

Casimir effect in critical $O(N)$ models with non-equilibrium methods

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ECT* Trento, Bridging analytical and numerical methods for quantum field theory

Based on [AB](#), [Caselle](#), [Nada](#), [Panero](#); 2505.20403



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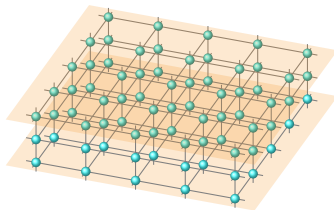
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Thermal CFTs

- A thermal CFT can be defined on the manifold $\mathcal{M}_D = S_l^1 \times \mathbb{R}^{D-1}$
- Enhanced number of CFT data
- **Critical Casimir amplitude**
→ one-point function of the stress-energy tensor

$$\begin{aligned}\langle T_{tt} \rangle_{\mathcal{M}_D} &= (1 - D) \langle T_{ii} \rangle_{\mathcal{M}_D} \\ &= -(D - 1) \Delta l^{-D}\end{aligned}$$

Experimental results



- Attractive force between the boundaries
- **Experimentally measured** in ^4He and binary liquid mixtures
[Garcia et al.; PRL **83** (1999)]
[Garcia et al.; PRL **88** (2002)]
[Hertlein et al.; Nature **451** (2008)]

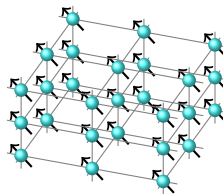
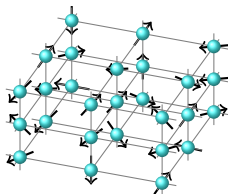
$O(N)$ models

$$H = -J \sum_{\langle ij \rangle} \mathbf{s}_i \cdot \mathbf{s}_j$$

$O(1)$: Ising $O(2)$: XY $O(3)$: Heisenberg

$O(4) \rightarrow$ particle physics $O(\infty)$: spherical model $\rightarrow$ AdS/CFT

Euclidean lattice, $D = 3$,
 $a = 1$

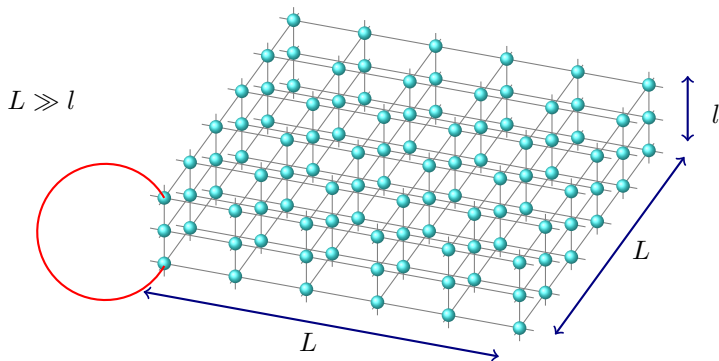


$J = 0$

$J = J_c$

$J = \infty$

Casimir effect: setup



General idea: long-range fluctuations are sensitive to the short direction l

[Vasilyev et al.; PRE **79** (2009)]

