



Update on the improved lattice calculation of direct CP-violation in K decays

Christopher Kelly & Tianle Wang
(RBC & UKQCD collaborations)

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The RBC & UKQCD collaborations

BNL and BNL/RBRC

Yasumichi Aoki (KEK)
Taku Izubuchi
Yong-Chull Jang
Chulwoo Jung
Meifeng Lin
Aaron Meyer
Hiroshi Ohki
Shigemi Ohta (KEK)
Amarjit Soni

UC Boulder

Oliver Witzel

CERN

Mattia Bruno

Columbia University

Ryan Abbot
Norman Christ
Duo Guo
Christopher Kelly
Bob Mawhinney
Masaaki Tomii
Jiqun Tu

Bigeng Wang
Tianle Wang
Yidi Zhao

University of Connecticut

Tom Blum
Dan Hoyal (BNL)
Luchang Jin (RBRC)
Cheng Tu

Edinburgh University

Peter Boyle
Luigi Del Debbio
Felix Erben
Vera Gülpers
Tadeusz Janowski
Julia Kettle
Michael Marshall
Fionn Ó hÓgáin
Antonin Portelli
Tobias Tsang
Andrew Yong
Azusa Yamaguchi

UAM Madrid

Julien Frison

University of Liverpool

Nicolas Garron

MIT

David Murphy

Peking University

Xu Feng

University of Regensburg

Christoph Lehner (BNL)

University of Southampton

Nils Asmussen
Jonathan Flynn
Ryan Hill
Andreas Jüttner
James Richings
Chris Sachrajda

Stony Brook University

Jun-Sik Yoo
Sergey Syritsyn (RBRC)

Motivation

- Significant goal of RBC & UKQCD is physical lattice calculation of ε' , the measure of direct CP-violation in $K \rightarrow \pi\pi$ decays.
- Direct CPV heavily suppressed in Standard Model hence ε' is very sensitive to BSM sources of CPV.
- When compared to experiment, any deviations may help shed light on origin of matter/antimatter asymmetry.
- Understanding $I=0$ $\pi\pi$ system is crucial:
 - Energy is needed for time dependence of correlation function from which extract finite-volume $K \rightarrow \pi\pi$ matrix element.
 - Phase shift enters Lellouch-Lüscher finite-volume correction to matrix element.

Calculation details

- 741 configurations of domain wall fermions with Iwasaki+DSDR gauge action ($a^{-1}=1.73$ GeV) and physical pion masses.
- $32^3 \times 64$ volume $\sim (4.7 \text{ fm})^3$ - large volume helps control finite-volume effects.
- G-parity BCs in 3-directions, lowest energy pion momentum $(1,1,1)\pi/L$. $E_{\pi\pi} \sim m_K$ ensuring physical decay.
- 3 $\pi\pi$ operators: $\pi\pi(111)$, $\pi\pi(311)$, σ (scalar)
- Significant improvement in determination of ground-state energy (cf Tianle Wang's talk)
- $E_{\pi\pi} = 0.3614(74)$ [2015, 216cfigs, 1 op] [Phys.Rev.Lett. 115 (2015) no.21, 212001]
 \longrightarrow $\sim 0.3488(9)^*$ [2019, 741 cfigs, 3 op]
- Result now agrees well with dispersive prediction.
- Good evidence that excited state contamination formerly underestimated.
- Effect on $K \rightarrow \pi\pi$ as yet undetermined, awaiting final $\pi\pi$ numbers

Multi-operator $\pi\pi$ fits

- Obtain parameters by simultaneous fitting to matrix of correlation functions

$$C_{ij}(t) = \langle 0 | O_i^\dagger(t) O_j(0) | 0 \rangle = C + \sum_{\alpha} A_{i,\alpha} A_{j,\alpha} e^{-E_{\alpha} t}$$

round-the-world single pion propagation
small compared to errors - drop

- Correlated fits to obtain most precise result and **goodness-of-fit metric**
- Jackknife procedure to determine statistical errors.
- Full variation in covariance matrix over jackknife samples taken into account with double-jackknife technique.

Outstanding issues

- $I=0$ $\pi\pi$ calculation looks very solid and excited state contamination appears negligible.
- Why the delay in publication then?
 - With vastly more statistics, now some evidence of autocorrelation effects which may lead to underestimation of errors.
 - Despite high degree of stability under changing fit ranges, goodness of fit typically quite poor and display unexpected dependence on bin size and fit ranges.
- Importance of reliable $\pi\pi$ fits strongly motivates resolving these issues.

