Improving the Five-Point Bootstrap

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50 + ϵ Years of Conformal Bootstrap



[Kos, DP, Simmons-Duffin, Vichi '16; Simmons-Duffin '16]

- The conformal bootstrap has had some surprising successes in computing low-lying CFT data in some theories
- This comes from applying crossing to 4-point functions involving scalars, and more recently fermions, currents, and stress tensors



[Kos, DP, Simmons-Duffin, Vichi '16; Simmons-Duffin '16]

- ▶ However, some basic data is not so easy to access using this approach, e.g. 3-point couplings like $\langle C^{\ell=4}C^{\ell=4}\epsilon \rangle$ and $\langle C^{\ell=4}C^{\ell=4}T^{\ell=2} \rangle$
- ▶ In principle this data can be computed using 4-point functions like $\langle C^{\ell=4}C^{\ell=4}C^{\ell=4}C^{\ell=4}\rangle$, but it has 881 tensor structures!

The five-point bootstrap



- ► Recently we started exploring what can be extracted from CFT 5-point functions like (\$\phi_1\phi_2\phi_3\phi_4\phi_5\$) [DP, Prilepina, Tadic, May '23; Dec '23]
- ▶ It gives a convenient probe of 3-point functions w/ 2 spinning operators: $\langle \mathcal{O}_{\Delta,\ell} \phi \, \mathcal{O}'_{\Delta',\ell'} \rangle^{(n_{IJ})} \propto V_1^{\ell-n_{IJ}} V_3^{\ell'-n_{IJ}} H_{13}^{n_{IJ}} \quad (n_{IJ} = 0, \dots, \min(\ell, \ell'))$

Five-point blocks

$$\begin{aligned} \langle \phi_1(x_1)\phi_2(x_2)\phi_3(x_3)\phi_4(x_4)\phi_5(x_5)\rangle &= \\ \sum_{(\mathcal{O}_{\Delta,\ell},\mathcal{O}'_{\Delta',\ell'})} \sum_{n_{IJ}=0}^{\min(\ell,\ell')} (\lambda_{\phi_1\phi_2\mathcal{O}_{\Delta,\ell}})(\lambda_{\phi_4\phi_5\mathcal{O}'_{\Delta',\ell'}})(\lambda_{\mathcal{O}_{\Delta,\ell}\phi_3\mathcal{O}'_{\Delta',\ell'}}^{n_{IJ}}) \\ &\times P(x_i)G_{(\Delta,\ell,\Delta',\ell')}^{(n_{IJ})}(u_1,v_1,u_2,v_2,w) \end{aligned}$$

$$u_1 = \frac{x_{12}^2 x_{34}^2}{x_{13}^2 x_{24}^2}, \ v_1 = \frac{x_{14}^2 x_{23}^2}{x_{13}^2 x_{24}^2}, \ u_2 = \frac{x_{23}^2 x_{45}^2}{x_{24}^2 x_{35}^2}, \ v_2 = \frac{x_{25}^2 x_{34}^2}{x_{24}^2 x_{35}^2}, \ w = \frac{x_{15}^2 x_{23}^2 x_{34}^2}{x_{24}^2 x_{13}^2 x_{35}^2}$$

Blocks with scalars exchanged can be computed as a series expansion [Rosenhaus '18; Parikh '19; Fortin, Ma, Skiba '19]

Blocks with spins exchanged can be computed via a couple methods:

- Recursion relations relating $\ell \rightarrow \ell 1$ [DP, Prilepina '21]
- Solving two quadratic Casimir equations order by order [Goncalves, Pereira, Zhou '19; DP, Prilepina, Tadic, May '23; Dec '23]

Cross ratios



[Buric, Lacroix, Mann, Quintavalle, Schomerus '21]