free energy density
in the limit $l \rightarrow L$

$$F = L^2 l f = L^2 (\beta^{-1} f_{\text{ex}} + l f_{\text{bulk}})$$

excess of free energy due
to the compact direction

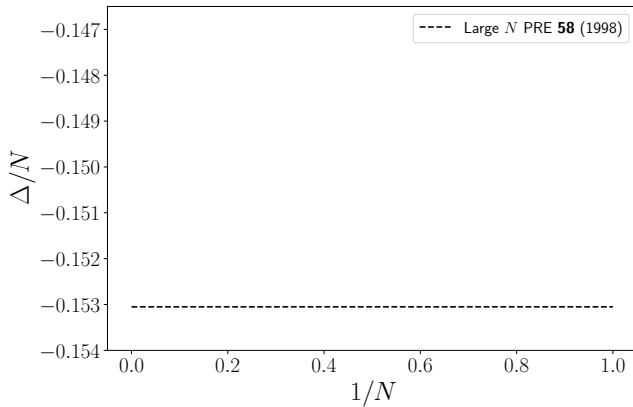
- In the scaling limit $L \gg l \gg 1$

$$f_{\text{ex}} = \Delta l^{-2}$$

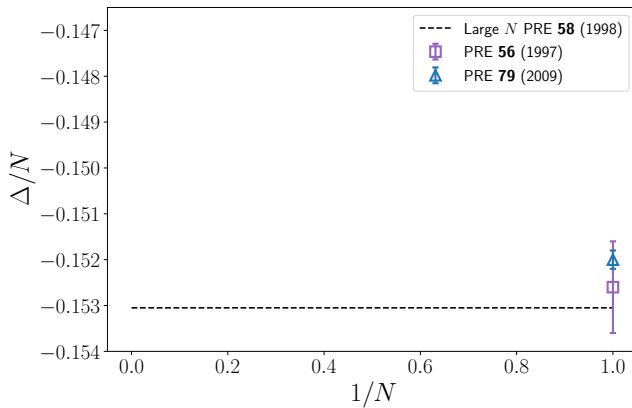
- Δ is the **critical Casimir amplitude**

- The Casimir amplitude is part of the **CFT data** of the theory
- Determined for different $O(N)$ models with different methods

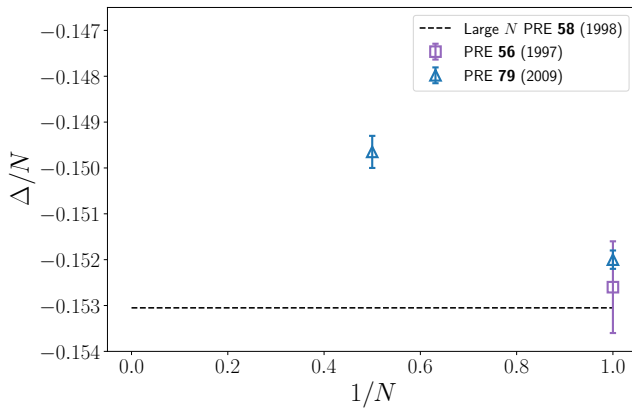
A puzzling quantity...



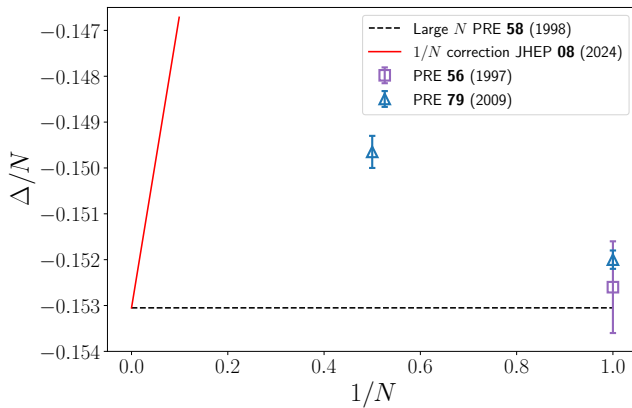
A puzzling quantity...



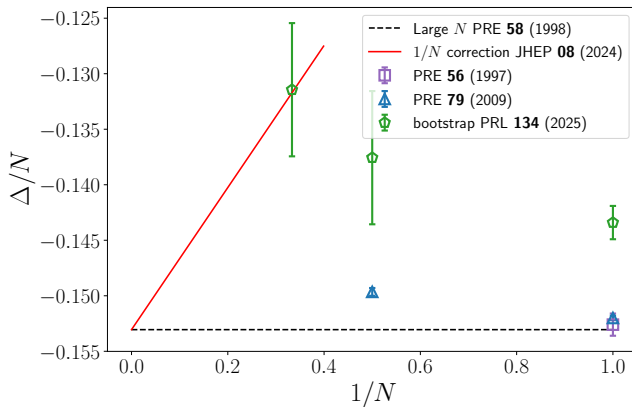
A puzzling quantity...



