

Title: Lecture 1: Introduction and Overview; Bootstrapping Ising mixed correlator

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Collection: Mini-Course of Numerical Conformal Bootstrap

Date: April 24, 2023 - 10:05 AM

URL: <https://pirsa.org/23040135>

# *Introduction; Bootstrapping Ising mixed correlator*

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24/04/2023 Perimeter Bootstrap Minicourse



SIMONS  
FOUNDATION



# Constraints for conformal field theory

1, Operators as conformal rep:

$\mathcal{O}_{\Delta, \ell}$   
 ↑ scaling dimension  
 ↘ spin

Primary :  $K_\mu \mathcal{O}(0) = 0$

Descendant :  $\partial_{\mu_1} \dots \partial_{\mu_\ell} \mathcal{O}$

Scaling dimensions  $\longleftrightarrow$  critical exponents

$$\alpha = 2 - d / (d - \Delta_\epsilon)$$

$$\eta = 2 \Delta_\sigma - d + 2$$

$$\langle \mathcal{O}^{\mu_1 \dots \mu_\ell}(x) \mathcal{O}_{\nu_1 \dots \nu_\ell}(y) \rangle = \frac{I_{\nu_1 \dots \nu_\ell}^{\mu_1 \dots \mu_\ell} - \text{trace}}{|x-y|^{2\Delta_\ell}}, \quad I_\nu^\mu = \delta_\nu^\mu - 2 \frac{(x-y)^\mu (x-y)_\nu}{(x-y)^2}$$

$$\langle \phi_1(x) \phi_2(y) \mathcal{O}^{\mu_1 \dots \mu_\ell}(z) \rangle = \lambda_{\phi_1 \phi_2 \mathcal{O}} \frac{Z^{\mu_1 \dots \mu_\ell} - \text{trace}}{|x-y|^{\Delta_i + \Delta_j - (\Delta_k - \ell)} |y-z|^{\Delta_j + (\Delta_k - \ell) - \Delta_i} |z-x|^{(\Delta_k - \ell) + \Delta_i - \Delta_j}}$$

$$Z^{\mu_1} = \frac{x_{13}^\mu}{x_{13}^2} - \frac{x_{23}^\mu}{x_{23}^2}$$

$$\langle \phi(x) \phi(y) \mathcal{O}^{\mu_1 \dots \mu_\ell}(z) \rangle = 0 \text{ for odd } \ell$$

## Constraints for conformal field theory

2, Operator Product Expansion :  $\phi_i(x) \phi_j(y) = \sum_k \lambda_{ijk} C_k(x-y, \partial_y) \mathcal{O}_k(y)$

↓  
Operator Product Expansion (OPE) coefficients

conformal symmetry fix  $C_k(x-y, \partial_y)$  :  $\langle \phi_i(x) \phi_j(y) \mathcal{O}_k(z) \rangle = \lambda_{ijk} C_k(x-y, \partial_y) \langle \mathcal{O}_k(y) \mathcal{O}_k(z) \rangle$

$$\begin{aligned} \langle \overline{\phi(x_1) \phi(x_2)} \overline{\phi(x_3) \phi(x_4)} \rangle &= \sum \lambda_{12a}^2 C_a(x_1-x_2, \partial_2) C_b(x_3-x_4, \partial_4) \langle \mathcal{O}^a(x_2) \mathcal{O}^b(x_4) \rangle \\ &= \sum_o \lambda_{12o}^2 x_{12}^{-\Delta_o} x_{34}^{-\Delta_o} g_{\Delta, \ell}(u, v) \end{aligned}$$

$$u = z \bar{z} = \frac{x_{12}^2 x_{34}^2}{x_{13}^2 x_{24}^2}, \quad v = (1-z)(1-\bar{z}) = \frac{x_{14}^2 x_{23}^2}{x_{13}^2 x_{24}^2}$$



## Constraints : crossing symmetry

3, Associativity :  $(O_i O_j) O_k = O_i (O_j O_k)$

$$\langle O_1 O_2 O_3 O_4 \rangle \sim \sum_k \begin{array}{c} 1 \\ \diagdown \\ \text{---} O_k \\ \diagup \\ 2 \end{array} \begin{array}{c} 4 \\ \diagdown \\ \text{---} \\ \diagup \\ 3 \end{array} = \sum_{k'} \begin{array}{c} 1 \\ \diagdown \\ \text{---} \\ \diagup \\ 2 \end{array} \begin{array}{c} 4 \\ \diagdown \\ \text{---} O_{k'} \\ \diagup \\ 3 \end{array}$$