One can go to a conformal frame which puts {x₂, x₃, x₄} at {0,∞,1}
 The position of x₁ on a plane is specified by {z₁, z
₁}, x₅ on a different plane by {z₂, z
₂}, and the angle between the planes w₁ = sin(φ/2)²

$$u_1 = z_1 \overline{z}_1, \qquad v_1 = (1 - z_1)(1 - \overline{z}_1), u_2 = z_2 \overline{z}_2, \qquad v_2 = (1 - z_2)(1 - \overline{z}_2), w = w_1(z_1 - \overline{z}_1)(z_2 - \overline{z}_2) + (1 - z_1 - z_2)(1 - \overline{z}_1 - \overline{z}_2)$$

Cross ratios



We've found it useful to define a set of "radial coordinates" on each plane: (4pt radial coords: [Pappadopulo, Rychkov, Espin, Rattazzi '12; Hogervorst, Rychkov' 13])

$$z_{i} = \frac{4\rho_{i}}{(1+\rho_{i})^{2}}, \qquad \rho_{i} = r_{i}e^{i\theta_{i}}, \qquad \eta_{i} = \cos\theta_{i}, \qquad i = 1, 2,$$
$$R = \sqrt{r_{1}r_{2}}, \qquad r = \sqrt{\frac{r_{1}}{r_{2}}}, \qquad \hat{w} = \left(\frac{1}{2} - w_{1}\right)\sqrt{(1-\eta_{1}^{2})(1-\eta_{2}^{2})}$$

In these coordinates the blocks have a nice expansion:

$$G_{(\Delta,\ell,\Delta',\ell')}^{(n_{IJ})}(R,r,\eta_{1},\eta_{2},\hat{w}) = \sum_{n=0}^{\infty} R^{\Delta+\Delta'+n} \sum_{m} \sum_{j_{1},j_{2}}^{\min(j_{1},j_{2})} \sum_{k=0}^{\min(j_{1},j_{2})} c\left(\frac{n+m}{2},\frac{n-m}{2},j_{1},j_{2},k\right) r^{\Delta-\Delta'+m} \eta_{1}^{j_{1}-k} \eta_{2}^{j_{2}-k} \hat{w}^{k}$$
$$m \in [-n,-n+2,\ldots,n-2,n]$$
$$j_{1} \in \left[\frac{n+m}{2}+\ell,\frac{n+m}{2}+\ell-2,\ldots,\operatorname{Mod}\left(\frac{n+m}{2}+\ell,2\right)\right]$$
$$j_{2} \in \left[\frac{n-m}{2}+\ell',\frac{n-m}{2}+\ell'-2,\ldots,\operatorname{Mod}\left(\frac{n-m}{2}+\ell',2\right)\right]$$

The power of R gives total exchanged dimension and there is a single ∞ -sum

The blocks satisfy two quadratic Casimir equations $\mathcal{D}_{12}^2 G = \mathcal{D}_{45}^2 G = 0$, giving two recursion relations for the *c*-coefficients:

$$\sum_{\{\hat{m}_1, \hat{m}_2, \hat{j}_1, \hat{j}_2, \hat{k}\} \in \mathcal{S}_j} q_j(\hat{m}_1, \hat{m}_2, \hat{j}_1, \hat{j}_2, \hat{k})$$

$$c\left(\frac{n+m}{2} + \hat{m}_1, \frac{n-m}{2} + \hat{m}_2, j_1 + \hat{j}_1, j_2 + \hat{j}_2, k + \hat{k}\right) = 0$$

They have 499 terms but can be easily solved in e.g. Mathematica
 To relate to structure n_{IJ} labeling block have boundary conditions:

$$c(0,0,\ell,\ell',k) = (-1)^{\ell+\ell'+k+n_{IJ}} 2^{k+2(\Delta+\Delta')} \binom{n_{IJ}}{k}$$

