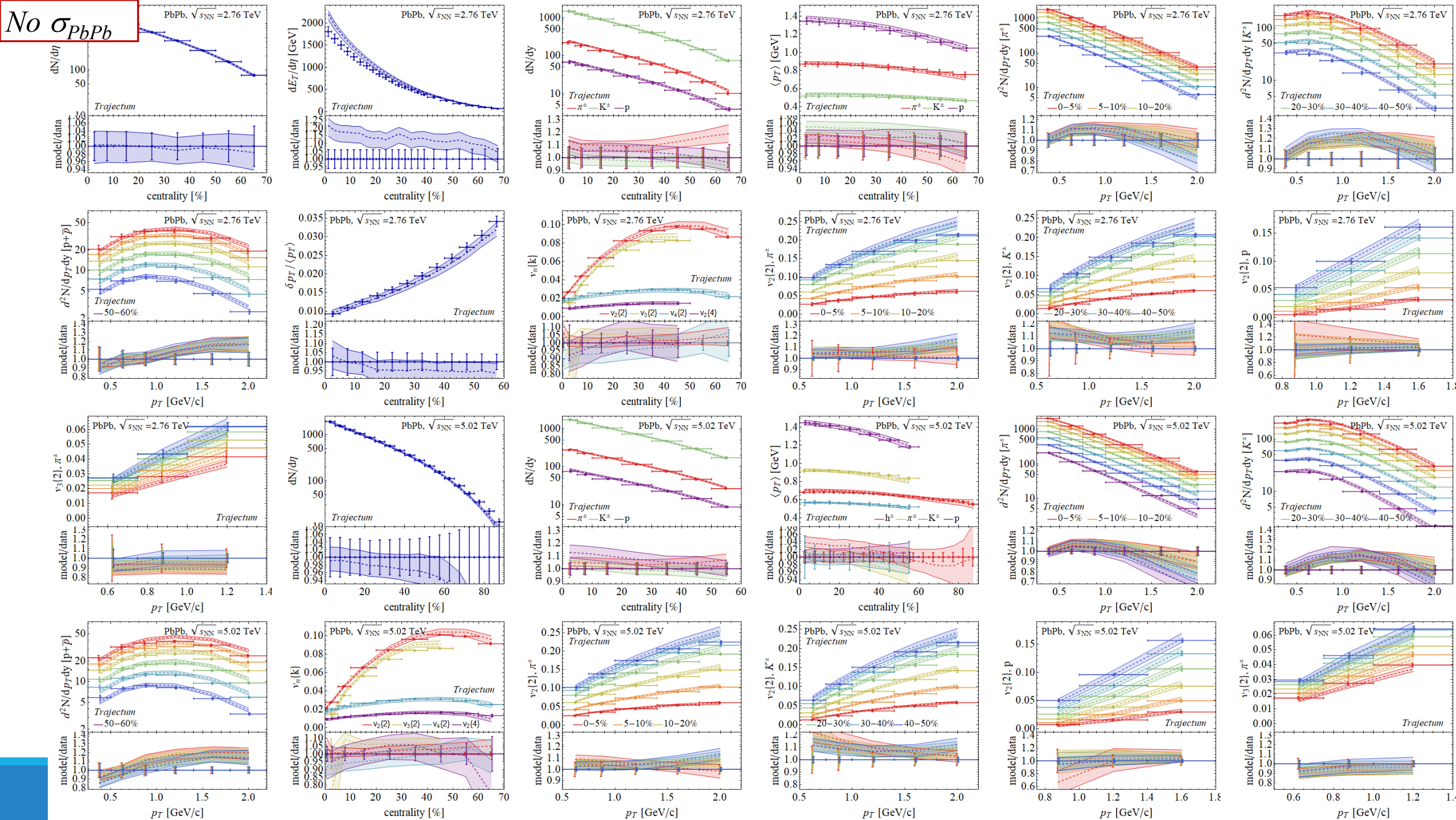
Nucleon width increased in 2018 (very significantly) and decreased (very significantly) in 2022

- Turns out that the width was inconsistent with the total hadronic PbPb cross section (measured in 2022)

What happened?

- In 2018: the initial stage model changed
- In 2022: new measurement, but also:
Many subtle changes in many data points 'pulled' to large width



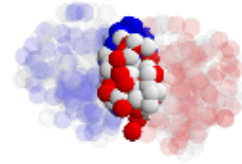


The nucleon width and the total PbPb hadronic cross section

What is easier to measure the width than by simply measuring the size?

Fix nucleon-nucleon cross section:

$$P_{\text{coll}} = 1 - \exp \left[-\sigma_{gg} \int dx dy \int dz \rho_A \int dz \rho_B \right]$$



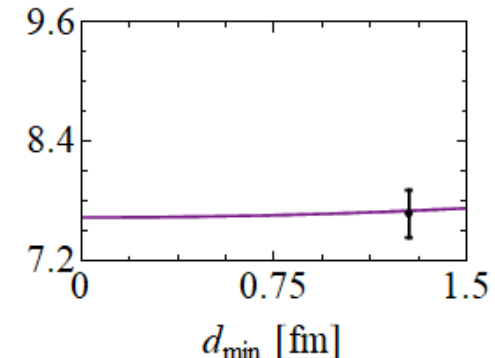
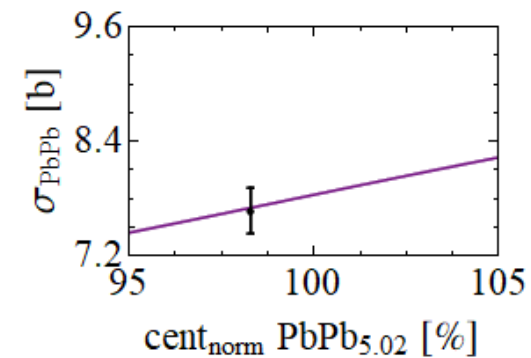
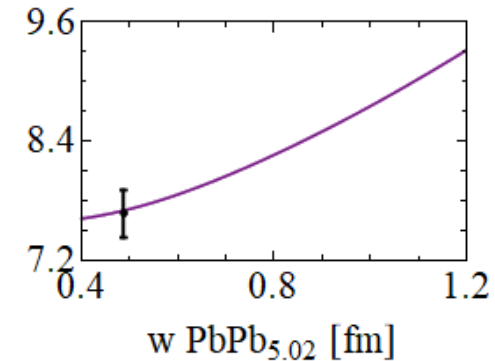
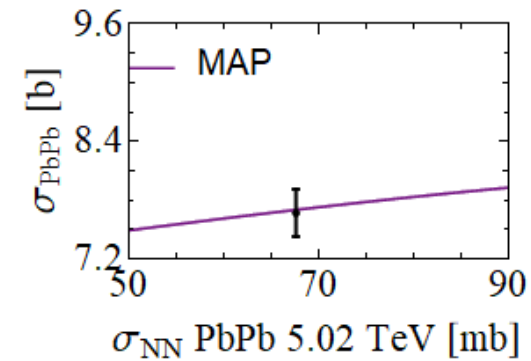
e.g. collision probability tuned to σ_{NN} for Gaussian profile ρ

Theoretically, cross section only depends on

- Nucleon-nucleon cross section
- Nucleon Gaussian width (dominant)
- Centrality normalisation
- Minimum inter-nucleon spacing

Makes the cross section a robust observable

- Basically implying every model needs to get this right
- Basically implying the nucleon width should be small



Why was the width overestimated?

	$\sigma_{\text{PbPb}}[b]$	$\sigma_{p\text{Pb}}[b]$
σ_{AA} & weights	8.03 ± 0.19	2.20 ± 0.06
weights	9.00 ± 0.34	2.50 ± 0.10
σ_{AA}	8.13 ± 0.19	2.23 ± 0.06
neither	8.72 ± 0.29	2.41 ± 0.09
ALICE	7.67 ± 0.24	2.06 ± 0.08