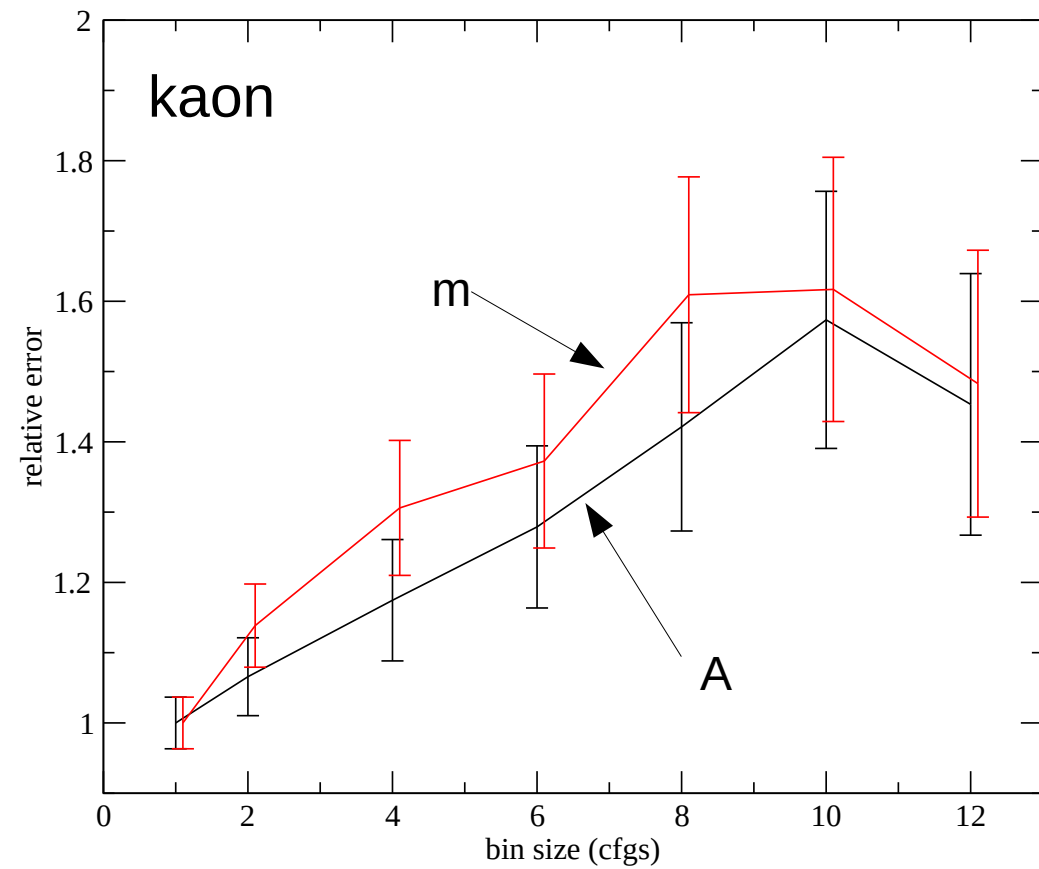
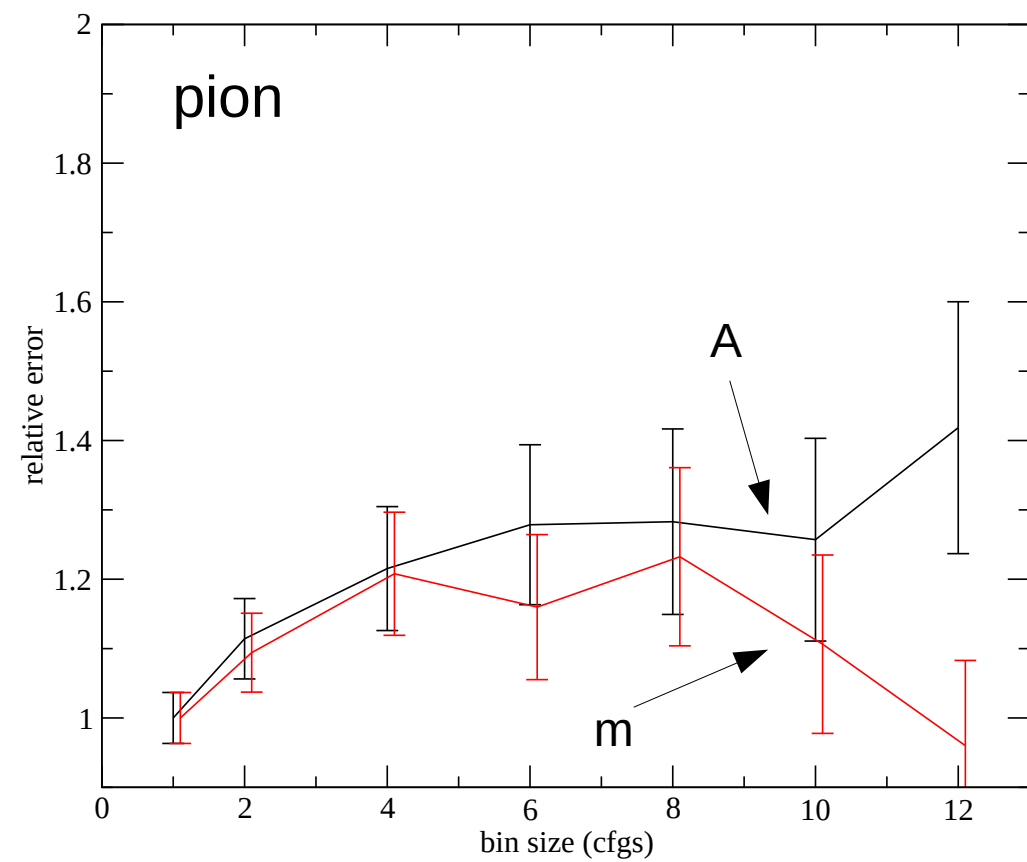


Autocorrelations

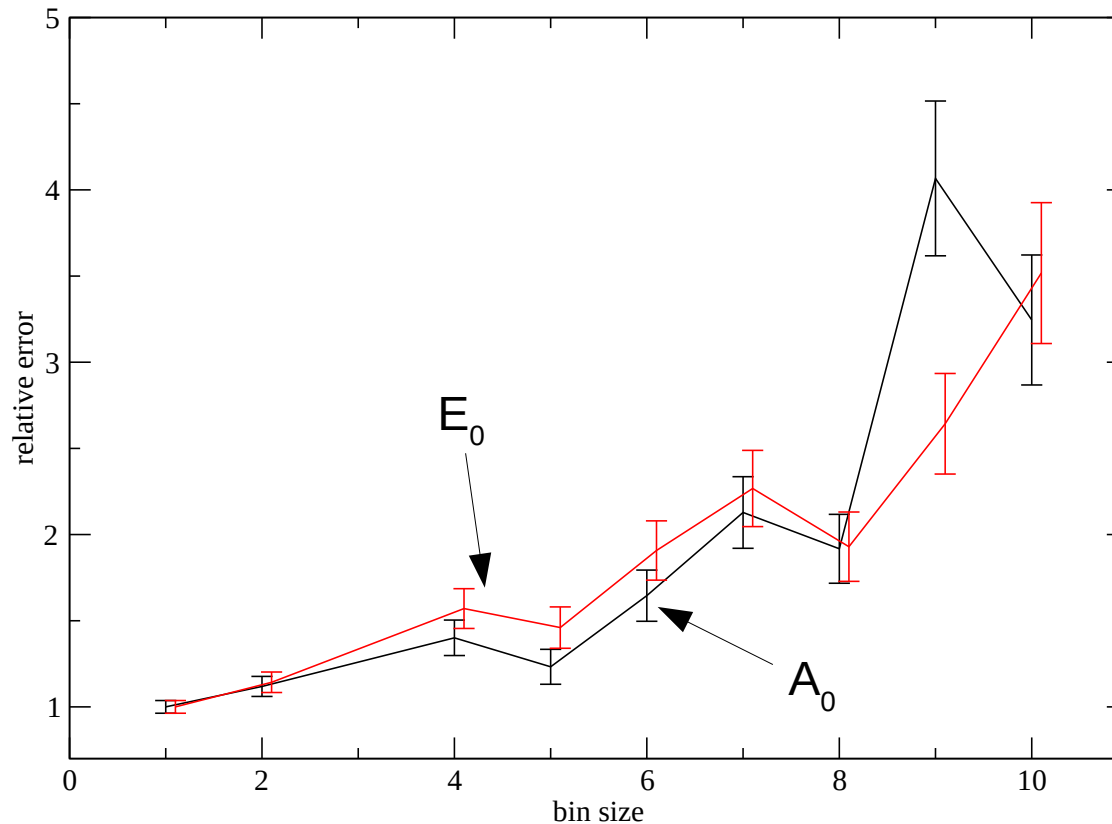
Dealing with autocorrelations

- Evidence suggests $\tau_{\text{int}} \sim 4$ MDTU (1 cfg)
- Naively $\sigma \propto \sqrt{\frac{2\tau_{\text{int}}}{n}}$ expect $\sim 1.4x$ larger errors.
- Standard (crude) strategy is to bin (average) the data over blocks sufficiently large to make the blocks independent.
- Typically see errors grow then stabilize when bin size large enough.