Mean-field theory

- One application is to expand a known 5-point function in blocks and read off OPE coefficients
- E.g., we can expand the MFT correlator $\langle \phi \phi \phi^2 \phi \phi \rangle$:

$$\langle \phi(x_1)\phi(x_2)\phi^2(x_3)\phi(x_4)\phi(x_5)\rangle = \left(\frac{x_{24}}{x_{12}x_{23}x_{34}x_{45}}\right)^{2\Delta} \times \\ \sqrt{2}\left((u_1)^{\Delta} + (u_2)^{\Delta} + (u_1u_2)^{\Delta} + \left(\frac{u_1u_2}{v_1}\right)^{\Delta} + \left(\frac{u_1u_2}{v_2}\right)^{\Delta} + \left(\frac{u_1u_2}{w}\right)^{\Delta}\right)$$

and read off the product of OPE coefficients

$$P_{n,\ell,n',\ell'}^{n_{IJ}} \equiv \lambda_{\phi\phi[\phi,\phi]_{n,\ell}} \lambda_{\phi\phi[\phi,\phi]_{n',\ell'}} \lambda_{[\phi,\phi]_{n,\ell}[\phi,\phi]_{0,0}[\phi,\phi]_{n',\ell'}}^{n_{IJ}}$$

(Here $[\phi,\phi]_{n,\ell}\sim\phi\partial^{\mu_1\dots\mu_\ell}\partial^{2n}\phi)$ are double-twist operators)

Mean-field theory

- The coefficients with leading twists (n = n' = 0) were computed in [Antunes, Costa, Goncalves, Vilas Boas '22]
- ▶ We were able to extract the general formula [DP, Prilepina, Tadic, May '23]:

$$\begin{split} P_{n,\ell,n',\ell'}^{n_{IJ}} &= \\ \frac{(-1)^{n_{IJ}} 2^{\frac{5}{2} - n_{IJ}} (\ell - n_{IJ} + 1)_{n_{IJ}} (\ell' - n_{IJ} + 1)_{n_{IJ}} (\Delta)_{\frac{\ell}{2} + n} (\Delta)_{\frac{\ell'}{2} + n'}}{\ell! \ell'! n! n'! n_{IJ}! (\ell + \nu + 1) (\ell' + \nu + 1) (\ell + \nu + 2)_{n-1} (\ell' + \nu + 2)_{n'-1}} \\ \frac{(\Delta - \nu)_n (\Delta - \nu)_{n'} (\Delta - \nu)_{n+n'}}{(\frac{\ell - 1}{2} + n + \Delta)_{\frac{\ell}{2}} (\frac{\ell' - 1}{2} + n' + \Delta)_{\frac{\ell'}{2}} (n + 2\Delta - 2\nu - 1)_n (n' + 2\Delta - 2\nu - 1)_{n'}} \\ \frac{(\Delta)_{\ell+n+n'} (\Delta)_{\ell'+n+n'}}{(\ell + n + 2\Delta - \nu - 1)_n (\ell' + n' + 2\Delta - \nu - 1)_{n'} (\Delta)_{n+n'+n_{IJ}}} \end{split}$$

We can also try the numerical bootstrap [DP, Prilepina, Tadic, May '23]:

- \blacktriangleright We expand $\langle \sigma \sigma \epsilon \sigma \sigma \rangle$ in the (12)(45) and (14)(25) OPEs
- After separating the $(\mathbb{1}, \epsilon) + (\epsilon, \mathbb{1})$ contributions, we get a sum rule:

$$\sum_{\mathcal{O},\mathcal{O}'\neq\mathbb{1}} \frac{\lambda_{\sigma\sigma\mathcal{O}}\lambda_{\mathcal{O}\epsilon\mathcal{O}'}^{n_{IJ}}\lambda_{\sigma\sigma\mathcal{O}'}\left(PG_{\mathcal{O},\mathcal{O}'}-\tilde{P}\tilde{G}_{\mathcal{O},\mathcal{O}'}\right)}{\lambda_{\sigma\sigma\epsilon}\left(\tilde{P}\tilde{G}_{\mathbb{1},\epsilon}+\tilde{P}\tilde{G}_{\epsilon,\mathbb{1}}-PG_{\mathbb{1},\epsilon}-PG_{\epsilon,\mathbb{1}}\right)} = 1$$