Bootstrap Equations :  $u^{-\Delta_\phi} \sum_{O \in \phi \times \phi} \lambda_{\phi\phi O}^2 g_{\Delta, \ell}(u, v) = v^{-\Delta_\phi} \sum_{O \in \phi \times \phi} \lambda_{\phi\phi O}^2 g_{\Delta, \ell}(v, u)$

$$\boxed{\sum_O \lambda_{\phi\phi O}^2 F_{\Delta, \ell}(u, v) = 0} \quad \text{with } F_{\Delta, \ell}^{\phi\phi; \phi\phi} = v^{\Delta_\phi} g_{\Delta, \ell}^{\phi\phi; \phi\phi}(u, v) - u^{\Delta_\phi} g_{\Delta, \ell}^{\phi\phi; \phi\phi}(v, u)$$

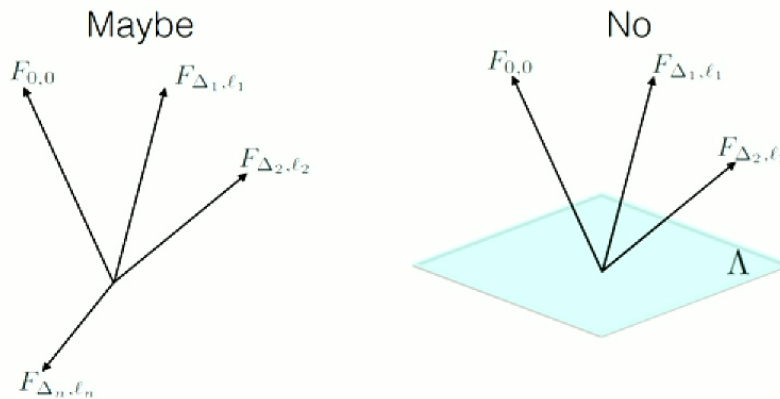
# Numerical bootstrap

- 1, Pick up a set of correlators. Write down bootstrap equations. Truncate equations to finite dimension.

For example :  $\langle \phi(x_1) \phi(x_2) \phi(x_3) \phi(x_4) \rangle : \sum_O \lambda_{\phi\phi O}^2 F_{\Delta,\ell}(u, v) = 0$  (hold for various  $x_i$ )

$$F_{\Delta,\ell}(z, \bar{z}) = \sum c_{mn} \partial_u^m \partial_{\bar{v}}^n F_{\Delta,\ell}(u, v) |_{z=\bar{z}=1/2} (z-1/2)^m (\bar{z}-1/2)^n \quad F_{\Delta,\ell}(z, \bar{z}) \rightarrow \text{vector } (c_{01}, c_{12}, c_{23} \dots)$$