Without cross section width is large, about 1.0 fm

With the cross section width is smaller, about 0.7 fm

- Still tension with cross section: other data pushes width higher

Need to capture 'trust' in observables: **weighting**

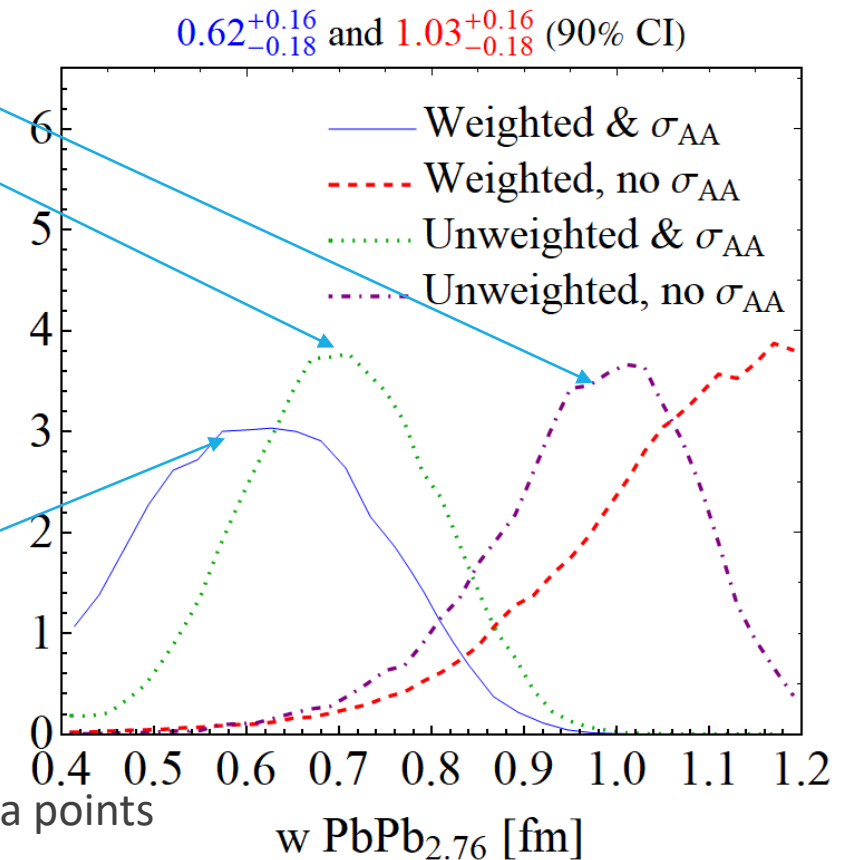
- Weight unity: cross section + integrated & unidentified
- Weight ½: integrated identified observables
- Weight ¼: p_{T} -differential identified observables
- Reduced weight: $p_{\text{T}} > 1.5$ GeV (π +K) and centrality > 50%

With weighting cross section comes out correctly

- Broader uncertainties: reflect less 'trust' due to weighting

Also: description of data not much worse with smaller width

- Important that Bayes factor is an addition of many (correlated!) data points



$$\mathcal{P}(\mathbf{x}|\mathbf{y}_{\text{exp}}) = \frac{e^{-\Delta^2/2}}{\sqrt{(2\pi)^n \det(\Sigma(\mathbf{x}))}} \mathcal{P}(\mathbf{x})$$

with $\Delta^2 = (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}}) \cdot \Sigma(\mathbf{x})^{-1} \cdot (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}})$

Four simple examples

Bayes theorem:

$$\mathcal{P}(\mathbf{x}|\mathbf{y}_{\text{exp}}) = \frac{e^{-\Delta^2/2}}{\sqrt{(2\pi)^n \det(\Sigma(\mathbf{x}))}} \mathcal{P}(\mathbf{x})$$

$$\text{with } \Delta^2 = (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}}) \cdot \Sigma(\mathbf{x})^{-1} \cdot (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}})$$

What's the likelihood?

1. For uncorrelated uncertainties?
 - 0.325
2. If 1 & 2 are fully correlated?
 - Ill defined / infinity
3. If 1 & 3 are fully correlated?
 - Ill defined / zero
4. For a realistic covariance matrix?
 - 1.439

$$\Sigma = \begin{pmatrix} 1 & 0.8 & 0.6 \\ 0.8 & 1 & 0.8 \\ 0.6 & 0.8 & 1 \end{pmatrix};$$

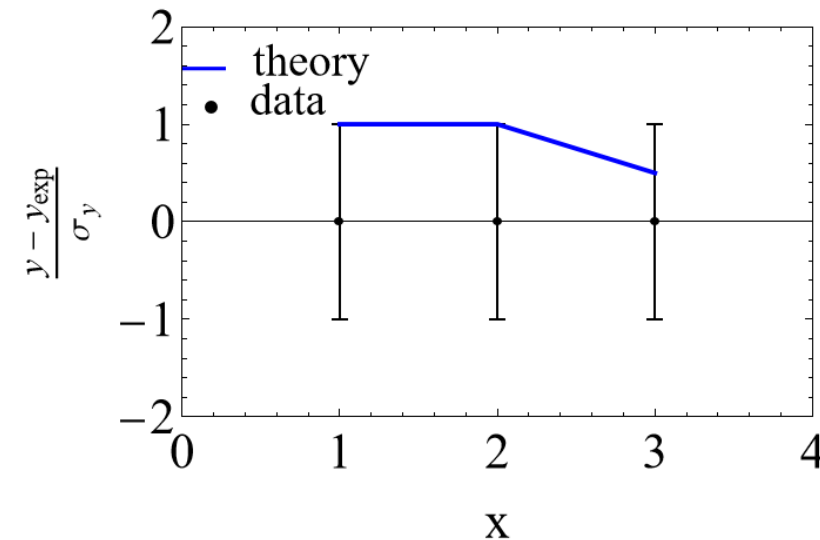
$$\delta\mathbf{y} = \{1, 1, 0.5\};$$

$$\Delta^2 = \delta\mathbf{y} \cdot \text{Inverse}[\Sigma] \cdot \delta\mathbf{y}$$

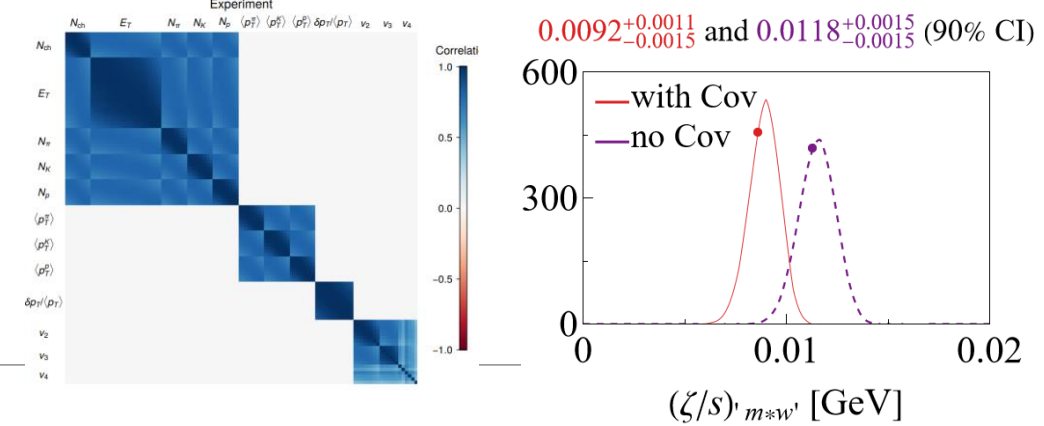
$$\text{likelihood} = e^{-\Delta^2/2} / \sqrt{\text{Det}[\Sigma]}$$

$$1.32813$$

$$1.43879$$



A real example: mean p_T



1. Most analyses likely follow the 'Nature Physics' approach*

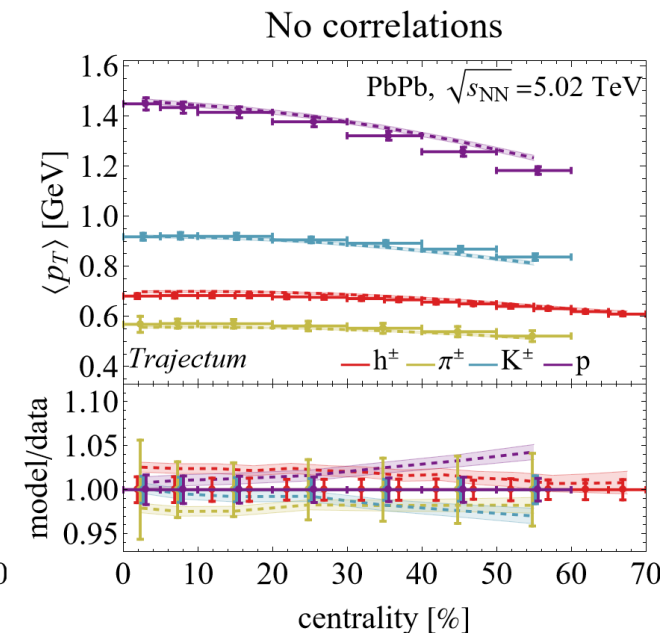
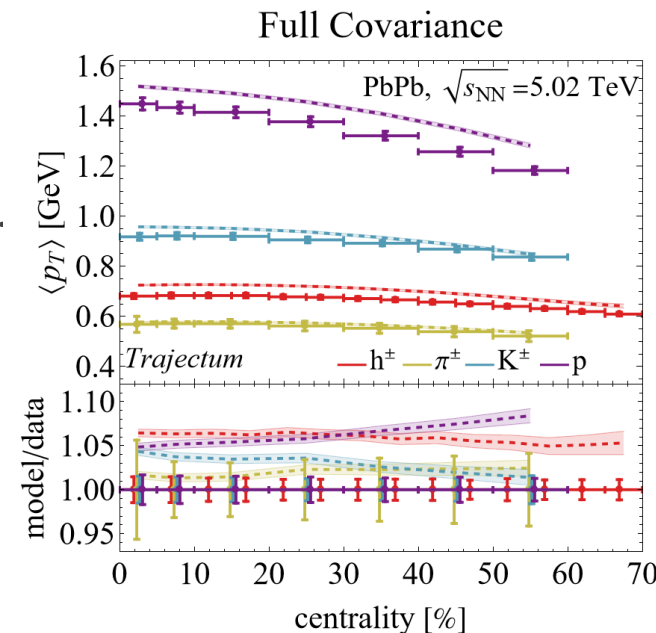
- Only correlations between equal observables
- If subobservables differ: multiply by 80%
- Covariance between two centralities:

$$e^{-\frac{1}{2}(c_1 - c_2)^2 / 2}$$

2. Likely not very realistic for many observables...

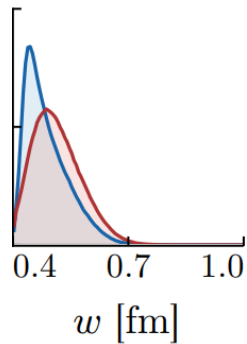
3. For e.g. mean p_T it really matters:

Changes bulk viscosity beyond uncertainty

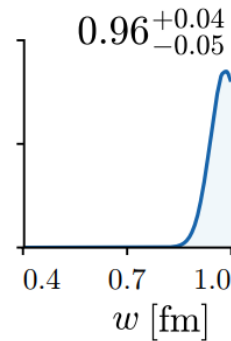


Bayes theorem: three words of caution

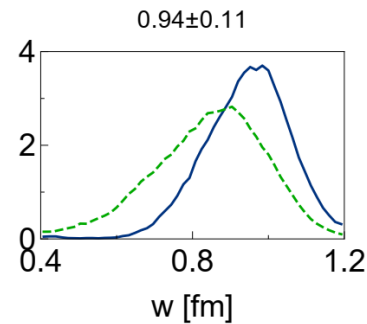
Duke + OSU, 2016



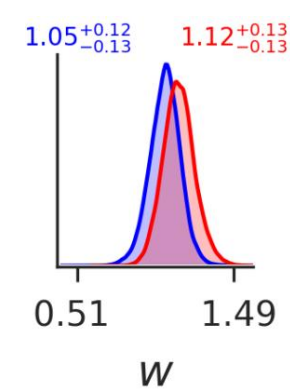
Duke nature physics, 2018



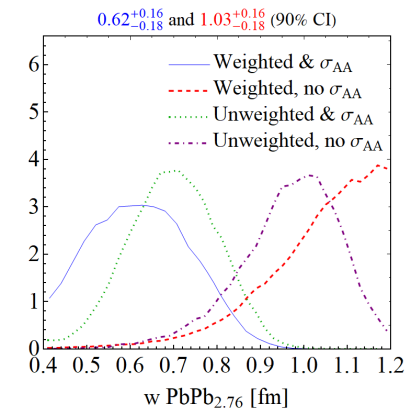
Trajectum, 2020



JETSCAPE, 2020



Trajectum, 2022



First lesson:

- Be careful with many data points that 'pull' the Bayes factor
- Are really all *known and unknown* uncertainties included in the covariance matrix? Including correlations?

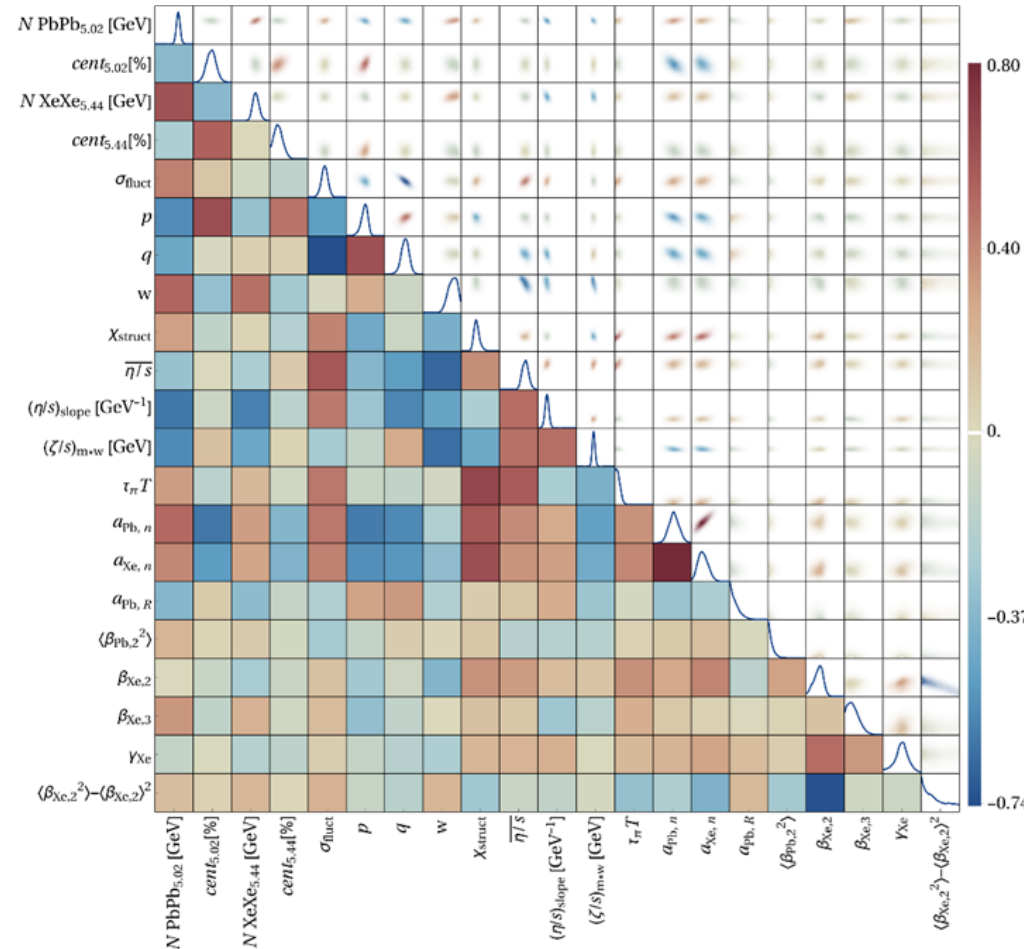
Bayes theorem: three words of caution

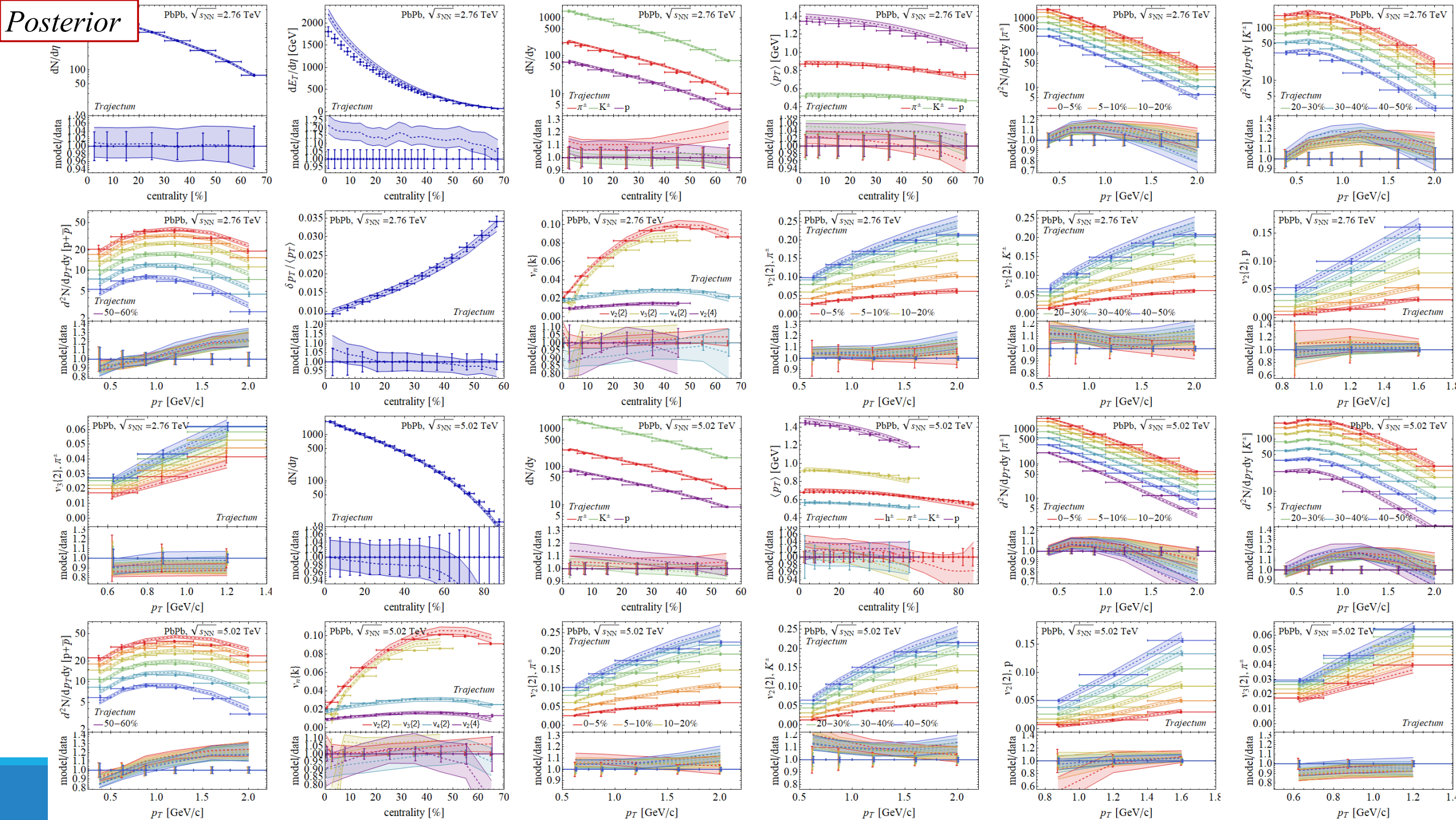
Second: does the model capture all the (relevant) physics?