- ▶ We truncate to include operators with $\Delta \leq \Delta_{\text{cutoff}}$, and fix exchanged dimensions and $\lambda_{\sigma\sigma\mathcal{O}}$ coefficients to their known values
- Then we choose a set of derivatives of the sum rule and search for the λ^{n_{IJ}}_{O∈O'} coefficients that make these equations closest to being satisfied (by minimizing a cost function)

Five-point configuration



• Expand around configuration: $\phi = \pi/2$ and $x_{12}^2 = x_{14}^2 = x_{52}^2 = x_{54}^2 = 1$ • In radial coordinates it is: $R = 2 - \sqrt{3}, r = 1, \eta_i = \hat{w} = 0$

Five-point block convergence



• The radius of convergence in R is controlled by the singularity $x_{15}^2 = 0$

► In general it is a complicated function of {r, η₁, η₂, ŵ} (determined by the smallest root of an 8th order polynomial)

▶ Above shows
$$\tilde{R} = R_{\max}/(2 - \sqrt{3})$$
 vs. \hat{w} at $r = 1, \eta_i = 0$

Five-point block convergence



$$R = 2 - \sqrt{3}, r = 1, \eta_1 = \eta_2 = \hat{w} = 0.$$

Blocks/derivatives eventually converge but first show big oscillations

▶ It is very helpful to accelerate the convergence of the series using a Padé approximant: $G_{\text{Padé}} \equiv \left[\frac{N_{\text{max}}}{2}/\frac{N_{\text{max}}}{2}\right]_G(R)$

• We studied $\langle \phi \phi \phi^2 \phi \phi \rangle$ and included the exchanged operators:

{1,
$$\phi^2$$
, $T_{\mu\nu} \sim \phi \partial_{\mu} \partial_{\nu} \phi$, $C_{\mu\nu\rho\sigma} \sim \phi \partial_{\mu} \partial_{\nu} \partial_{\rho} \partial_{\sigma} \phi$ }

Treat {Δ_φ, λ_{Tφ²T}, λ_{Tφ²C}, λ_{Cφ²C}} as unknowns, input other data
 Pick a set of derivatives of the sum rule D_i(Δ_φ, λ), minimize cost function Σ w_i (D_i(Δ_φ, λ)/D_i(Δ_φ, 0))² with randomly chosen weights w_i ∈ [0, 1]
 Take D_i which gives Δ_φ ~ 0.5 with smallest deviation as w_i varied

Results:

	truncation	ovact	
	truncation	exact	
Δ_{ϕ}	0.5000(3)	0.500000	
$\lambda^0_{T\phi^2T}$	0.52(1)	0.530330	
$\lambda^0_{T\phi^2C}$	0.21(2)	0.226428	
$\lambda_{C\phi^2C}^4$	0.03(2)	0.022097	
$\lambda^3_{C\phi^2C}$	-1.0(3)	-0.618718	
$\lambda_{C\phi^2C}^2$	5.1(9)	2.320194	
$\lambda^{1}_{C\phi^{2}C}$	2(2)	-1.546796	
$\lambda^0_{C\phi^2C}$	1.2(5)	0.096675	

 $\left(\mathsf{Inputs:}\ \Delta_{\phi^2}, \Delta_T, \Delta_C, \lambda_{\phi\phi\phi^2}, \lambda_{\phi\phi T}, \lambda_{\phi\phi C}, \lambda_{\phi^2\phi^2\phi^2}, \lambda_{\phi^2\phi^2 T}, \lambda_{\phi^2\phi^2 C}\right)$