- 2, Unitarity :  $\lambda_{\phi\phi O}^2 \geq 0$

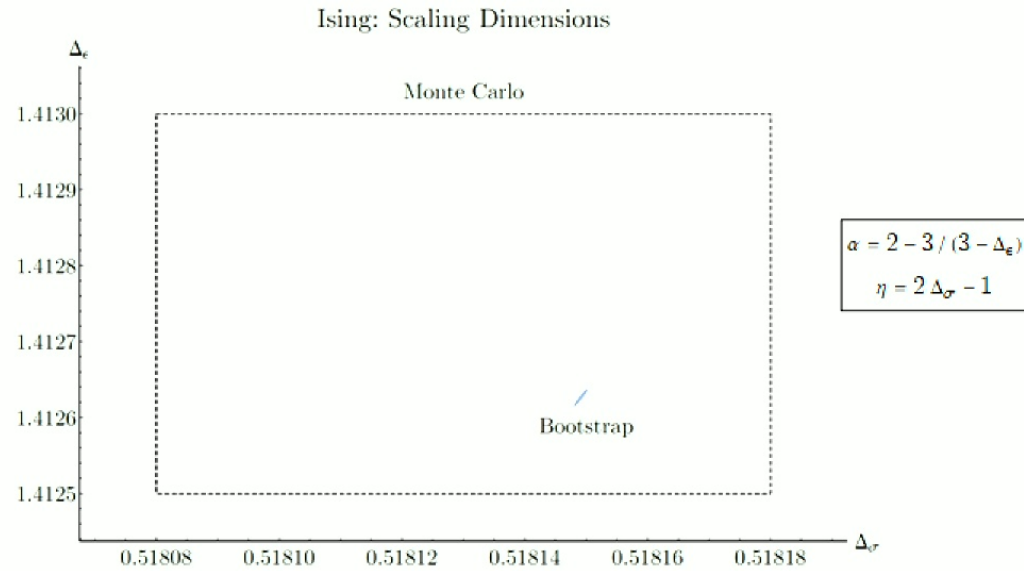


Semidefinite Program (SDP)  $\Rightarrow$   $\begin{cases} \text{plane exist : disallowed} \\ \text{not exist : allowed} \end{cases}$

# Bootstrap 3D Ising CFT

Correlators :  $\langle \sigma\sigma\sigma\sigma \rangle$ ,  $\langle \epsilon\epsilon\epsilon\epsilon \rangle$ ,  $\langle \epsilon\epsilon\sigma\sigma \rangle$

Assumptions :  $\sigma$ ,  $\epsilon$  are the only two relevant scalars.



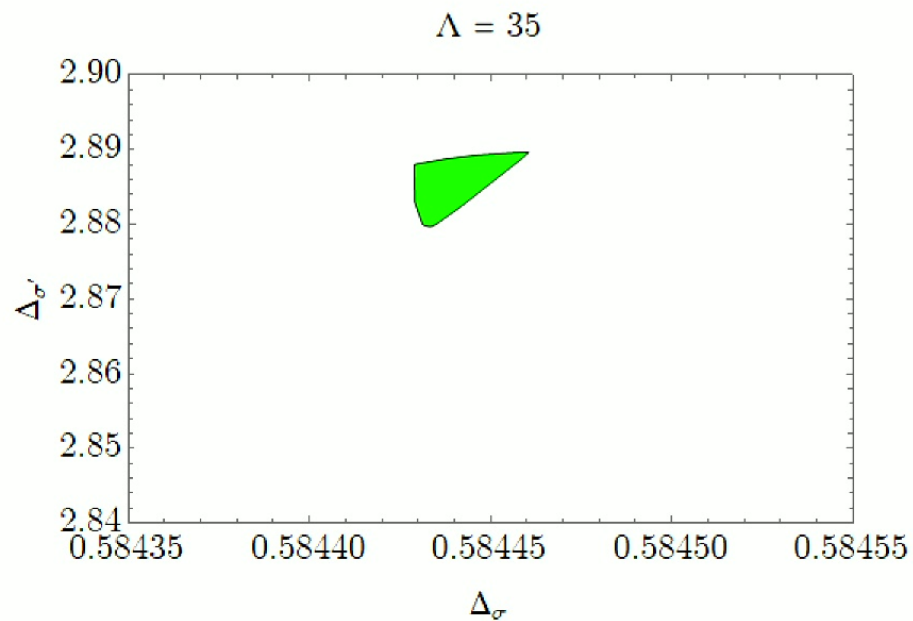
$\eta = 0.036298$  (2),  $\alpha = 0.11008$  (1)

(Kos, Poland, Simmons-Duffin, Vichi 2016)

# Bootstrap 3D super-Ising

$$\mathcal{L}_{\text{SuperIsing}} = \frac{1}{2} (\partial_\mu \sigma)^2 + \frac{1}{2} \bar{\psi} \not{\partial} \psi + \frac{\lambda}{2} \sigma \bar{\psi} \not{\partial} \psi + \frac{\lambda^2}{8} \sigma^4$$

Bootstrapping  $\langle \sigma \sigma \sigma \sigma \rangle$  with 3D  $\mathcal{N} = 1$  SUSY (Rong, NS 2018)