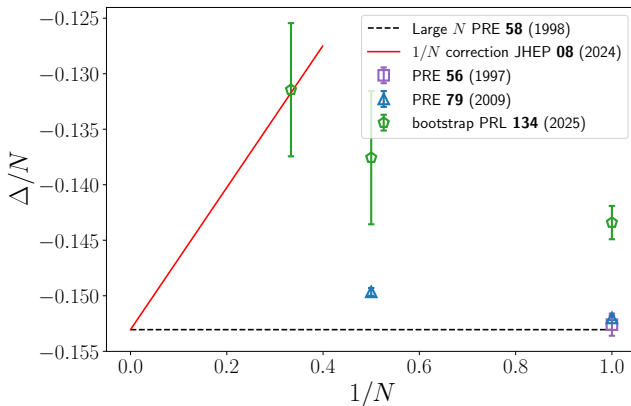
A puzzling quantity...



A puzzling quantity...



A puzzling quantity...



Goals

Better understanding
of large- N behavior

Discriminating between
conflicting results

How do we compute Δ ?

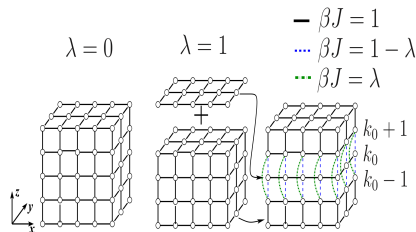
only depends on L

$$F = L^2 l f = L^2 (\beta^{-1} f_{\text{ex}} + l f_{\text{bulk}})$$

Δl^{-2} plus subleading dependence on L

- Markov chain Monte Carlo can be used to compute **free energy differences** (although it is not an ideal method...)

$$F_{\lambda=1} - F_{\lambda=0} = F(l-1) + F_{\text{slab}} - F(l) \simeq -\frac{\partial F}{\partial l} + F_{\text{slab}}$$



From [Vasilyev et al.; PRE **79** (2009)]

unwanted term,
only depends on L

$$\frac{1}{L^2} \frac{\partial F}{\partial l} = \beta^{-1} \frac{\partial f_{\text{ex}}}{\partial l} + f_{\text{bulk}}$$

we want to isolate this
 $\propto \Delta l^{-3}$

- An iterative procedure is employed to remove f_{bulk} and f_{slab}

Can we design a better strategy?



Non-equilibrium simulations
for efficient calculations of ΔF



Different target ΔF to automatically get rid of f_{bulk} and f_{slab}

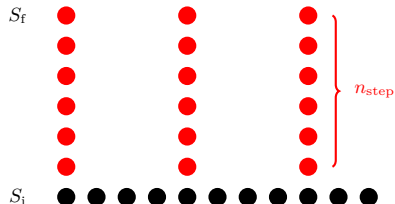
Strategy

- We want to compute $F_f - F_i$ defined by S_f and S_i
- Introduce a **discrete protocol** $c(n)$ connecting S_i and S_f
- **Non-equilibrium Markov Chain** following $c(n)$ in n_{step}

$$\phi_0 \xrightarrow{P_{c(1)}} \phi_1 \xrightarrow{P_{c(2)}} \dots \xrightarrow{P_{c(n_{\text{step}})}} \phi_{n_{\text{step}}}$$

Realization

[Caselle et al.; PRD **94** (2016)]