3d free scalar with hard truncation



• We studied $\langle \sigma \sigma \epsilon \sigma \sigma \rangle$ and included the exchanged operators:

$$\{1, \epsilon, \epsilon', T_{\mu\nu}, C_{\mu\nu\rho\sigma}\}$$

- ► Treat {Δ_σ, λ_{T∈T}, λ_{T∈C}, λ_{C∈C}, λ_{ϵ'ϵϵ'}, λ_{ϵ'ϵC}} as unknowns, input other data using best 4-point bootstrap results
- Pick a set of derivatives of the sum rule D_i(Δ_σ, λ), minimize cost function Σ w_i (D_i(Δ_σ,λ)/D_i(Δ_σ,0))² with randomly chosen weights w_i ∈ [0, 1]
 T b D = bit bit is a constraint of the set of the set
- ▶ Take D_i which gives $\Delta_{\sigma} \sim 0.51815$ with smallest deviation as w_i varied

3d Ising with hard truncation

Results:

	truncation	
Δ_{σ}	0.518(2)	
$\lambda_{T\epsilon T}^0$	0.81(5)	
$\lambda_{T\epsilon C}^0$	0.30(6)	
$\lambda_{C\epsilon C}^4$	-0.3(1)	
$\lambda^3_{C\epsilon C}$	-2(2)	
$\lambda_{C\epsilon C}^2$	2(5)	
$\lambda^1_{C\epsilon C}$	-5(11)	
$\lambda^0_{C\epsilon C}$	-3(11)	
$\lambda_{\epsilon'\epsilon\epsilon'}$	1(3)	
$\lambda_{\epsilon' \epsilon C}$	0(2)	

 $(\mathsf{Inputs:}\ \Delta_{\epsilon}, \Delta_{\epsilon'}, \Delta_{T}, \Delta_{C}, \lambda_{\sigma\sigma\epsilon}, \lambda_{\epsilon\epsilon\epsilon}, \lambda_{\sigma\sigma\epsilon'}, \lambda_{\epsilon\epsilon\epsilon'}, \lambda_{\sigma\sigma T}, \lambda_{\epsilon\epsilon T}, \lambda_{\sigma\sigma C}, \lambda_{\epsilon\epsilon C})$

3d Ising with hard truncation



Disconnnected-correlator improvement

- These results can be improved further by introducing an approximation to the truncated contributions [Li, Dec '23; DP, Prilepina, Tadic, Dec '23]
- E.g. for the 4-point function, we can write:

$$\langle \sigma \sigma \sigma \sigma \rangle = \langle \sigma \sigma \sigma \sigma \rangle_{MFT} + \frac{1}{x_{12}^{2\Delta_{\sigma}} x_{34}^{2\Delta_{\sigma}}} \sum_{\mathcal{O}}^{\Delta_{\max}} (P_{\mathcal{O}} g_{\Delta,\ell} - P_{\mathcal{O}}^{MFT} g_{\Delta_{MFT},\ell})$$

where we use $\langle \sigma \sigma \sigma \sigma \rangle_{MFT} = \langle \sigma \sigma \rangle \langle \sigma \sigma \rangle + (\text{perm.})$ and subtract and replace a finite number of MFT contributions

Crossing symmetry then gives a sum rule

$$0 = \sum_{\mathcal{O}}^{\Delta_{\max}} (P_{\mathcal{O}} F_{\Delta,\ell} - P_{\mathcal{O}}^{MFT} F_{\Delta_{MFT},\ell})$$

	no MFT	with $\rm MFT$	Numerical Bootstrap
Δ_{σ}	0.514(5)	0.5182(4)	0.5181489(10)
$\mathcal{P}[\epsilon]$	1.15(4)	1.106(5)	1.106396(9)
$\mathcal{P}[\epsilon']$	-0.010(8)	0.003(2)	0.002810(6)
$\mathcal{P}[T_{\mu\nu}]$	0.33(5)	0.422(2)	0.425463(1)
$\mathcal{P}[C_{\mu\nu\rho\sigma}]$	0.115(9)	0.0768(5)	0.0763(1)

- Here we fixed scaling dimensions of $\{\epsilon, \epsilon', T_{\mu\nu}, C_{\mu\nu\rho\sigma}\}$ to their known values and computed OPE coefficients using the improved truncation
- As before, minimized cost function after selecting "optimal" derivative set and estimated errors by varying random weights of each constraint