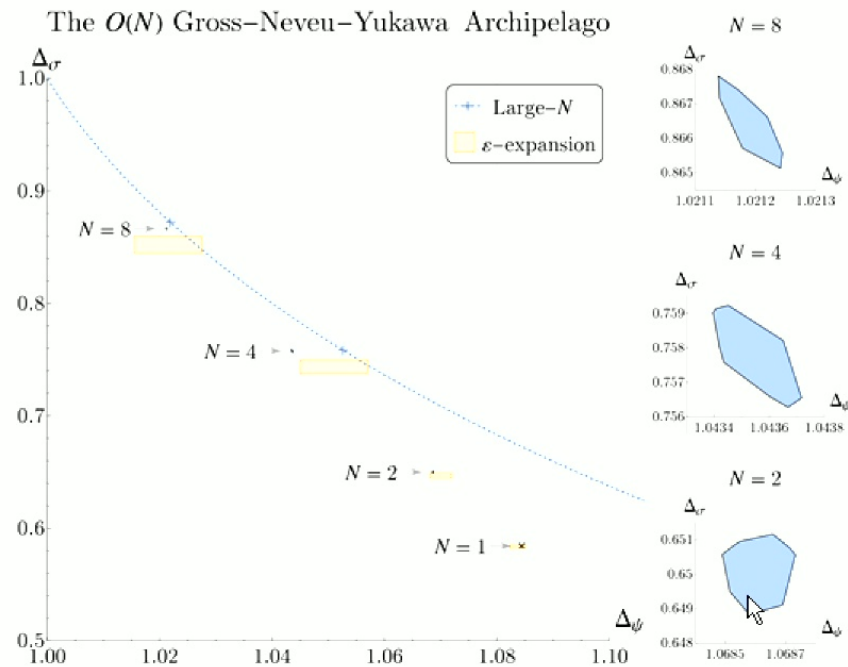
$$\Delta_\sigma = 0.584444 (30)$$

$$\eta = 0.168888 (60)$$

# Bootstrap 3D GNY

$$\mathcal{L}_{\text{GNY}} = \frac{1}{2} (\partial_\mu \phi)^2 + \frac{1}{2} \bar{\psi} \not{\partial} \psi + \frac{\lambda}{4} \phi^4 + \frac{g}{2} \phi \bar{\psi} \not{\partial} \psi$$

Bootstrapping all correlators involving  $\{\psi, \phi, \phi^2\}$



[Erramilli, Iliesiu, Kravchuk, Liu, Poland, Simmons-Duffin, 2022]

## Computational perspective

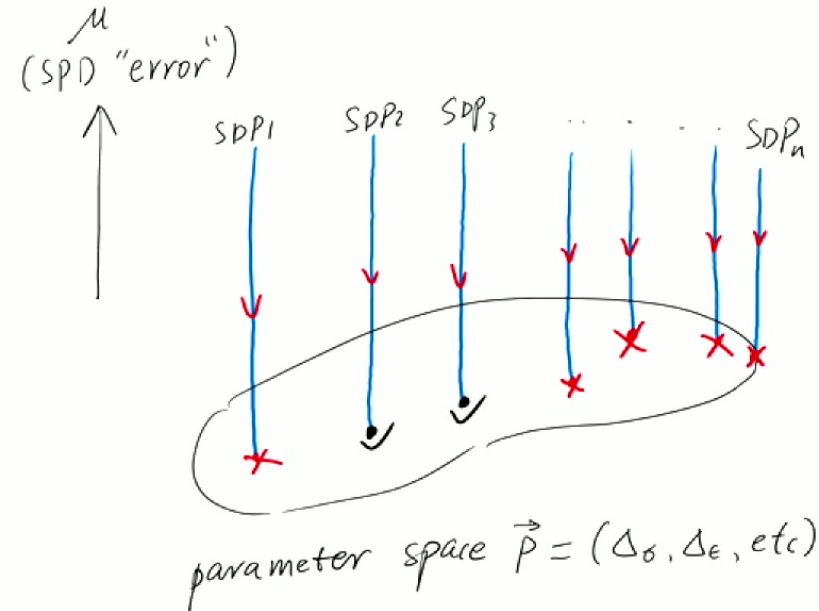
The formal theory of CFT : constraints are easy to understand, but hard to compute

- 1960s : “Bootstrap” ideas by Polyakov
- 1980s : Many results for 2D CFTs
- 2001 : Conformal blocks  $F_{\Delta, \ell}(u, v)$  in 2D and 4D are solved by Dolan, Osborn
- 2008-present : Effective numerical method for  $F_{\Delta, \ell}(u, v)$  in arbitrary dimension
  - Effective numerical method to check crossing symmetry (SDP approach)
  - Effective numerical method to scan theory space
  - ...
  - + Greatly improved analytical understanding of bootstrap equation, especially at large spin