Pion and kaon 2pt functions



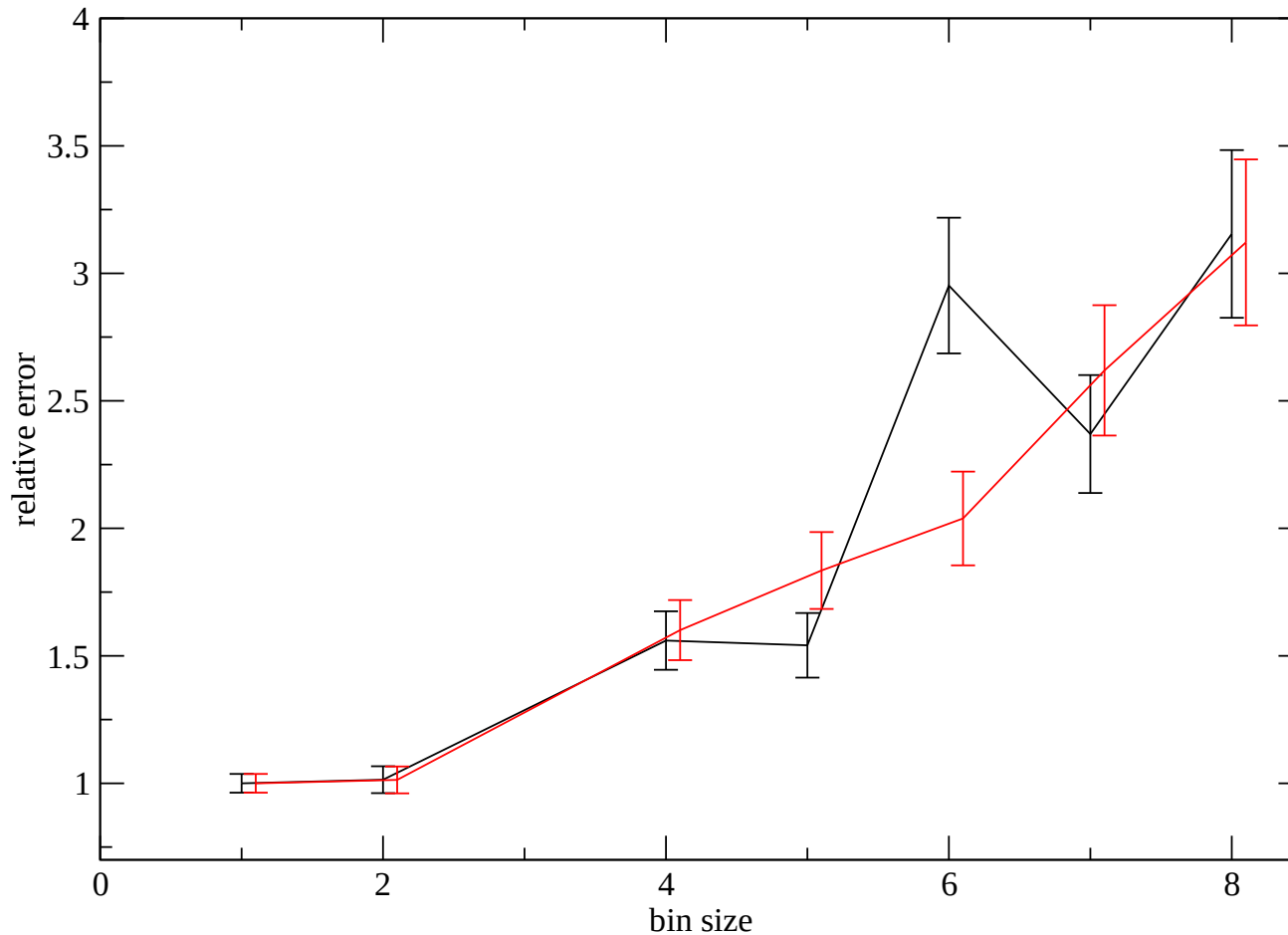
$I=0$ $\pi\pi$ 2pt function



- $\pi\pi$ errors continue growing with bin size and do not stabilize. Why?
- Covariance matrix is 66x66 here!
- As bin size increased, fewer data points enter determination of covariance matrix → matrix becomes less and less well resolved.
- Fluctuations of low eigenvalues increase, causing error growth unrelated to autocorrelation

Scrambled data

- Isolate effect of loss of resolution of covariance matrix by randomly scrambling data to destroy autocorrelations

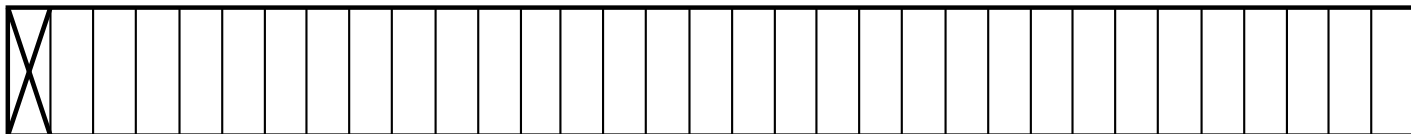
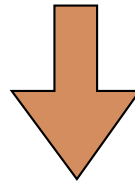


- Error growth essentially the same!

Block jackknife

- To prevent loss of resolution of covariance matrix while still take into account autocorrelations, we perform **block jackknife**

Regular jackknife: generate n “reduced ensembles” of size $n-1$ by dropping successive samples