Jarzynski's theorem

- How to compute equilibrium quantities?
- Jarzynski's theorem [Jarzynski; PRL 78 (1997)]

$$\Delta F = -\frac{1}{\beta} \log \langle e^{-\beta W} \rangle_{\text{non-eq}}$$

$$W = \sum_{n=0}^{n_{\text{step}}-1} S_{c(n+1)}(\phi_n) - S_{c(n)}(\phi_n)$$

- Mathematically exact result, also for expectation values

$$\langle \mathcal{O} \rangle_{\text{eq}} = \frac{\langle \mathcal{O} e^{-W} \rangle_{\text{non-eq}}}{\langle e^{-W} \rangle_{\text{non-eq}}}$$

Advantages

- ✓ Systematics well under control
- ✓ Linear scaling with the number of d.o.f.
[Bonanno et al.; JHEP 04 (2024)]
[AB et al.; PRD 111 (2025)]
- ✓ Systematically improvable by increasing n_{step} or using **Stochastic Normalizing Flows**

$$\times \quad \frac{\beta}{L^2} \frac{\partial F}{\partial l} = \frac{\partial f_{\text{ex}}}{\partial l} + \beta f_{\text{bulk}} \qquad \checkmark \quad \frac{\beta}{L^2} \frac{\partial^2 F}{\partial l^2} = \frac{\partial^2 f_{\text{ex}}}{\partial l^2} \propto \Delta l^{-4}$$

On the lattice second derivatives are computed as differences of first derivatives

$$\frac{\partial^2 F}{\partial l^2} \simeq [F(l+1) - F(l)] - [F(l) - F(l-1)]$$



Forward derivative Backward derivative

Drawback: not a primary observable, the two derivatives are uncorrelated.