## Computational perspective : bootstrap algorithms

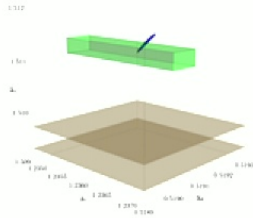
Choose a point in theory space (such as  $\Delta_\sigma$ )  $\implies$  Solving SDP  $\implies$  Repeat and carving out theory space



Algorithms : 1, Scan algorithm; 2, SDP solver algorithm

# Majors challenges in bootstrap numerics

- 1, The dimension of parameters could be high



→ Scan over  $\left\{ \Delta_\phi, \Delta_S, \Delta_t, \frac{\lambda_{\phi\phi t}}{\lambda_{\phi\phi S}}, \frac{\lambda_{tts}}{\lambda_{\phi\phi S}}, \frac{\lambda_{SSS}}{\lambda_{\phi\phi S}} \right\}$

Naive scan : cost  $\sim e^{\text{dimension}}$  (the curse of dimensionality)

- 2, Solving a SDP is slow, if SDP problem is too big

Ising :  $\langle \sigma\sigma\sigma\sigma \rangle, \langle \epsilon\epsilon\epsilon\epsilon \rangle, \langle \epsilon\epsilon\sigma\sigma \rangle$

$O(2)$  :  $\langle \phi\phi\phi\phi \rangle, \langle tttt \rangle, \langle ssss \rangle, \langle \phi s\phi s \rangle, \langle \phi\phi ts \rangle, \langle \phi\phi tt \rangle, \langle \phi\phi ss \rangle, \langle sstt \rangle$

$O(3)$  :  $\langle \phi\phi\phi\phi \rangle, \langle tttt \rangle, \langle ssss \rangle, \langle \phi s\phi s \rangle, \langle \phi\phi ts \rangle, \langle \phi\phi tt \rangle, \langle \phi\phi ss \rangle, \langle sstt \rangle, \langle sttt \rangle, \langle \phi\phi ts \rangle$

More complicated CFTs : need more correlators

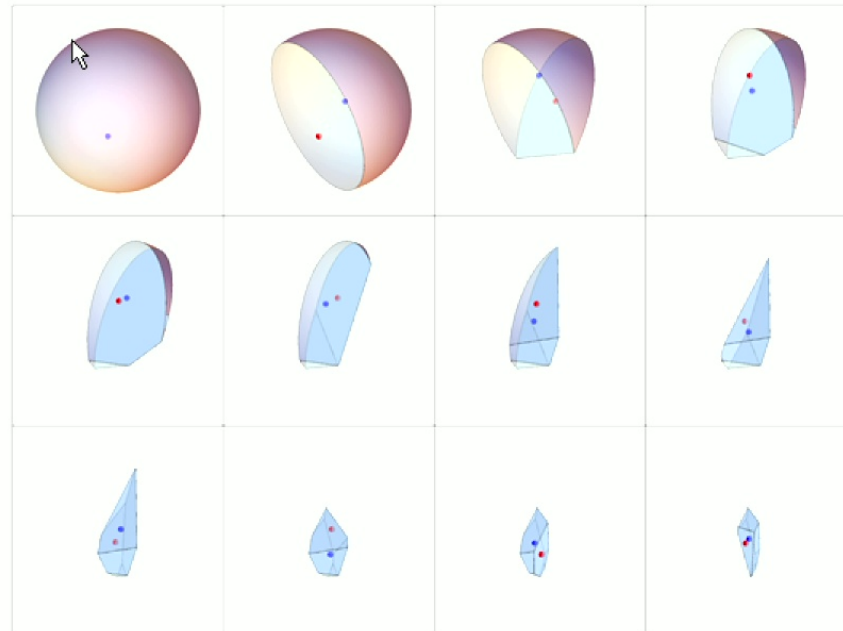
More powerful computational method → explore bigger set of constraints → access to more CFTs and data



# Cutting surface algorithm

Scan  $\left\{ \frac{\lambda_{\text{out}}}{\lambda_{\text{OOS}}}, \frac{\lambda_{\text{ts}}}{\lambda_{\text{OOS}}}, \frac{\lambda_{\text{SSS}}}{\lambda_{\text{OOS}}} \right\}$  : cutting surface algorithm : cost linear to number of OPE coeff