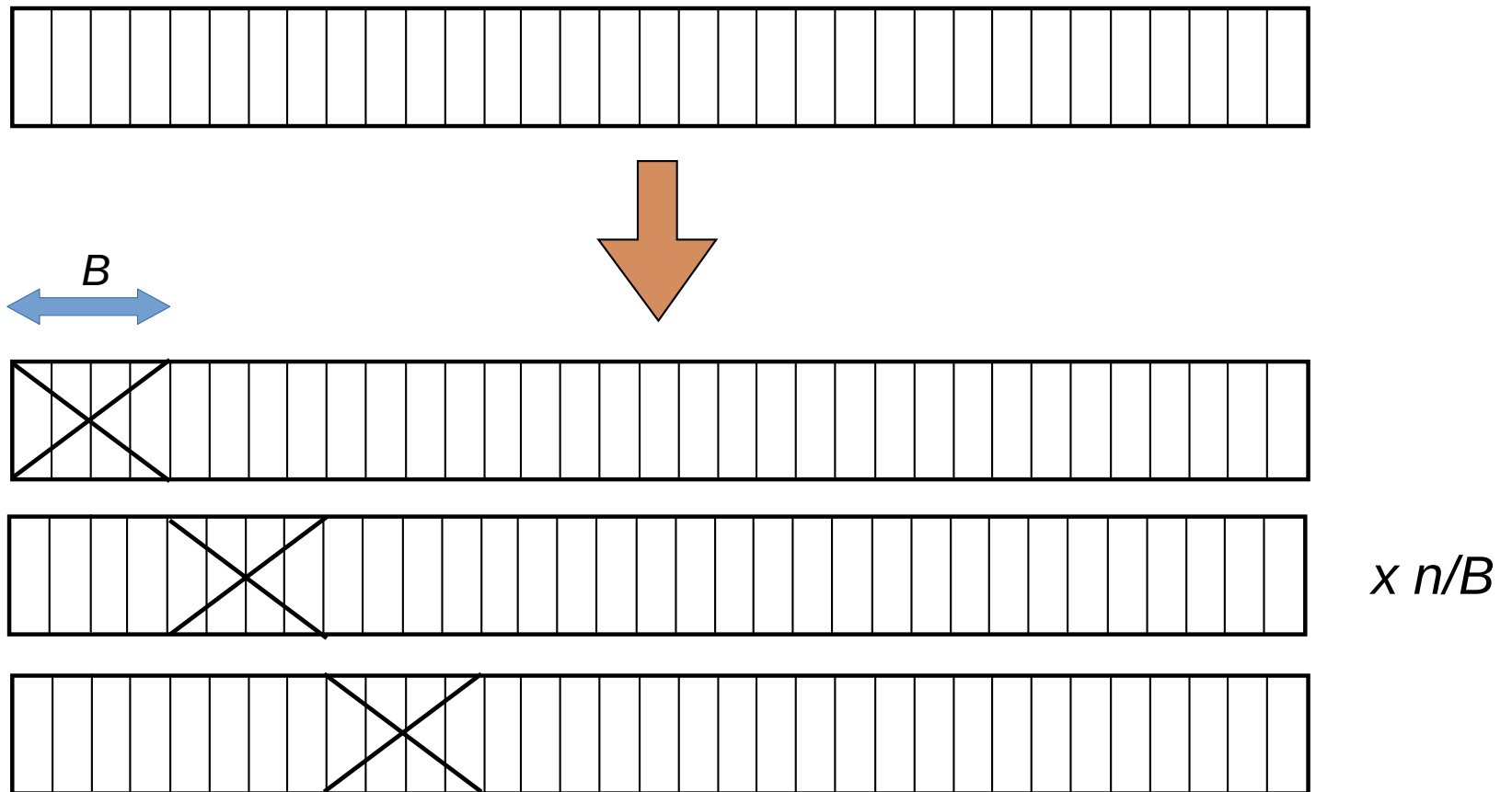
$\times n$



- With binning, covariance matrix obtained from just $n/B-1$ samples

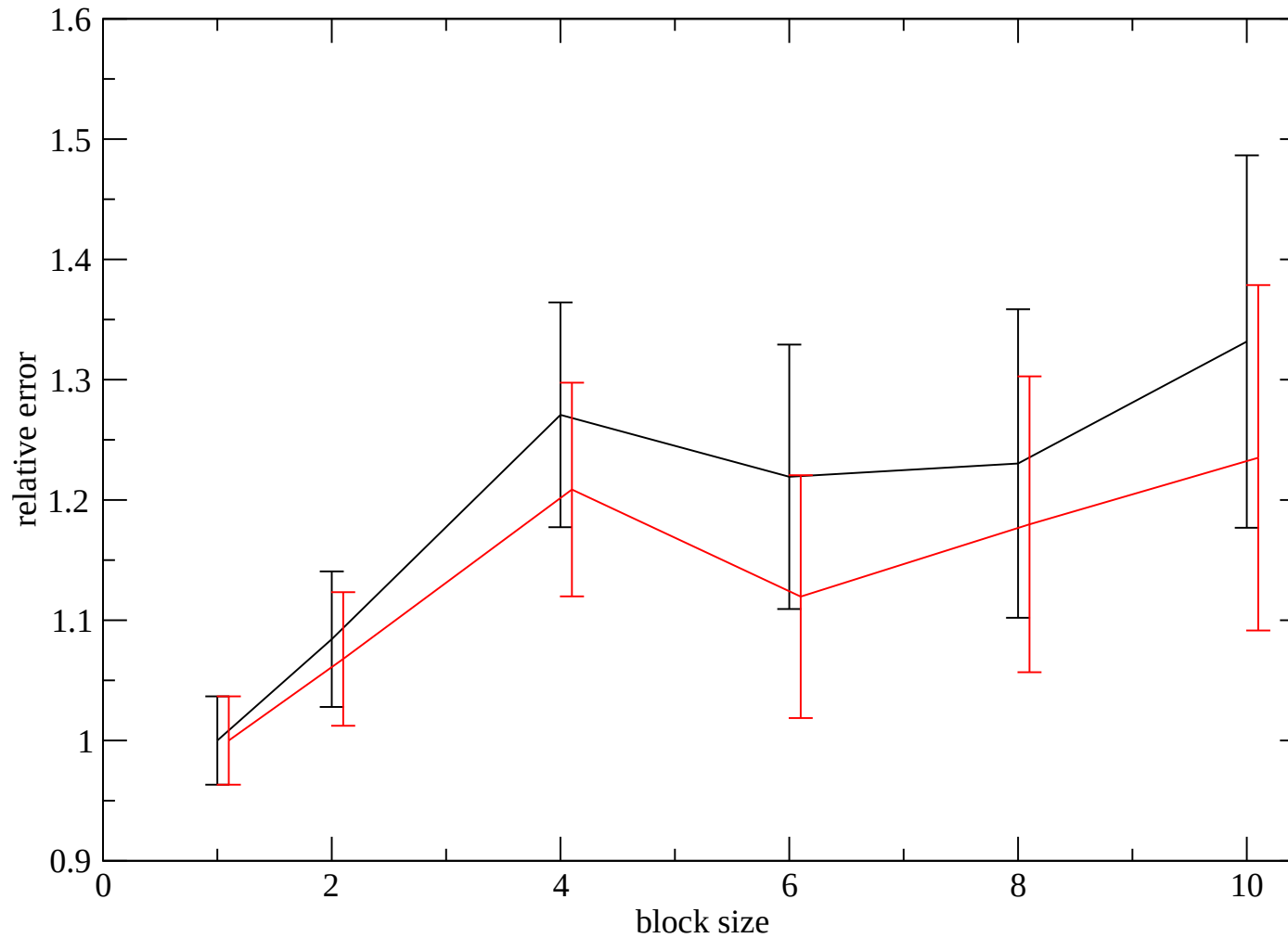
Block jackknife II

block jackknife: generate n/B reduced ensembles of size $n-B$ by throwing away successive **blocks** of size B



- Covariance matrix obtained from $n-B$ samples !
- Jackknife procedure ensures correct statistical error

$l=0$ $\pi\pi$ 2pt function with block jackknife



- Bravissimo! Problem solved



Goodness of fit

Techniques

- Test statistic is q^2 (often called χ^2).

$$q^2 = \sum_{t, t' = t_{\min}}^{t_{\max}} [\bar{v}_t - f(t, \vec{p})] C_{tt'}^{-1} [\bar{v}_{t'} - f(t', \vec{p})]$$

With covariance matrix obtained from sample covariance:

$$C_{tt'} = \frac{1}{n(n-1)} \sum_{i=1}^n [v_{i,t} - \bar{v}_t] [v_{i,t'} - \bar{v}_{t'}]$$

- **Stone age:** compare q^2/dof to 1.
Doesn't take into account fact that deviations from 1 less likely for large dof k as std.dev of distribution

$$\text{std. dev}(q^2/\text{dof}) \sim \sqrt{2/k}$$

- **Bronze age:** compare q^2 to $\chi^2(k = \text{dof})$ distribution, compute **p-value:**

$$p(q^2, k) = \int_{q^2}^{\infty} dx \text{PDF}_{\chi^2}(x, k)$$

Doesn't take into account fluctuations in covariance matrix between experiments when cov. mat obtained from data

Techniques II

- **Iron age:** For *normally-distributed, independent data*, effects of fluctuations of covariance matrix incorporated in the Hotelling T^2 distribution:

$$q^2 \sim T^2(k, n - 1)$$

Doesn't take into account effects of non-normality or autocorrelations.