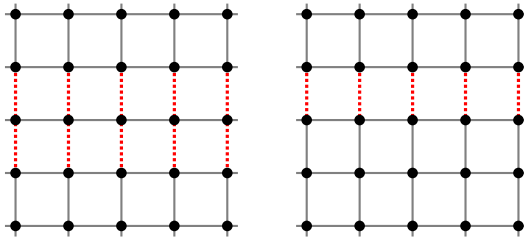
We propose an alternative strategy

$$\frac{\partial^2 F}{\partial l^2} \simeq F(l+1) + F(l-1) - 2F(l)$$

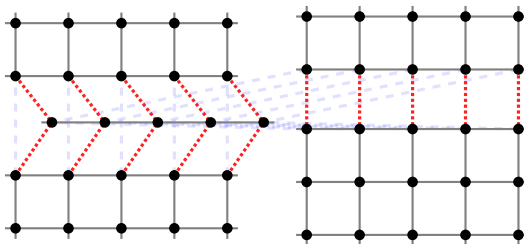


Target F_f Prior F_i

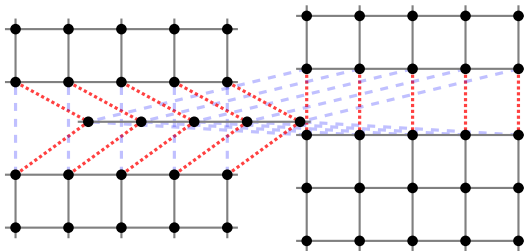
The slab-exchange protocol



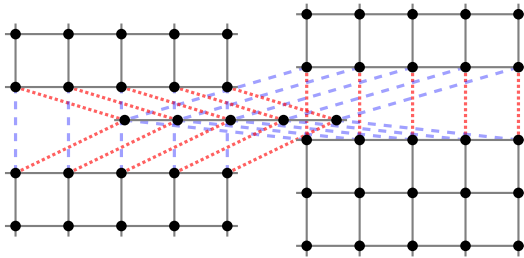
The slab-exchange protocol



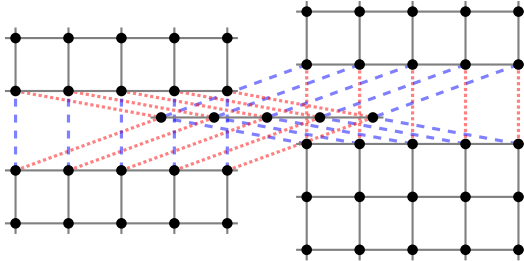
The slab-exchange protocol



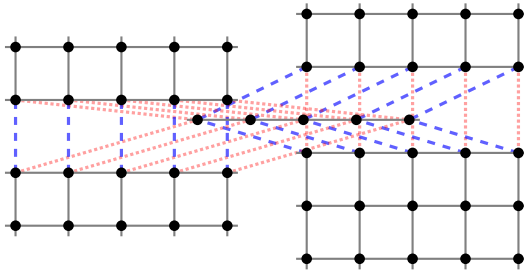
The slab-exchange protocol



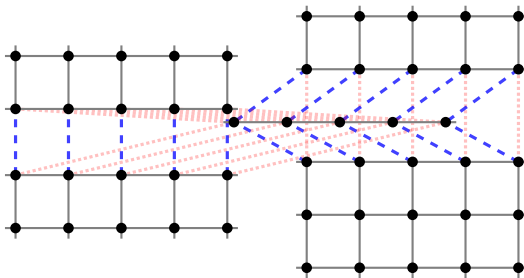
The slab-exchange protocol



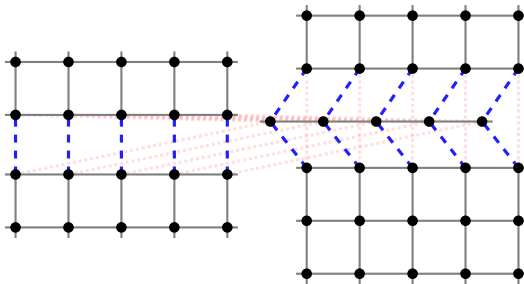
The slab-exchange protocol



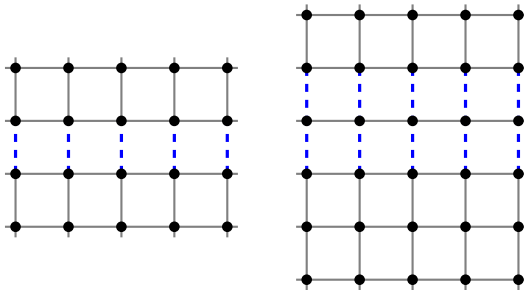
The slab-exchange protocol



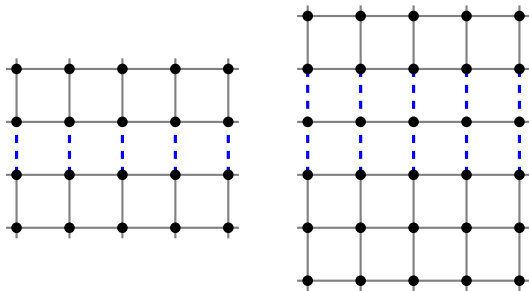
The slab-exchange protocol



The slab-exchange protocol

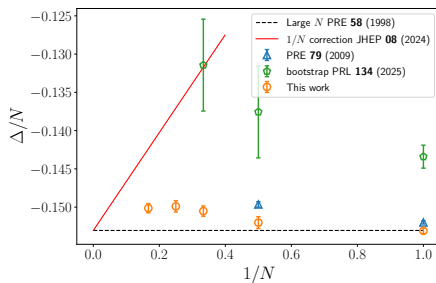
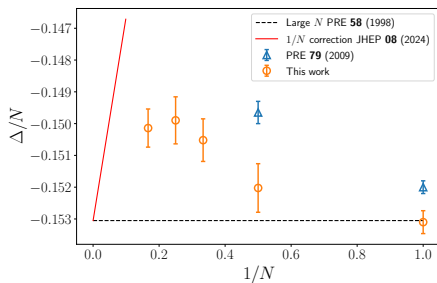