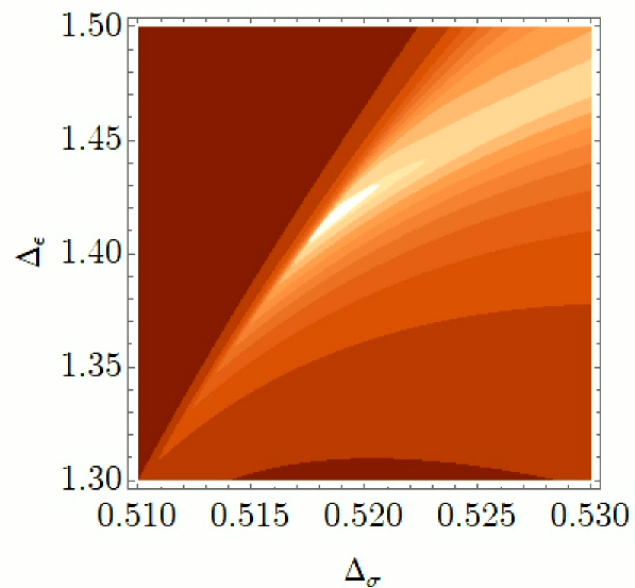
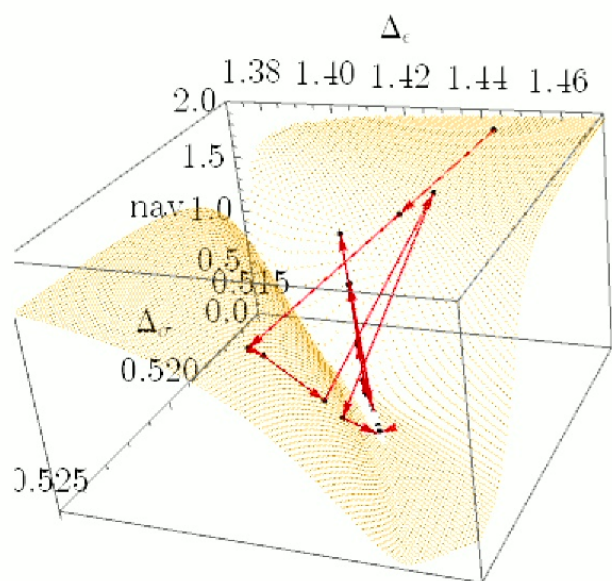
→ success in O(2), O(3), GNY



# The navigator function

“allowed/disallowed” → a continuous measure of success (navigator function)

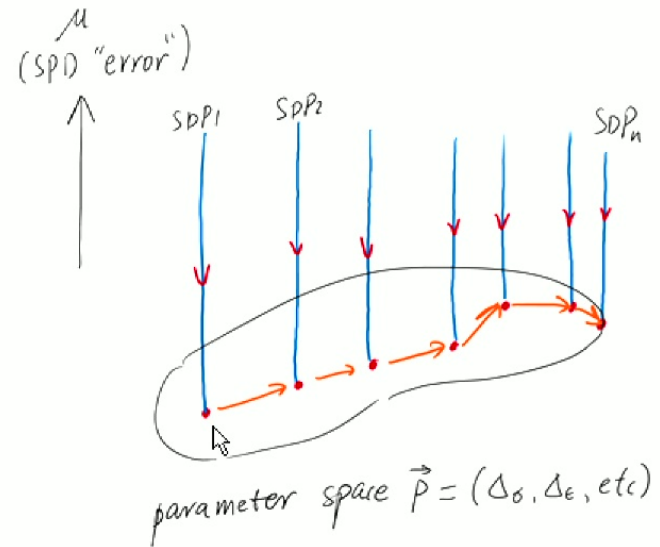
Unspecified  $\theta_{sc}$ ,  $\Lambda=11$



[Reehorst, Rychkov, Simmons-Duffin, Sirois, SN, van Rees 2021]

# Algorithm in navigator computation

A section of SDPs over a parameter space. We solve SDPs one by one.