- For a 5-point function like $\langle \sigma \sigma \epsilon \sigma \sigma \rangle$ it is not a good idea to use a MFT correlator like $\langle \sigma \sigma \sigma^2 \sigma \sigma \rangle_{MFT}$, since σ^2 is very different from ϵ .
- Instead, one can use $\langle \sigma \sigma \epsilon \sigma \sigma \rangle_d = \langle \sigma \sigma \rangle \langle \epsilon \sigma \sigma \rangle + (\text{perm.})$, which in an AdS dual is the leading contribution of a bulk 3-point interaction



$$\langle \sigma(x_1)\sigma(x_2)\epsilon(x_3)\sigma(x_4)\sigma(x_5)\rangle_d = \frac{\lambda_{\sigma\sigma\epsilon}}{x_{12}^{2\Delta\sigma}x_{45}^{2\Delta\sigma}x_{34}^{\Delta\epsilon}} \left(\frac{x_{24}}{x_{23}}\right)^{\Delta\epsilon} \times \\ \left(u_1^{\frac{\Delta\epsilon}{2}} + u_2^{\frac{\Delta\epsilon}{2}} + \left(\frac{u_1u_2}{v_1v_2}\right)^{\Delta\sigma} \left(v_1^{\frac{\Delta\epsilon}{2}} + v_2^{\frac{\Delta\epsilon}{2}}\right) + \left(\frac{u_1u_2}{w}\right)^{\Delta\sigma} \left(w^{\frac{\Delta\epsilon}{2}} + 1\right)\right)$$

▶ Decomposing in terms of conformal blocks, one finds $(1, \epsilon)$ exchange as well as all the expected $[\sigma, \sigma]_{n,\ell} \sim \sigma \partial^{2n} \partial^{\ell} \sigma$ double-twist contributions

$$\langle \sigma \sigma \epsilon \sigma \sigma \rangle_d = P(x_i) \Big(\lambda_{\sigma \sigma \epsilon} G^{(0)}_{(0,0,\Delta_{\epsilon},0)} + \lambda_{\sigma \sigma \epsilon} G^{(0)}_{(\Delta_{\epsilon},0,0,0)} \\ + \sum_{n,\ell,n',\ell',n_{IJ}} \mathcal{P}(n,\ell,n',\ell',n_{IJ}) G^{(n_{IJ})}_{(2\Delta_{\sigma}+2n+\ell,\ell,2\Delta_{\sigma}+2n'+\ell',\ell')} \Big) \,,$$

with

$$\mathcal{P}(0,2,0,2,0) = \frac{\Delta_{\sigma}^2 \Delta_{\epsilon}^2 (\Delta_{\epsilon} + 2)^2 \lambda_{\sigma\sigma\epsilon}}{4 (2\Delta_{\sigma} + 1)^2},$$

$$\mathcal{P}(0,2,0,4,0) = \frac{\Delta_{\sigma}^2 (\Delta_{\sigma} + 1) \Delta_{\epsilon}^2 (\Delta_{\epsilon} + 2)^2 (\Delta_{\epsilon} + 4) (\Delta_{\epsilon} + 6) \lambda_{\sigma\sigma\epsilon}}{96 (2\Delta_{\sigma} + 1) (2\Delta_{\sigma} + 3) (2\Delta_{\sigma} + 5)},$$

etc.