Correction to χ^2 significant for large k (dof) :

$$\langle T^2(k, n) \rangle - \langle \chi^2(k) \rangle \approx k^2 / n$$

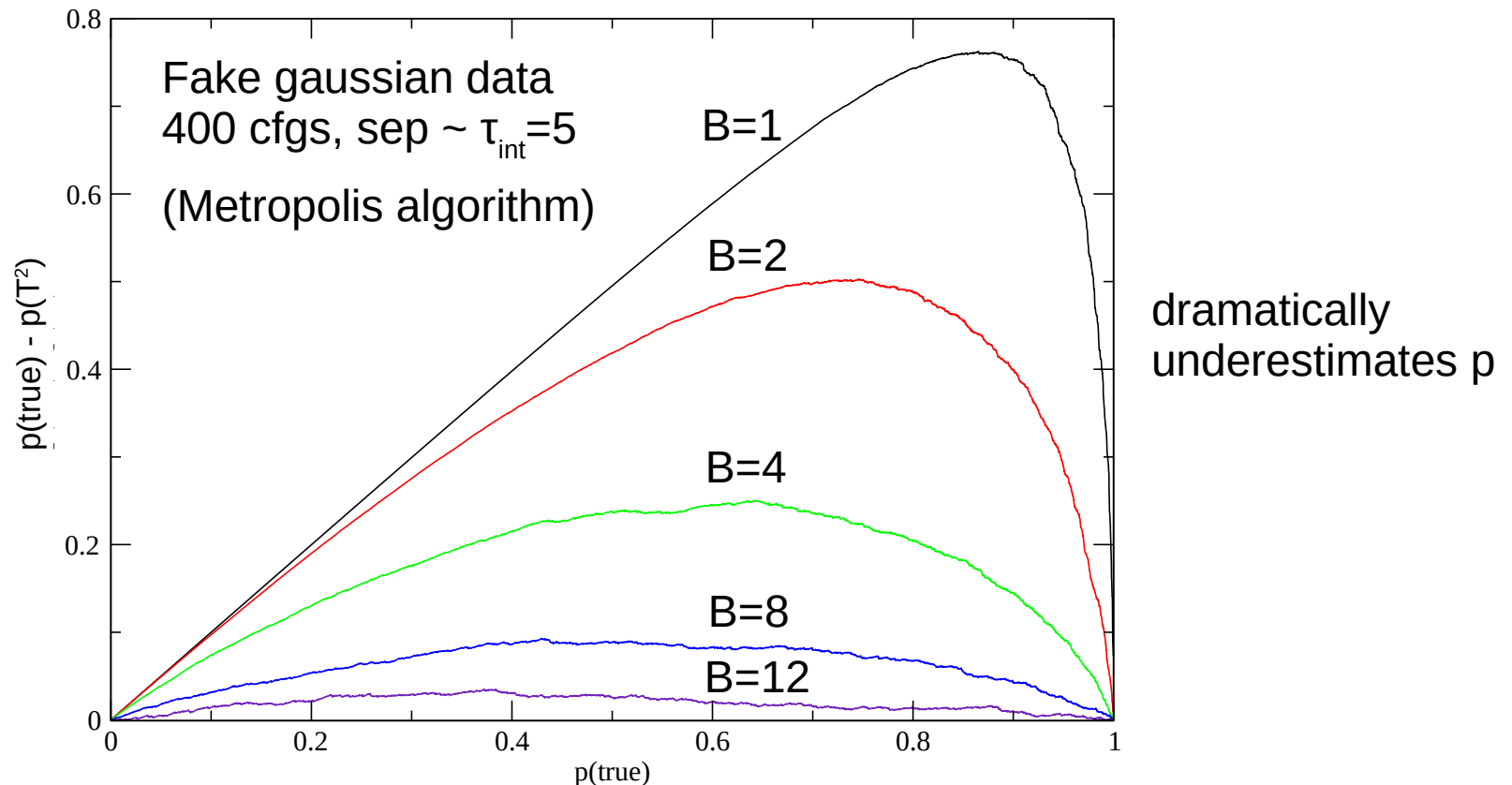
Example:

$l=0$ $\pi\pi$ fit, range 5-15 (no binning): $k=54$, $q^2=74$

$$q^2 / \text{dof} = 1.37 \quad p(\chi^2) = 0.04 \quad p(T^2) = 0.1$$

Is the Hotelling distribution sufficient?

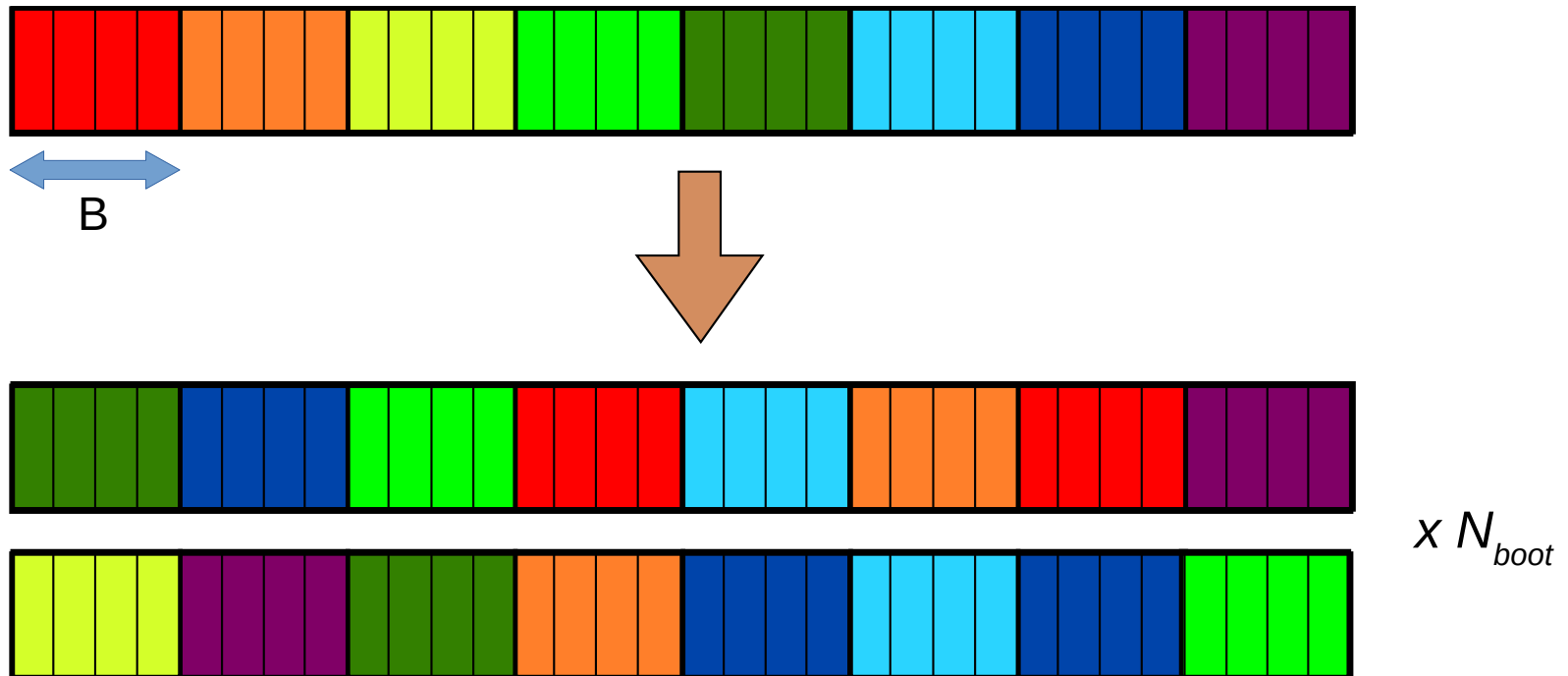
- Numerical experiments with fake data show Hotelling T^2 relatively tolerant of non-normality.
- **However** T^2 relies on independent configurations: *extremely* sensitive to autocorrelations.
- Even with binning, slow convergence to true distribution:



- Wish to avoid binning due to explosion in statistical error from reduced resolution of covariance matrix

Space age: Non-overlapping block bootstrap (NBB)

- The **bootstrap** technique allows us to estimate properties of the population from just one ensemble, by randomly resampling (with replacement).
- The (non-overlapping) block variant resamples blocks rather than single configurations, much like block jackknife, in order to account for autocorrelations:



Space age II

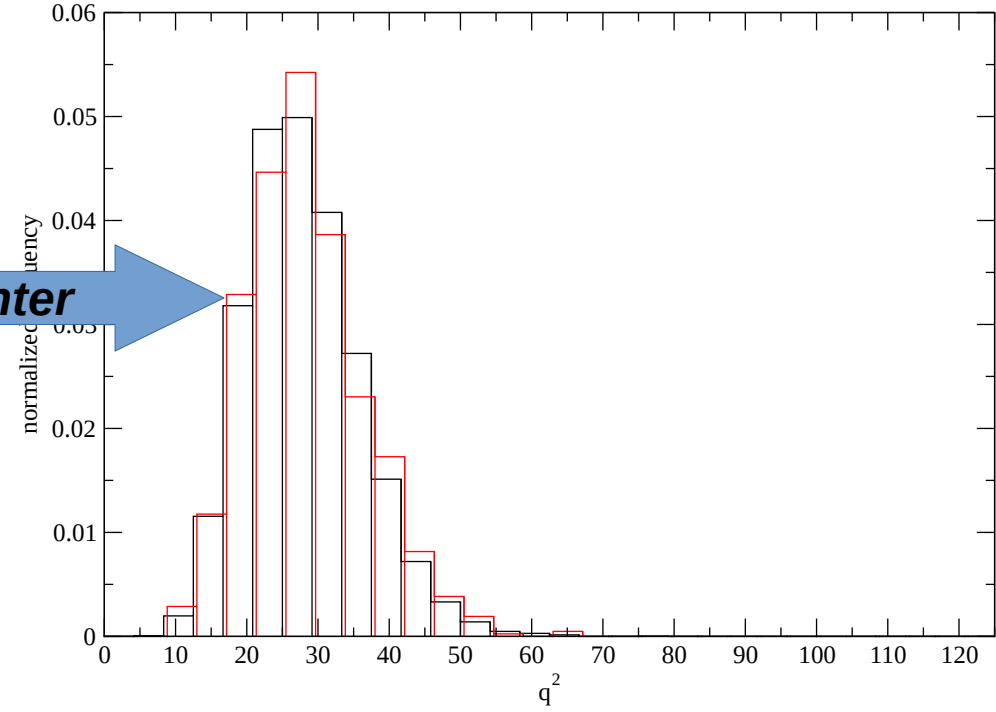
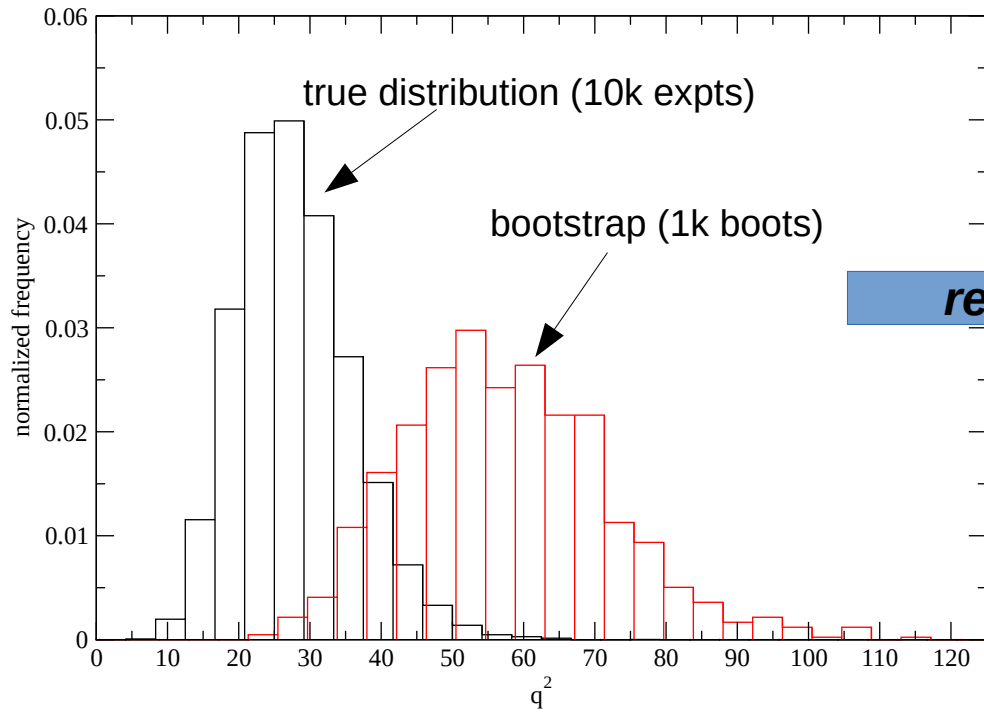
- Use NBB to directly compute the distribution of q^2 !
 - ✓ No normality assumption
 - ✓ Blocking accounts for autocorrelations without binning
- Minor subtlety: bootstrap ensemble means \bar{b}^α distributed about ensemble mean \bar{v} **not population mean**
- Results in broader distribution of q^2 with higher mean
- Correct by “recentering”: $\bar{b}^\alpha(t) \rightarrow \bar{b}^\alpha(t) + [f(t, \vec{p}) - \bar{e}(t)]$