- In practice always compromise between computational cost and completeness of the model

Hints can be seen in correlation matrix

- When 'fixing' one parameter this affects all other (correlated!) parameters
- 'Posterior' density unreliable if 'artificially' restricting a relevant parameter
- Meaning of 'relevant' depends on computation



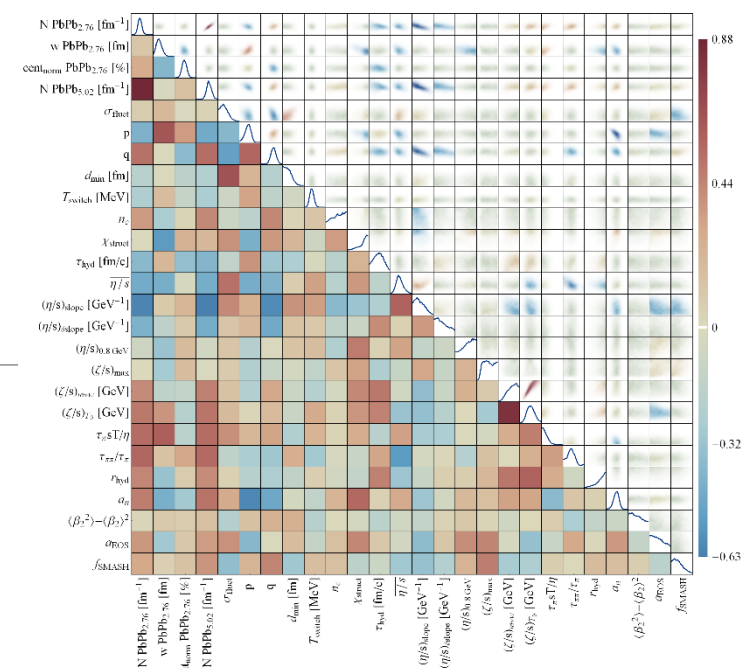
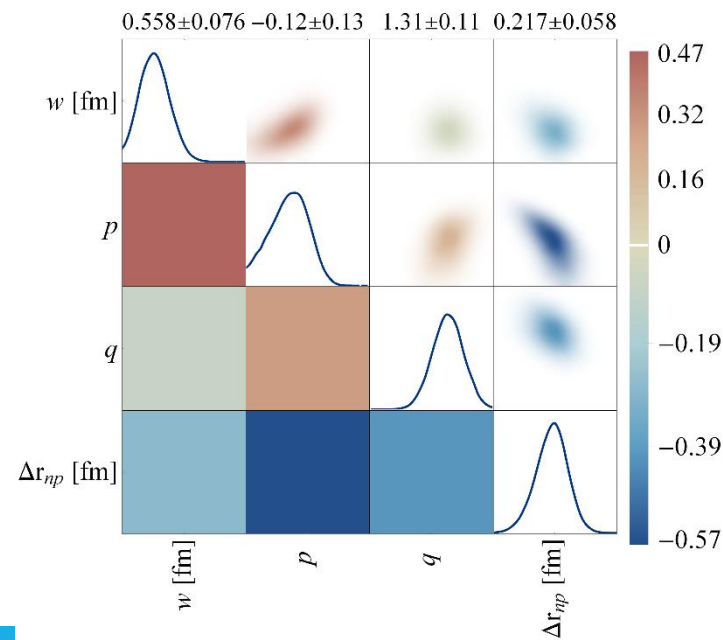
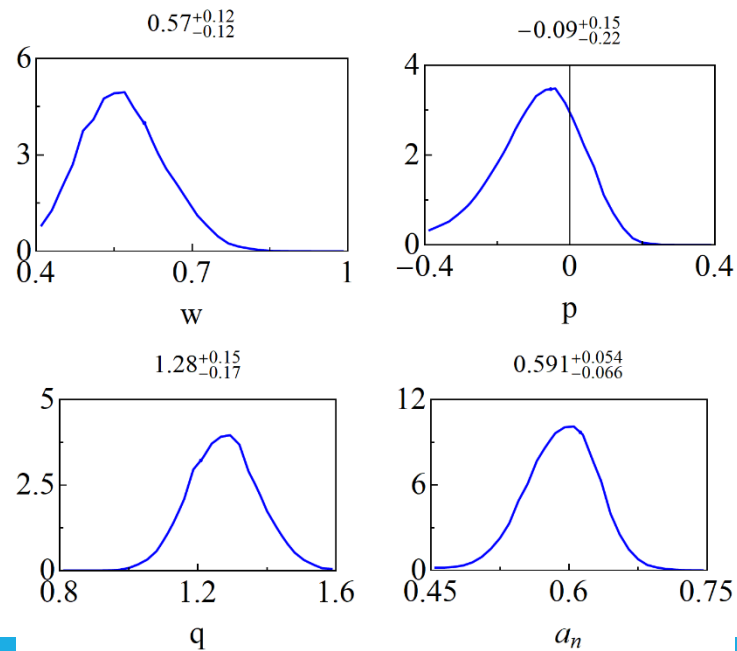


Trento: $\mathcal{E} \propto \left(\frac{1}{2}T_A^p + \frac{1}{2}T_B^p\right)^{q/p} = (T_A T_B)^{q/2} \big|_{p=0}$

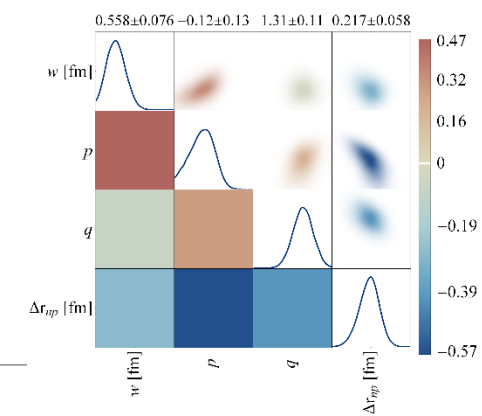
The neutron skin - posterior

1. Three parameters are most sensitive to the neutron skin:

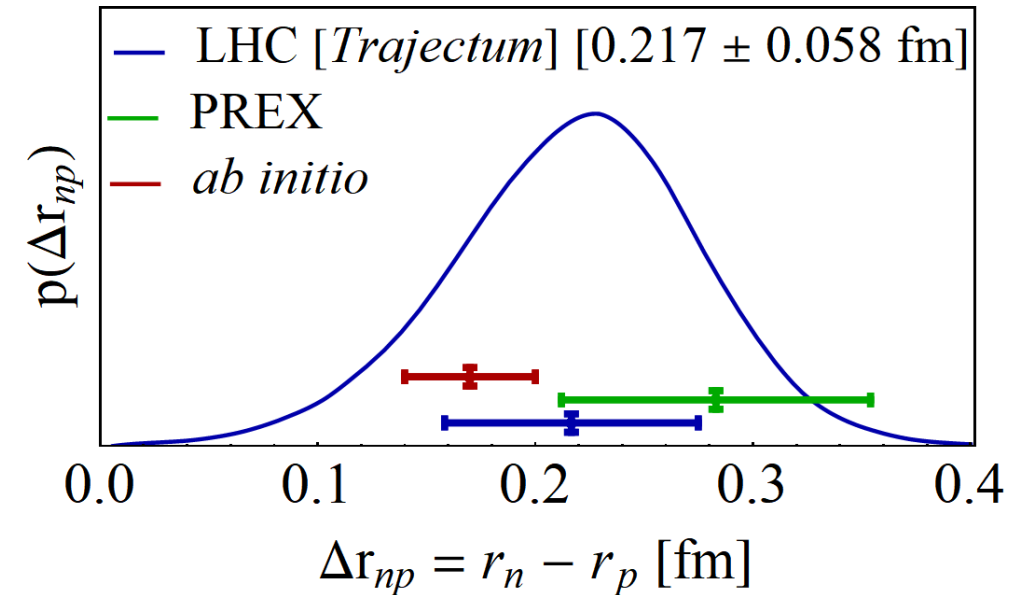
- The nucleon width and the Trento parameters p and q
- Small correlation with width (cross section is highly sensitive to w)
- Very strong anticorrelation with p ; centrality dependence is important