	$\langle \sigma \sigma \epsilon \sigma \sigma \rangle_d$]		$\langle \sigma \sigma \epsilon \sigma \sigma \rangle_d$
$\lambda^2_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,2}}$	-0.353885]	$\lambda^2_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,6}}$	1.034441
$\lambda^1_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,2}}$	-2.892385		$\lambda^1_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,6}}$	-2.351395
$\lambda^0_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,2}}$	0.988418		$\lambda^0_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,6}}$	0.238792
$\lambda^2_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,4}}$	0.580279		$\lambda^4_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,6}}$	-0.426110
$\lambda^1_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,4}}$	-3.100420		$\lambda^3_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,6}}$	-1.729458
$\lambda^0_{[\sigma,\sigma]_{0,2}\epsilon[\sigma,\sigma]_{0,4}}$	0.484382		$\lambda^2_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,6}}$	5.210505
$\lambda^4_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,4}}$	-0.439644		$\lambda^1_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,6}}$	-2.475238
$\lambda^3_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,4}}$	-0.231194		$\lambda^0_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,6}}$	0.117022
$\lambda^2_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,4}}$	3.849048		$\lambda^6_{[\sigma,\sigma]_{0,6}\epsilon[\sigma,\sigma]_{0,6}}$	-0.318259
$\lambda^1_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,4}}$	-3.276278		$\lambda^5_{[\sigma,\sigma]_{0,6}\epsilon[\sigma,\sigma]_{0,6}}$	0.419036
$\lambda^0_{[\sigma,\sigma]_{0,4}\epsilon[\sigma,\sigma]_{0,4}}$	0.237375		$\lambda^4_{[\sigma,\sigma]_{0,6}\epsilon[\sigma,\sigma]_{0,6}}$	0.882017

▶ Plugging in $\{\Delta_{\sigma}, \Delta_{\epsilon}, \lambda_{\sigma\sigma\epsilon}\}$, we can get an approximation to the unknown 3d Ising data (inversion of $(1, \epsilon) + (\epsilon, 1)$ exchange)

5pt bootstrap for 3d Ising with disc. improvement (DI)

Now we can use it as the starting point for an improved 5pt truncation:

$$0 = \sum_{(\mathcal{O}_{\Delta,\ell},\mathcal{O}'_{\Delta',\ell'})\in\mathcal{S}} \sum_{n_{IJ}=0}^{\min(\ell,\ell')} \lambda_{\sigma\sigma\mathcal{O}_{\Delta,\ell}} \lambda_{\sigma\sigma\mathcal{O}'_{\Delta',\ell'}} \lambda_{\mathcal{O}_{\Delta,\ell}\epsilon\mathcal{O}'_{\Delta',\ell'}}^{n_{IJ}} \mathcal{F}_{\Delta,\ell,\Delta',\ell'}^{n_{IJ}} - \sum_{([\sigma,\sigma]_{n,\ell},[\sigma,\sigma]_{n',\ell'})\in\mathcal{S}_d} \sum_{n_{IJ}=0}^{\min(\ell,\ell')} \mathcal{P}(n,\ell,n',\ell',n_{IJ}) \mathcal{F}_{2\Delta_{\sigma}^d+2n+\ell,\ell,2\Delta_{\sigma}^d+2n'+\ell',\ell'}^{n_{IJ}}$$

- S contains all pairs from $\{\epsilon, \epsilon', T_{\mu\nu}, C_{\mu\nu\rho\sigma}\}$
- S_d contains all pairs from $\{[\sigma,\sigma]_{0,0}, [\sigma,\sigma]_{1,0}, [\sigma,\sigma]_{0,2}, [\sigma,\sigma]_{0,4}\}$
- ► All data except $\{\Delta_{\sigma}, \lambda^0_{T \epsilon T}, \lambda^0_{T \epsilon C}, \lambda^{n_{IJ}=0...4}_{C \epsilon C}, \lambda_{\epsilon' \epsilon C}\}$ fixed to known values

5pt bootstrap for 3d Ising with disc. improvement (DI)

	$\mathcal{S}, \mathrm{no}\mathrm{DI}$	$\mathcal{S}, \mathrm{with}\mathrm{DI}$	
Δ_{σ}	0.518(2)	0.5181(6)	
$\lambda_{T\epsilon T}^0$	0.81(5)	0.96(1)	
$\lambda^0_{T\epsilon C}$	0.30(6)	0.48(3)	
$\lambda_{C\epsilon C}^4$	-0.3(1)	-0.27(2)	
$\lambda_{C\epsilon C}^{\tilde{3}}$	-2(2)	-0.5(4)	
$\lambda_{C\epsilon C}^2$	$_{\epsilon C}$ 2(5) 0.1(9)		
$\lambda_{C\epsilon C}^1$	-5(11)	-10(1)	
$\lambda^0_{C\epsilon C}$	-3(11)	-4(1)	
$\lambda_{\epsilon' \epsilon C}$	$\lambda_{\epsilon'\epsilon C} \mid 0(2) \mid 0.9(4)$		

