# Bootstrap: a nuclear physics point of view

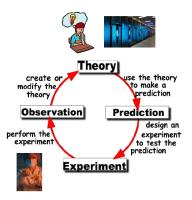
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## Introduction



#### The scientific method

- A problem is identified
- Relevant data are gathered

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- An hypothesis is formulated from the data
- The hypothesis is empirically tested

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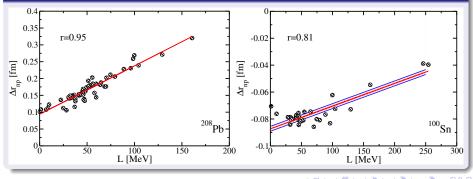
Enhancing the interaction between nuclear experiment and theory through information and statistics (ISNET) (2015)

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# What is Bootstrap?

- Non-Parametric Bootstrap is a statistical method introduced by Efron in 1979 B. Efron and R. Tibshirani, An introduction to the bootstrap (CRC press, 1994).
- Based on resampling (with replacement) of data
- No specific assumptions on the *parent* distribution
- Possible extension: parametric Bootsrap





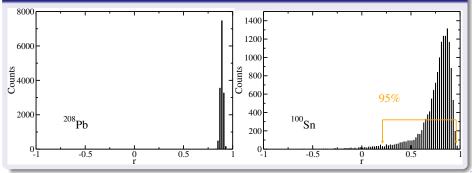
### Bootstrap @ work!

• We build BB new data sets with n points (as original) by resampling

$$\left(\begin{array}{c}2n-1\\n\end{array}\right) = \frac{(2n-1)!}{n!(n-1)!}$$

- We calculate for each data-set  $\hat{r}$
- We build the distribution

#### Errors on $\hat{r}$ : confidence interval 68% quantile.



# Bootsrap fit

- Possible alternative ot  $\chi^2$
- Bootsrap of residuals  $y = f(\vec{a}, x) + \varepsilon$  (4000 samples)
- Creation of new data sets  $y^* = y + \varepsilon^*$
- $\bullet~$  Minimisation of a distance  $(y^*-f(\vec{a},x))^2$  or  $|y^*-f(\vec{a},x)|$

#### Example: Liquid-Drop model

We want to fit a linear model (as a start point!)

$$f(N,Z;av,as,ac,aa) = av - as(N+Z)^{-1/3} - ac\frac{Z^2 - Z}{(N+Z)^{4/3}} - aa\frac{(N-Z)^2}{(N+Z)^2}$$

We use 2236 *experimental* points. No error on the data (as a start)

#### MATHEMATICA results

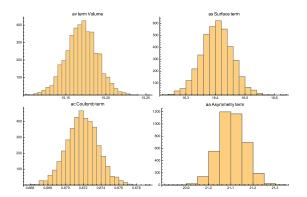
	Estimate	Standard Error	t - Statistic	P- Value	
av	15.1697	0.0197484	768.15	1.35875104398	×10 <sup>-2806</sup>
as	16.4042	0.0525803	311.985	4.85895485094	×10 <sup>-1904</sup>
ac	0.671887	0.00180058	373.151	8.83790086779	×10 <sup>-2082</sup>
aa	21.1048	0.0608933	346,586	2.61381401095	×10 <sup>-2008</sup>

We also obtain the covariance matrix

$$\left(\begin{array}{ccccc} av & as & ac & aa \\ 1 & 0.981 & 0.970 & 0.654 \\ & 1 & 0.921 & 0.590 \\ & & 1 & 0.588 \\ & & & 1 \end{array}\right)$$

We use these results to benchmark the Bootstrap. We fix the residual  $\varepsilon$  for the best fit.

## Bootsrap to assess errors (Gaussian here)



#### Comparing values

The mean values are av = 15.1699; as = 16.404; ac = 0.672; aa = 21.104.

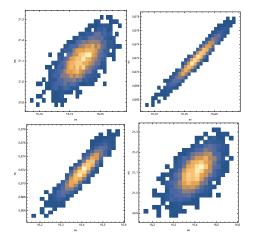
The differences with MATHEMATICA:  $\delta \approx 0.0003/4$ 

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### Bootsrap to assess correlations



#### Bootsrap distribution

We have access to the distribution of the parameters around the minimum!

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Bootstrap

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# Conclusions

Bootsrap is a very simple tool

- Simple to implement
- We have used at moment only on error analysis
- Need more CPU than  $\chi^2$ , but more informations

#### Possible ideas

- Bootsrap to go beyond parabolic approximations
- $\bullet\,$  Tool to avoid derivative perform numerical derivatives of  $\chi^2$
- Possible to assess non-Gaussian errors

#### THANK YOU!!!