fit to original ensemble



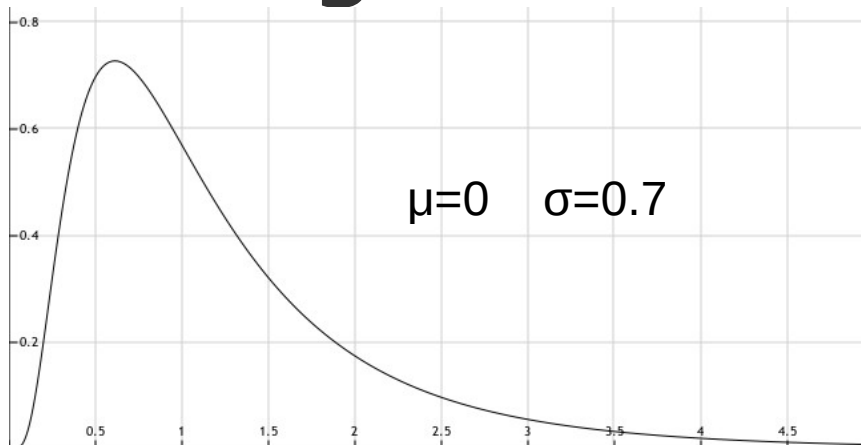
Demonstration

gaussian data, no autocorrelations, 400 samples

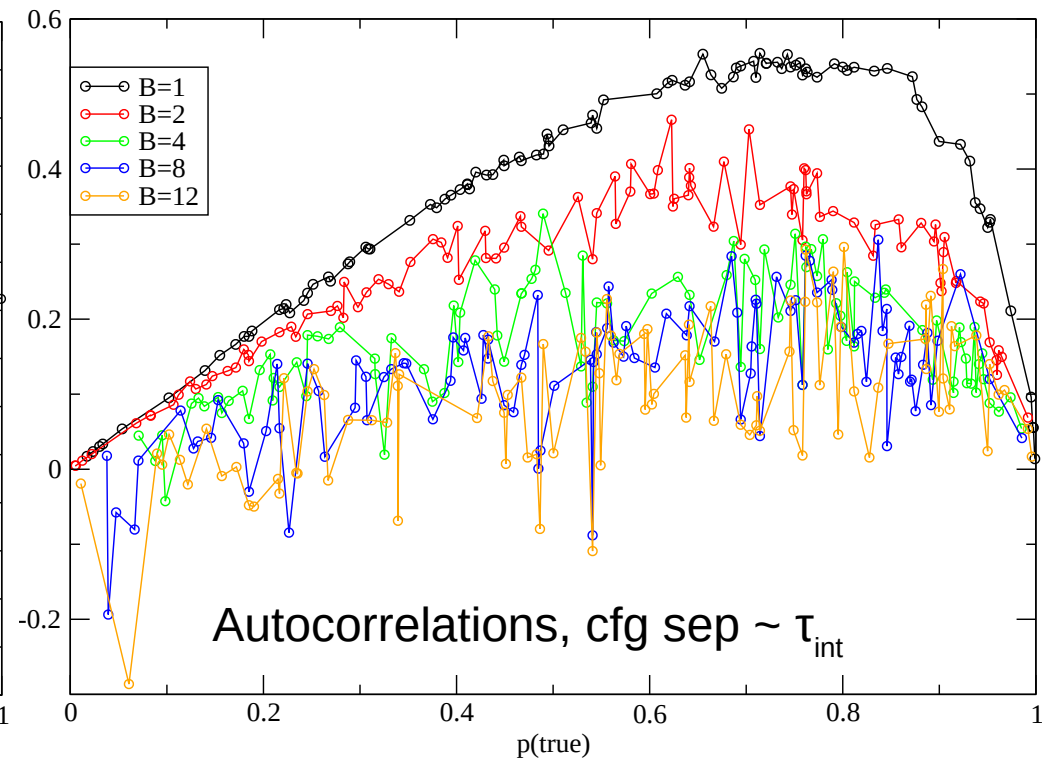
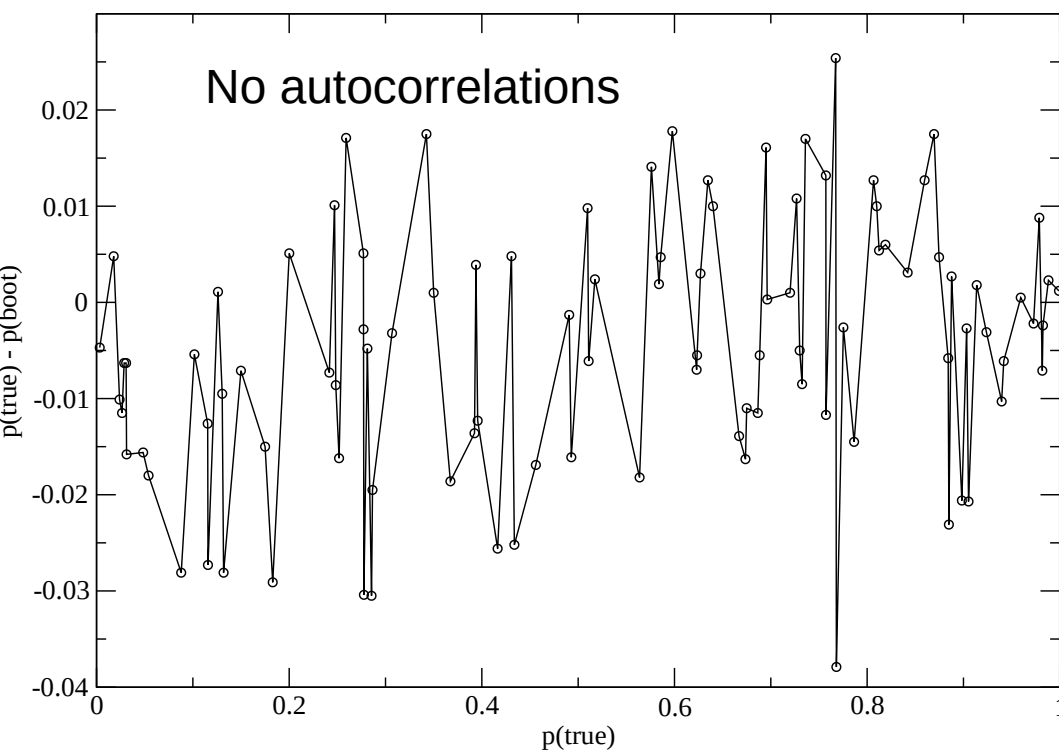