- Results significantly more constrained
- ▶ See noticable upward shift in $\lambda_{T\epsilon T}^0$ and $\lambda_{T\epsilon C}^0$

	$\mathcal{S}, \mathrm{no}\mathrm{DI}$	$\mathcal{S}, \mathrm{with}\mathrm{DI}$	$+(\epsilon, S_{\mu\nu\rho\sigma\alpha\delta})$	$+(\epsilon, \mathcal{E}_{\mu_1\dots\mu_8})$
Δ_{σ}	0.518(2)	0.5181(6)	0.5181(6)	0.5181(7)
$\lambda_{T\epsilon T}^0$	0.81(5)	0.96(1)	0.959(8)	0.958(7)
$\lambda^0_{T\epsilon C}$	0.30(6)	0.48(3)	0.48(2)	0.48(2)
$\lambda_{C\epsilon C}^4$	-0.3(1)	-0.27(2)	-0.28(2)	-0.28(2)
$\lambda^3_{C\epsilon C}$	-2(2)	-0.5(4)	-0.4(2)	-0.4(2)
$\lambda_{C\epsilon C}^2$	2(5)	0.1(9)	0.3(6)	0.6(9)
$\lambda^1_{C\epsilon C}$	-5(11)	-10(1)	-10.3(7)	-10(2)
$\lambda^0_{C\epsilon C}$	-3(11)	-4(1)	-4.9(6)	-4(2)
$\lambda_{\epsilon' \epsilon C}$	0(2)	0.9(4)	1.0(3)	1.1(3)

Results appear stable against adding spin-6 and spin-8 contributions

5pt bootstrap for 3d Ising with DI



- Let us focus on our determination $\lambda_{T\epsilon T}^0 \simeq 0.958(7)$
- This coefficient was bounded by [Cordova, Maldacena, Turiaci '17] to satisfy:

 $|\lambda_{T\epsilon T}^0| \le 0.981(2)$

Also recently computed using the "fuzzy sphere" method [He, He, Zhu '23]:

$$\lambda_{T\epsilon T}^0 \simeq 0.8057(65)$$

This coefficient can also be probed from the stress tensor bootstrap...

 $n_{\rm max}=6,10,14$ single correlator



[Dymarsky, Kos, Kravchuk, Poland, Simmons-Duffin '17]

[Chang, Dommes, Erramilli, Homrich, Kravchuk, Liu, Mitchell, Poland, Simmons-Duffin]

From stress-tensor 4-point functions one can get general bounds on parity-even and parity-odd scalar gaps, producing map of CFT landscape



[Chang, Dommes, Erramilli, Homrich, Kravchuk, Liu, Mitchell, Poland, Simmons-Duffin]

 \blacktriangleright We have computed bounds on the $\langle TT\epsilon\rangle$ coefficient after assuming gaps compatible with 3d Ising



[Chang, Dommes, Erramilli, Homrich, Kravchuk, Liu, Mitchell, Poland, Simmons-Duffin]

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• Expect definitive results from $\{T, \sigma, \epsilon\}$ mixed system (stay tuned...)

- The 5-point bootstrap works!
- ▶ We should explore other channels/correlators and extend to 6-points
- Blocks are under control, main bottleneck to adding more operators is handling the many unknown OPE coefficients
- Some low-lying observables (e.g. $\lambda_{TT\epsilon}$) are sensitive to hard truncations and greatly benefit from introducing a disconnected approximation
- Approximating the truncated spectrum may be useful more generally in other bootstrap problems where truncation methods are used