The slab-exchange protocol



$$\frac{\partial^2 F}{\partial l^2} \simeq \textcolor{brown}{F}(l+1) + \textcolor{brown}{F}(l-1) - \textcolor{blue}{2F}(l) = -\frac{1}{\beta} \log \langle e^{-\beta W} \rangle_{\text{non-eq}}$$

Our results



Conclusions

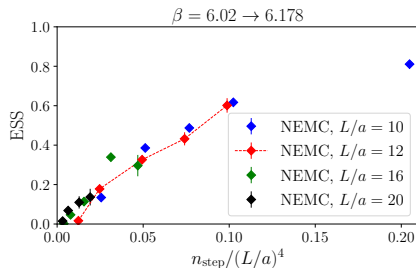
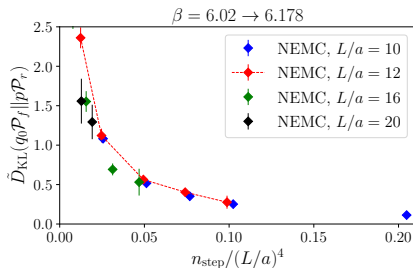
- **New methods** and range of parameters **never explored before**
- **Careful analysis** of extrapolations and systematic effects
- $N = 6$ is not large- N (differently from $SU(N)$ gauge theories)
- Our analysis **aligns with other MC studies over bootstrap**

Outlook

- For better understanding N dependence:
 - simulations at **larger N** (~ 10)
 - calculations at **higher orders** ($1/N^2$)
- Understanding the **discrepancy with bootstrap**
- Why is $N = 1$ so close to $N = \infty$?

Backup slides

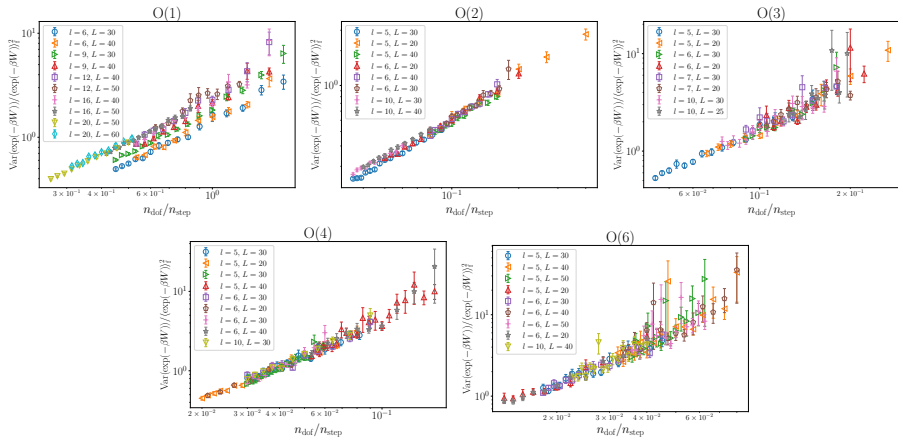
- Theory: SU(3) pure gauge theory in 4D
- Protocol: linear interpolation from $\beta = 6.02$ (mild autocorrelation) to $\beta = 6.178$ (larger autocorrelation)



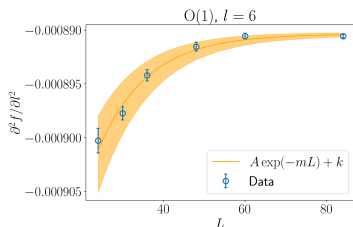
$$\text{ESS} = \frac{\langle e^{-\beta W} \rangle_{\text{non-eq}}^2}{\langle e^{-2\beta W} \rangle_{\text{non-eq}}} \qquad \frac{\text{Var}(e^{-\beta W})}{\langle e^{-\beta W} \rangle_{\text{non-eq}}^2} = \frac{1}{\text{ESS}} - 1$$

$$\frac{\text{Var}(e^{-\beta W})}{\langle e^{-\beta W} \rangle_{\text{non-eq}}^2} = \sum_{k=1}^{k_{\max}} v_k \left(\frac{n_{\text{dof}}}{n_{\text{step}}} \right)^k$$