The neutron skin – final result



1. Transform to neutron radius minus proton radius
2. Final result consistent but smaller than PREX II
3. Uncertainty is about 20% smaller than PREX II
4. Cross section is crucially important, but also centrality dependence
 - Important to vary Trento parameters in particular

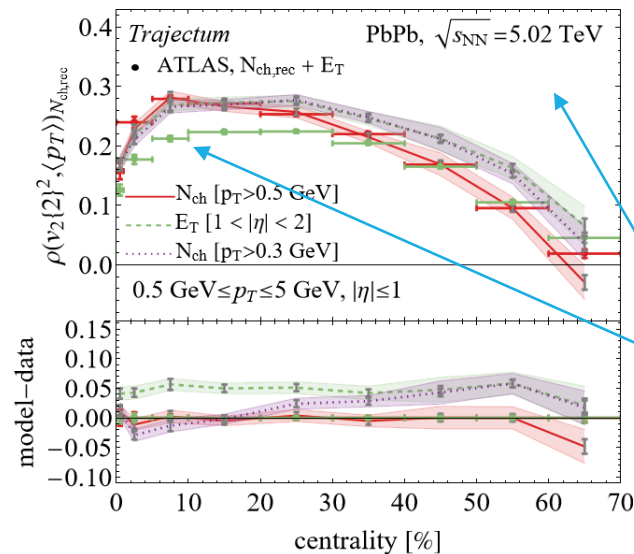


More recent progress: more precision

Slightly technical, message: with more precision a better (model) understanding is needed

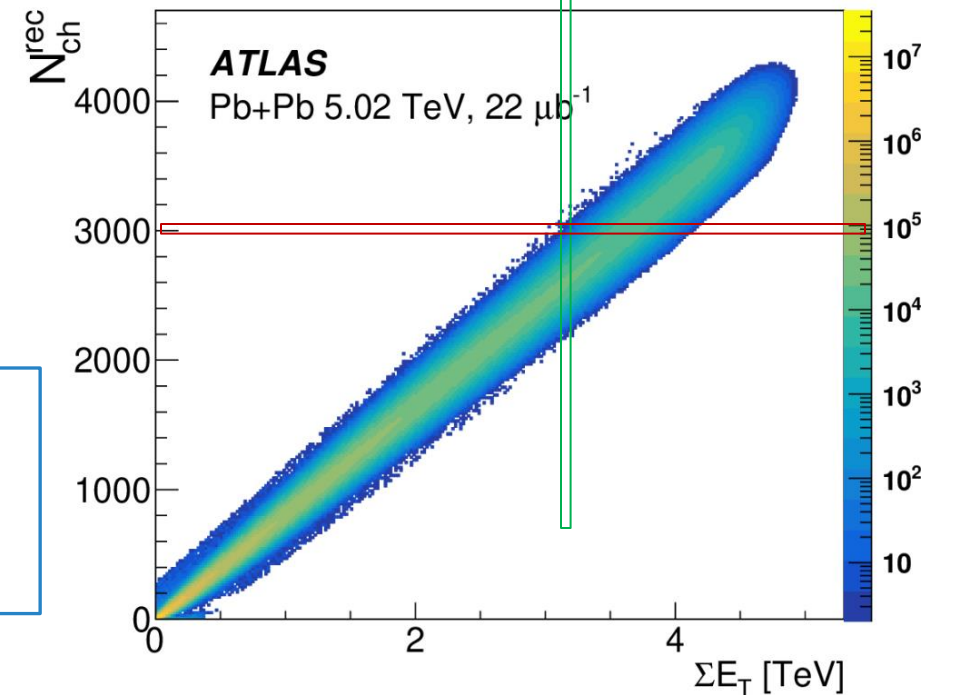
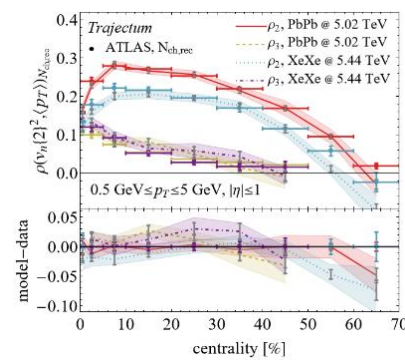
Forward (E_T) versus mid-rapidity (N_{ch})

- Forward considered 'better' (ATLAS, lower spread)
- Mid-rapidity closer to our model (boost invariant)



Somewhat ironic:
ATLAS sees big effect at central collisions
Trajectum has big effect at peripheral collisions

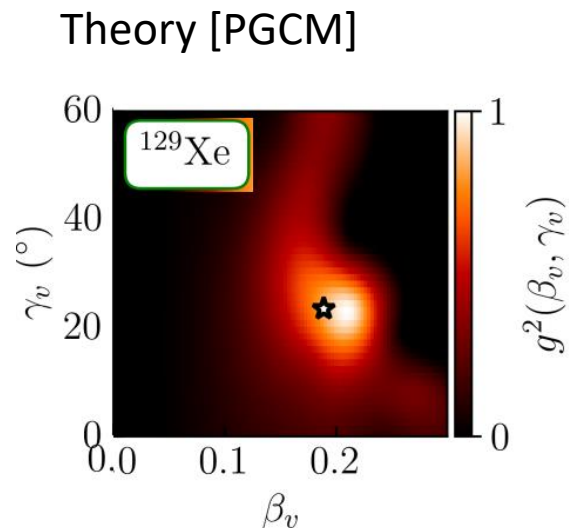
Trajectum: big effect p_T cut on N_{ch} (purple)



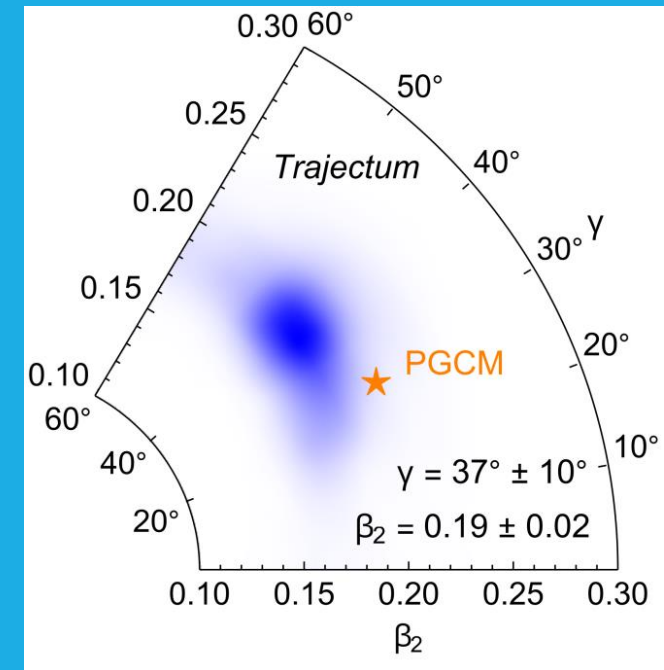
More recent progress: Xenon deformation

First quantitative measurement of Xenon triaxiality

Matches well with theoretical PGCM estimate



LHC [*Trajectum*]



“Bayesians address the questions everyone is interested in by using assumptions that no one believes. Frequentist use impeccable logic to deal with an issue that is of no interest to anyone.”- Louis Lyons