Demonstration II - log-normal

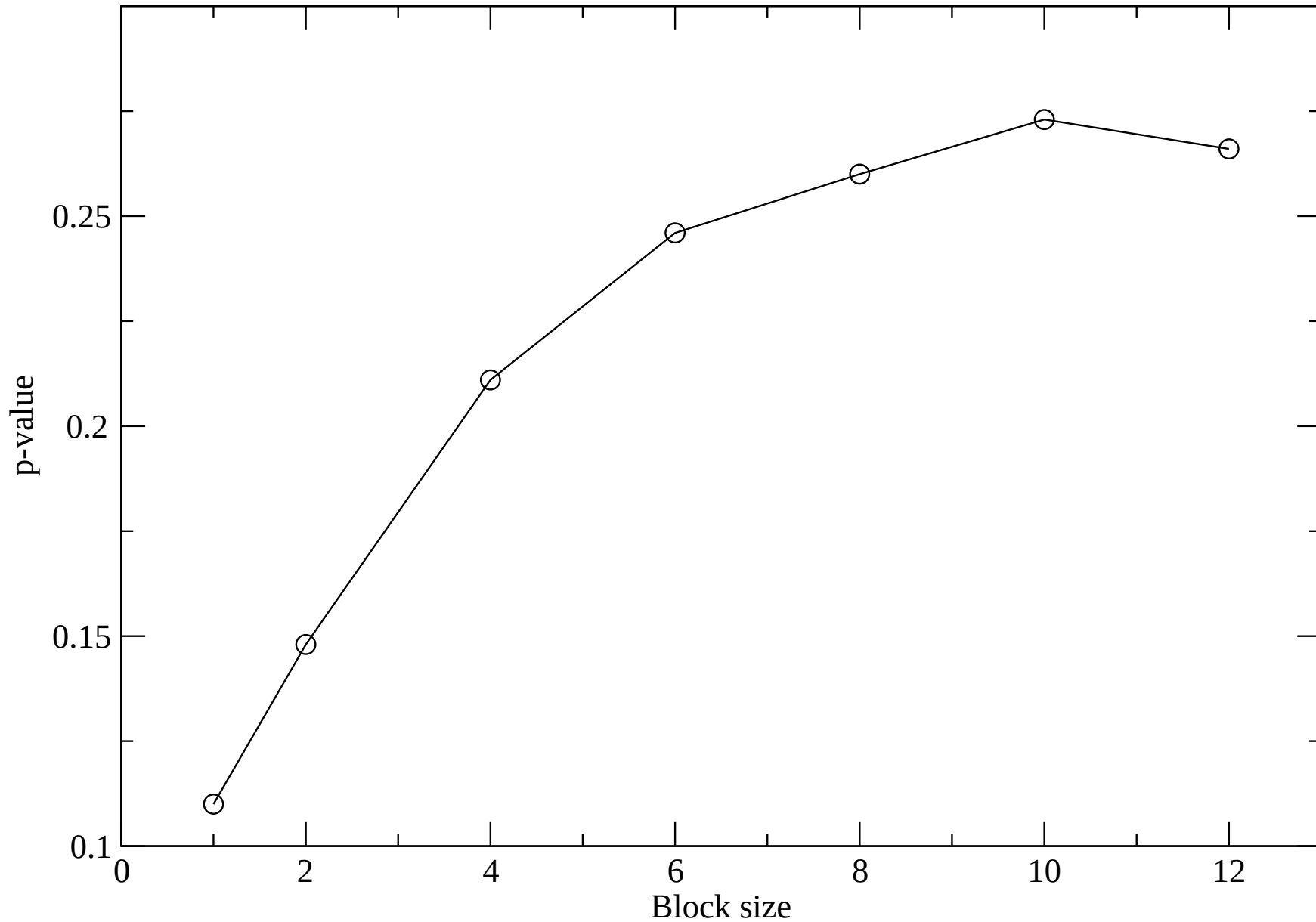
400 cfgs, log-normal



Stat error and bias fall as $n, B \rightarrow \infty$ ($B \ll n$)

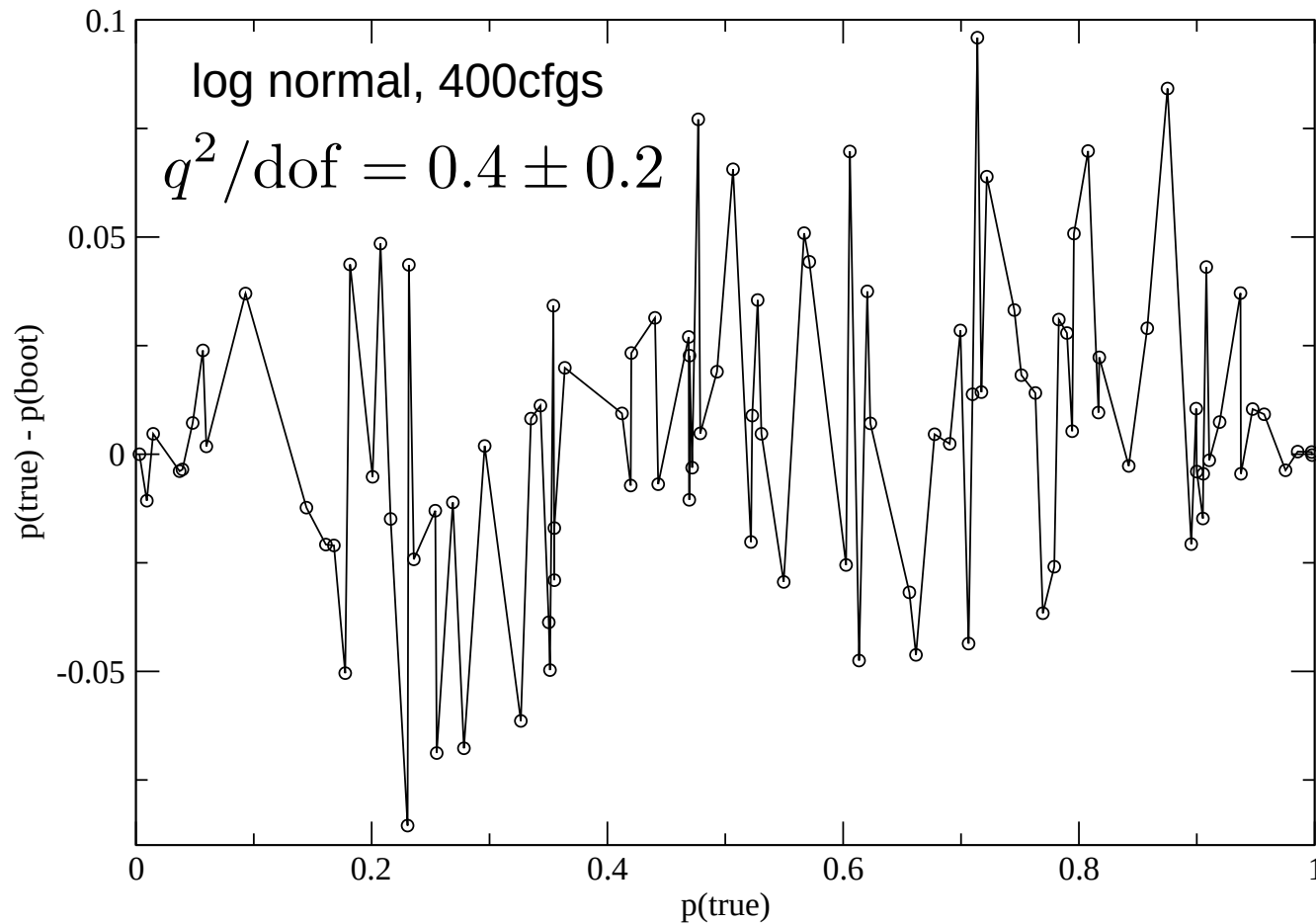


I=0 $\pi\pi$ fit bootstrap p-value



Incidentally....

- Conventional wisdom is that one cannot obtain the goodness-of-fit for uncorrelated fits. Space-age man can!





Conclusions

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- Publication delayed by desire to account for autocorrelations and obtain reliable goodness-of-fit metrics for fits.
- Binning causes uncontrolled error growth due to loss of resolution of covariance matrix.
- Solve by performing block jackknife and computing covariance matrix from unbinned data.
- Even most advanced of old-fashioned approaches to computing goodness-of-fit fail to account for effects of autocorrelations or non-normality.
- Use non-overlapping block bootstrap to directly predict the distribution of q^2 from which a p-value can be obtained without any assumptions.
- Expect no further hurdles to completion of project and we aim to publish within the next few months.

Thank you!