Calibration of the algorithm II



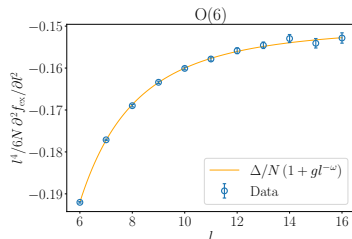
$L \rightarrow \infty$ limit



$$\frac{\partial^2 f_{\text{ex}}}{\partial l^2}(L; A, k, m) = A \exp(-mL) + k$$

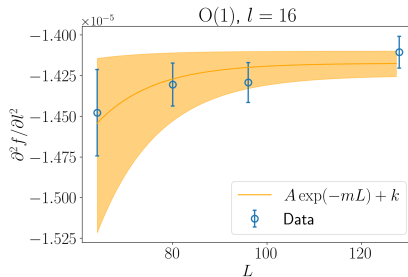
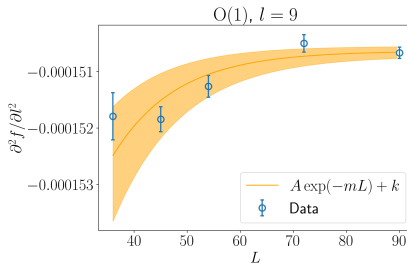
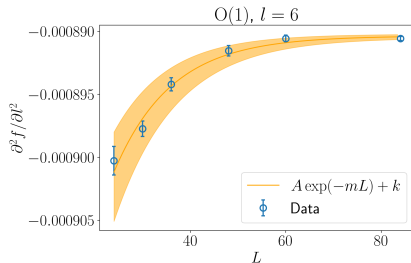
- Removing fine size effects in the transverse direction
- m global parameter for fixed N

Scaling in l

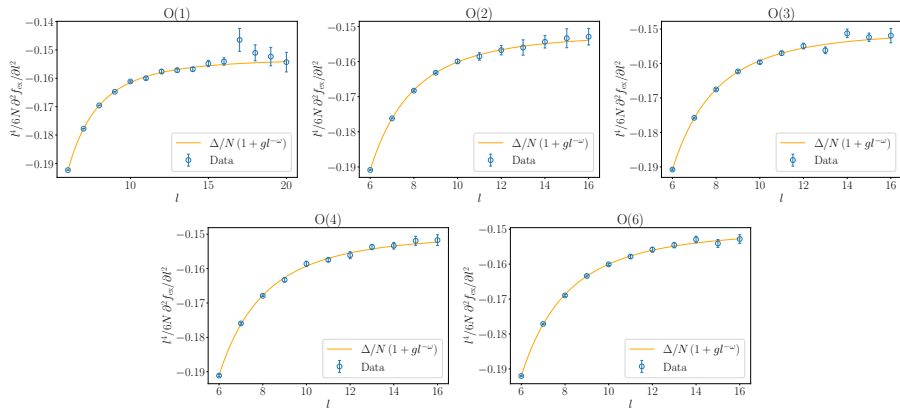


$$\frac{\partial^2 f_{\text{ex}}}{\partial l^2}(l; \Delta, g, \omega) = 6 \Delta l^{-4} (1 + gl^{-\omega})$$

- Fit function motivated by **finite size scaling** [Vasilyev et al.; PRE 79 (2009)]
- Good quality of the fit for all N s



Scaling in l



	O(1)	O(2)	O(3)	O(4)	O(6)
$\chi^2_{\text{red}}[\#\text{dof}]$	1.18 [12]	0.21 [8]	1.33 [8]	0.78 [8]	0.46 [8]
g	57(4)	63(10)	59(9)	49(9)	44(6)
ω	3.02(5)	3.08(10)	3.01(10)	2.90(11)	2.83(8)