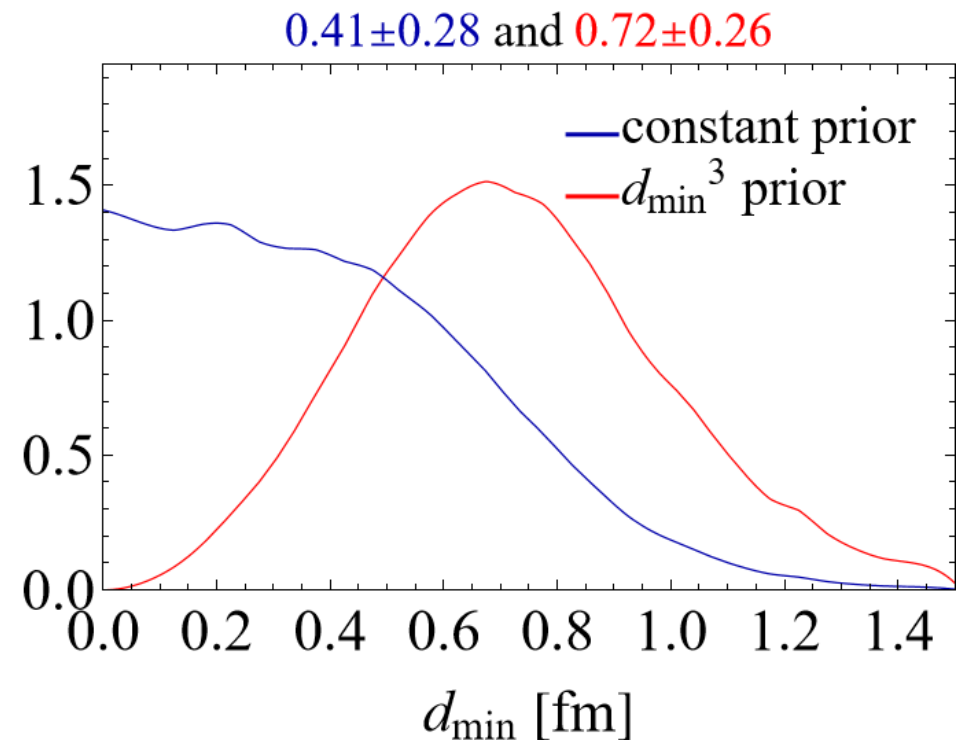
Bayes theorem: three words of caution

Third: the prior is important

- If the data is ‘strong’ the prior doesn’t matter
- In practice it often matters...

Bayesian analysis or global analysis

- The only way to ‘properly’ combine all the wealth of data and knowledge we have
- Be careful about the **known and unknown** uncertainties, including correlations



Discussion

1. Bayesian or global analyses

- Essential tool to assess parameters in view of experimental data
- Becoming more mainstream: public tools by e.g. JETSCAPE
- The uncertainties are essential and *sometimes difficult to control*

2. A personal biased overview of an answer to “Bayesian inference”

- Not covered: Technical implementation, non-flow effects, rapidity gaps, very peripheral or high pT, hadronic afterburner, collisions across beam energies and system sizes (light ions)

3. Heavy ions as a precision science

- Uncertainties are small: apples-to-apples comparison is essential
- Many ‘basic’ observables turn out to be interesting 😊 (no time to cover ultracentral..)

Back up / discussion

$$\mathcal{P}(\mathbf{x}|\mathbf{y}_{\text{exp}}) = \frac{e^{-\Delta^2/2}}{\sqrt{(2\pi)^n \det(\Sigma(\mathbf{x}))}} \mathcal{P}(\mathbf{x})$$

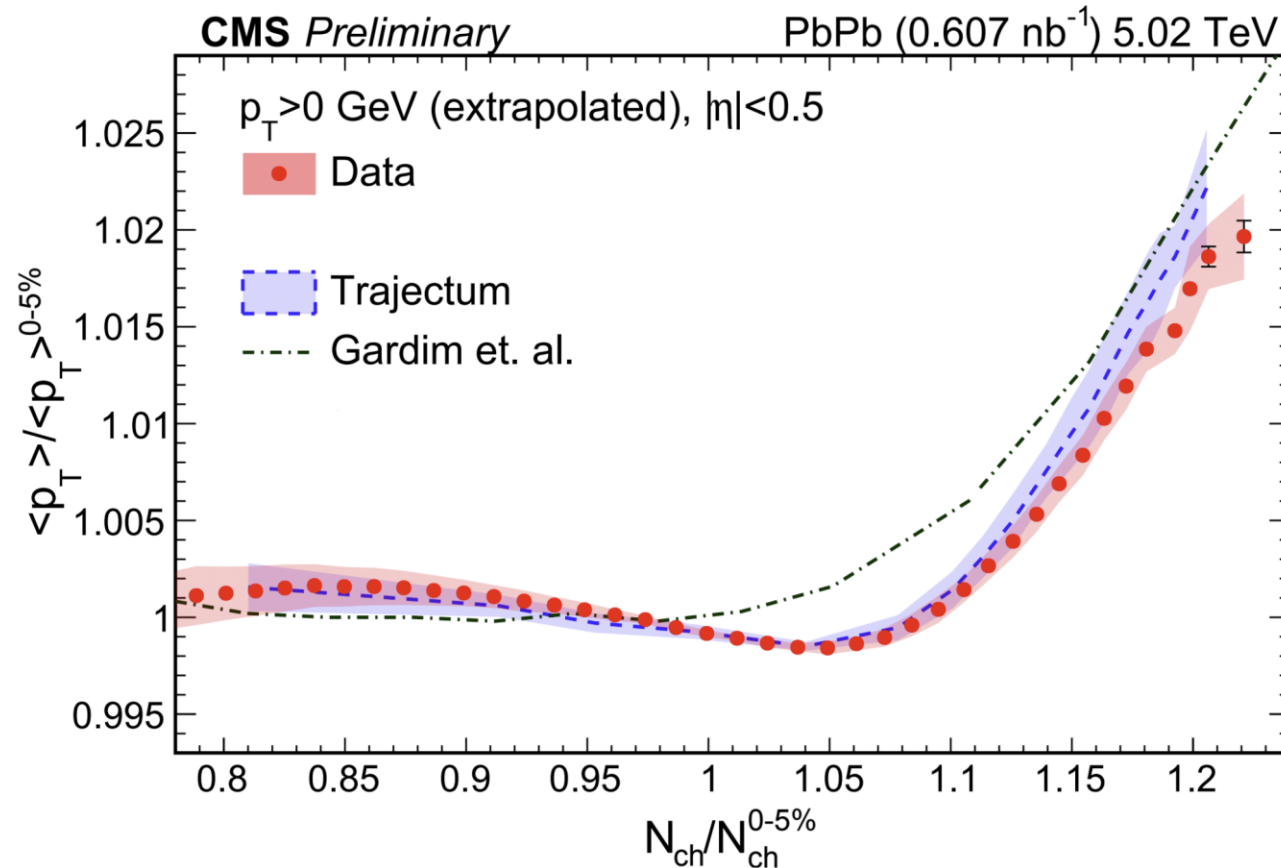
with $\Delta^2 = (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}}) \cdot \Sigma(\mathbf{x})^{-1} \cdot (\mathbf{y}(\mathbf{x}) - \mathbf{y}_{\text{exp}})$

Global analysis, technical practicalities

1. $P(\mathbf{x}|\mathbf{y}_{\text{exp}})$ can be a 25-dimensional function where $\mathbf{y}(\mathbf{x})$ is expensive to evaluate
 - Markov Chain Monte Carlo (emcee) can require order 10M evaluations
2. Technical steps to evaluate it
 - Compute $\mathbf{y}(\mathbf{x})$ for nicely spreaded sample of points (latin hypercube sampling, order 500 points)
 - PCA on observables, keep order 20 principal components (PCs)
 - Build a Gaussian Process Emulator for those PCs
 - Basically a glorified interpolating function in higher dimensions
 - 'Easier' if observables do not depend much on parameters (large length scale)
 - Validated and includes emulation uncertainty **(often dominant)**
3. Sample posterior of \mathbf{x} to obtain systematic model uncertainty **due to parameters**

Speed of Sound in QGP

- Slope of **data** matches models closely!



CMS PAS HIN-23-003

**Cesar
Bernardes's talk**
Wed. 15:40,
Ballroom C



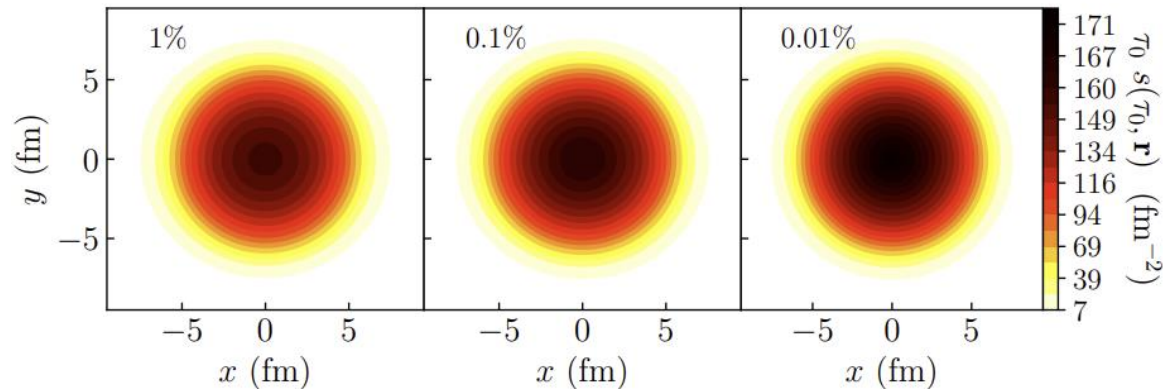
13

1. A true *prediction* using hydrodynamics (rare these days...)
 - Non-trivial curve; non-monotonicity was unexpected
 - Extremely precise (<0.1%) due to uncertainty *cancellations* in the ratio (theory & experiment (!))

What is happening?

1. (Extremely) ultracentral collisions are not driven by impact parameter

- Size does not extend beyond radius Lead nucleus
- (Local) **fluctuations** drive up multiplicity
- Average **temperature** can still increase



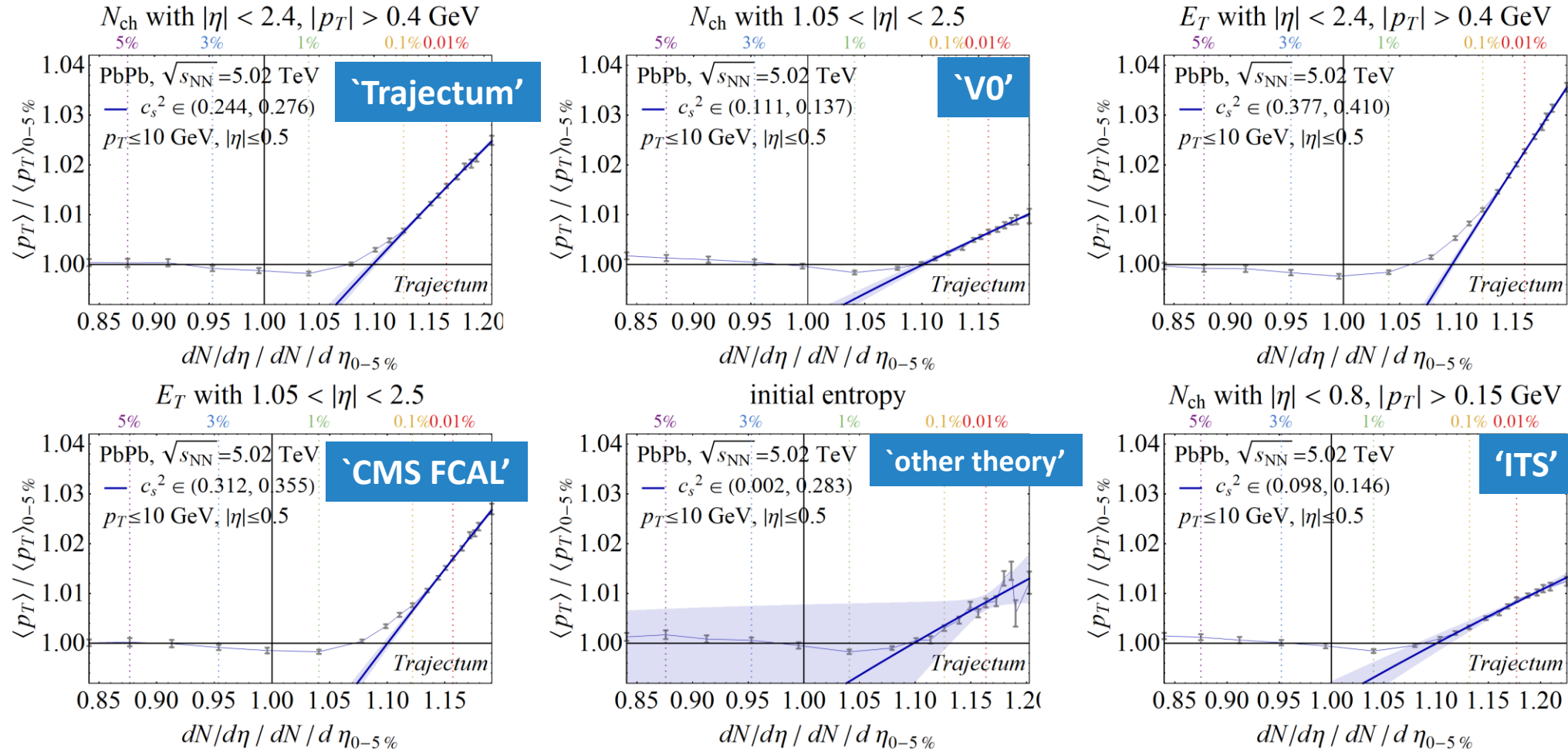
2. A simplified picture:

- Entropy increase \rightarrow multiplicity increases
- Temperature increases \rightarrow mean p_t increases (assuming ideal hydro + constant 'work' done in rapidity direction)

- Slope: sensitive to speed of sound (solely ?!):
$$c_s^2 = \frac{dP}{d\epsilon} = \frac{d \ln T}{d \ln s} = \frac{d \ln \langle p_t \rangle}{d \ln N_{\text{ch}}}$$

What then determines the slope?

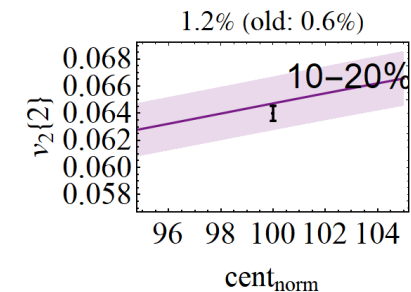
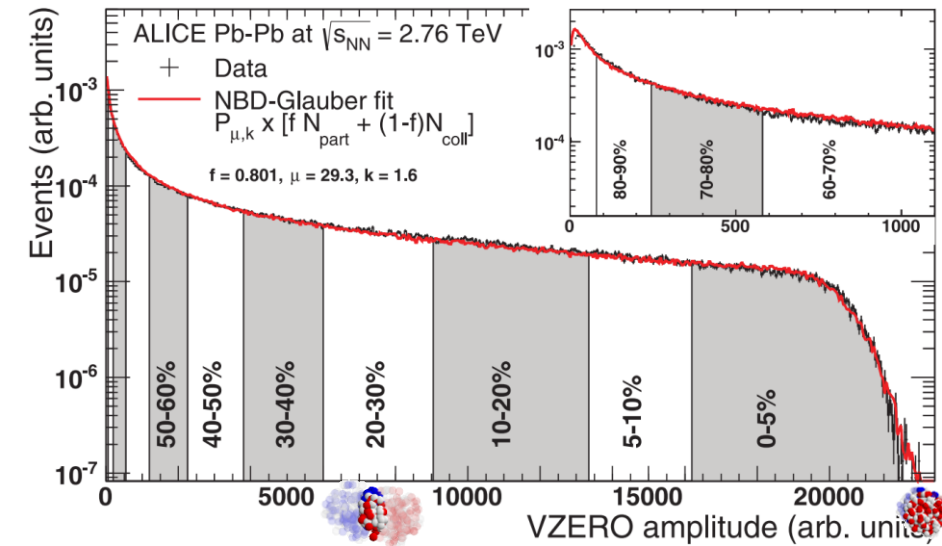
Centrality cuts are very important.;
boost invariant, self-correlation, computationally expensive for $|\eta| > 1.5$



An example analysing data: centrality

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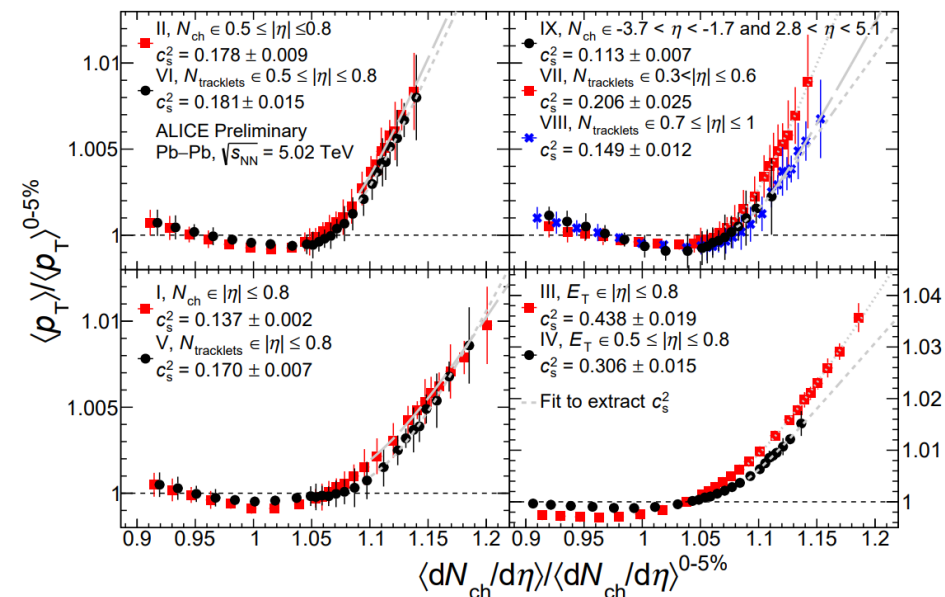
1. Historically: *centrality meant impact parameter*
 - Impact parameter is however never observable
 - Traditionally V0 signal (multiplicity at forward rapidity) as a proxy
 - Especially with small systems (p-Pb) the definition matters
2. Modern point of view: *centrality is event activity*
 - But where?
 - Also, 90-100% cannot be measured but is modelled
Do we have the anchor point at 90% under control?
 - If not, all observables shift as a function of centrality (!)
 - 'the ultimate' correlation of observables
 - Related to total hadronic PbPb cross section



An example analysing data: centrality

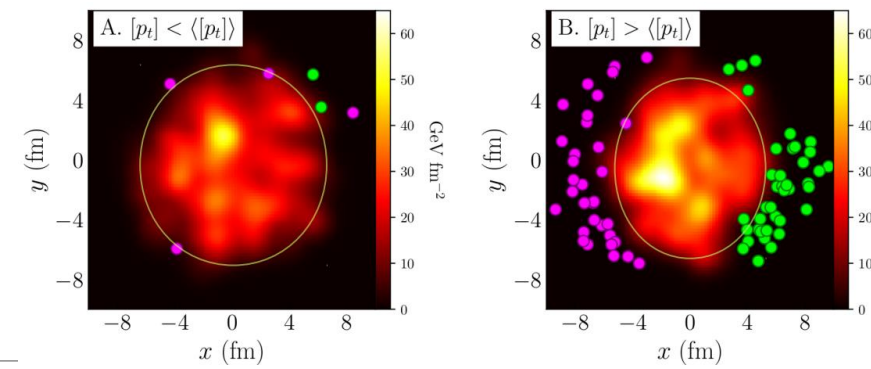
1. An even more modern perspective: apples-to-apples comparison theory/experiment
 - Even relevant in PbPb collisions; for instance in ultracentral:

Observable	Label	Centrality estimation	$\langle p_T \rangle$ and $\langle dN_{ch}/d\eta \rangle$	Minimum $ \Delta\eta $
N_{ch} in TPC	I	$ \eta \leq 0.8$	$ \eta \leq 0.8$	0
	II	$0.5 \leq \eta < 0.8$	$ \eta \leq 0.3$	0.2
E_T in TPC	III	$ \eta \leq 0.8$	$ \eta \leq 0.8$	0
	IV	$0.5 \leq \eta < 0.8$	$ \eta \leq 0.3$	0.2
$N_{tracklets}$ in SPD	V	$ \eta \leq 0.8$	$ \eta \leq 0.8$	0
	VI	$0.5 \leq \eta < 0.8$	$ \eta \leq 0.3$	0.2
	VII	$0.3 < \eta < 0.6$	$ \eta \leq 0.3$	0
	VIII	$0.7 \leq \eta < 1$	$ \eta \leq 0.3$	0.4
N_{ch} in V0	IX	$-3.7 < \eta < -1.7$ and $2.8 < \eta < 5.1$	$ \eta \leq 0.8$	0.9



2. Apples-to-apples only possible if theoretical modelling is valid at the centrality definition
 - Can be problematic especially for boost invariant models (forward rapidity poorly described)

Centrality selection: the ρ_2 correlator

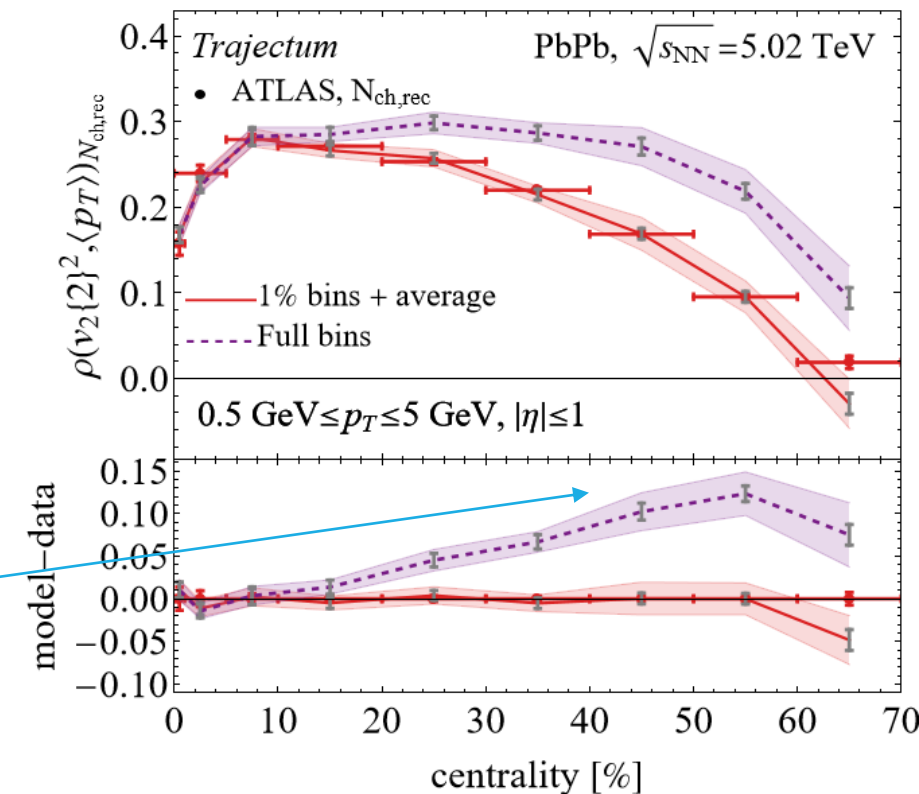


$\rho(v_n\{2\}^2, \langle p_T \rangle)$: correlation anisotropic flow and transverse momentum

- True correlator: zero if flow and momentum are independent
- Intuition: smaller system \rightarrow larger p_T
- Intuition: more elliptic \rightarrow higher $v_2\{2\}$
- Intuition: central: smaller system often a bit more elliptical

Centrality is an important detail:

- The correlator is done in small centrality bins (1% for us)
- Obtain larger bins by averaging
- Relatively big effect (up to 0.1, exp uncert. is 0.01)

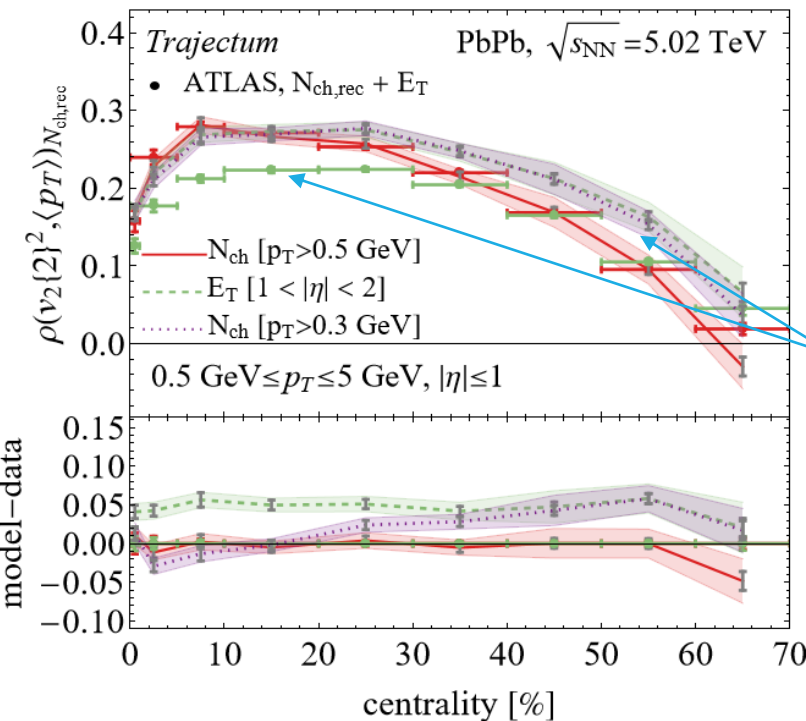


But how do we define centrality?

Centrality selection: the ρ_2 correlator

Forward (E_T) versus mid-rapidity (N_{ch}): apples-to-apples

- Forward considered 'better' (ATLAS, lower spread)
- Mid-rapidity closer to model (boost invariant)



Somewhat ironic:

ATLAS sees big effect at central collisions
Trajectum has big effect at peripheral collisions

Trajectum: big effect p_T cut on N_{ch} (purple)