Typical Newtonian iterations per SDP : 100 to 400



## The software

**scalar\_blocks, block3d** : generate conformal blocks

**SDPB, SDP\_skydiving** : solvers for SDP, dynamical SDP problems.

**autoboot** : derive bootstrap equation for many global symmetries

**Hyperion, simpleboot** : bootstrap frameworks that integrate scalarblocks, block3d, SDPB, SDP\_skydiving and various scan algorithms.

other software : **Juliboot, pyCFTboot, qboot**

## Plan for this week

☞ Morning : Lectures

☞ Afternoon : 1h Tutorial A

1h Break : running sample code, doing exercises and enjoying coffee

1h Tutorial B

1h Break : running sample code, doing exercises and enjoying coffee

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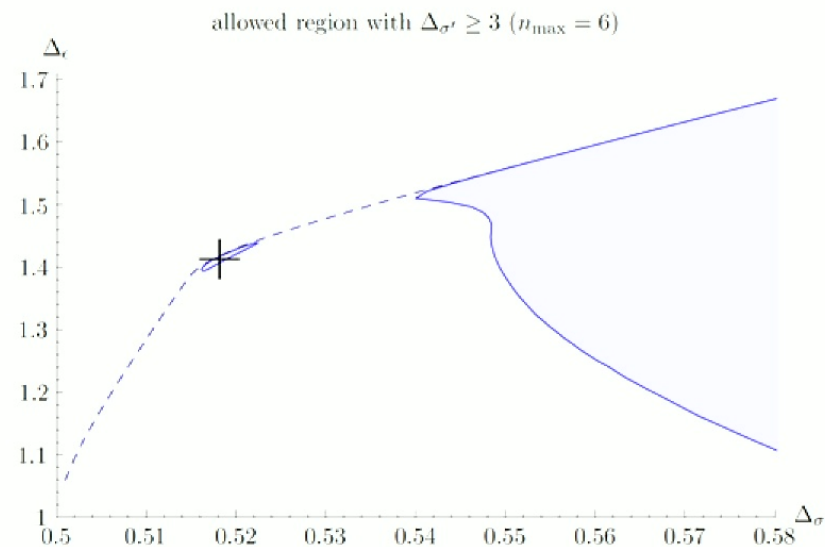
## Plan for this week

- Monday : 3D Ising  $\{\sigma, \epsilon\}$  system
- Tuesday : 3D  $O(N)$   $\{v, s, t\}$  system, cutting Surface algorithm
- Wednesday : 3D GNY model  $\{\psi, \phi, \epsilon\}$
- Thursday : Navigator method
- Friday : Skydiving algorithm

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# Today's goal : bootstrap 3D Ising

## Bootstrap 3D Ising $\{\sigma, \epsilon\}$ system



Reference : Weizmann Lectures on the Numerical Conformal Bootstrap. Shai M. Chester



## Bootstrap equations for scalars

Consider four different scalar  $\phi_1, \phi_2, \phi_3, \phi_4$ , the conformal partial wave decomposition is

$$\langle \overline{\phi_1(x_1)} \overline{\phi_2(x_2)} \overline{\phi_3(x_3)} \overline{\phi_4(x_4)} \rangle = \frac{1}{x_{12}^{\Delta_1+\Delta_2} x_{34}^{\Delta_3+\Delta_4}} \left( \frac{x_{24}}{x_{14}} \right)^{\Delta_{12}} \left( \frac{x_{14}}{x_{13}} \right)^{\Delta_{34}} \sum_O \lambda_{ijO} \lambda_{klO} g_{\Delta,l}^{\Delta_{12}, \Delta_{34}}(u, v), \quad \Delta_{ij} = \Delta_i - \Delta_j.$$

Crossing :  $\langle \overline{\phi_1 \phi_2 \phi_3 \phi_4} \rangle_{u,v} - \langle \overline{\phi_3 \phi_2 \phi_1 \phi_4} \rangle_{u,v} = 0$ . Under  $1 \leftrightarrow 3$  exchange,  $u \leftrightarrow v, z \leftrightarrow 1-z, \bar{z} \leftrightarrow 1-\bar{z}$

$$\sum_O \lambda_{12O} \lambda_{34O} F_{\pm, \Delta, l}^{12, 34} \mp \lambda_{32O} \lambda_{14O} F_{\pm, \Delta, l}^{32, 14} = 0 \text{ where } F_{\pm, \Delta, l}^{12, 34}(u, v) = v^{\frac{\Delta_2+\Delta_3}{2}} g_O^{\Delta_{12}, \Delta_{34}}(u, v) \pm u^{\frac{\Delta_2+\Delta_3}{2}} g_O^{\Delta_{12}, \Delta_{34}}(v, u)$$

Denote  $\langle \overline{\phi_1 \phi_2 \phi_3 \phi_4} \rangle = \langle \overline{\phi_3 \phi_2 \phi_1 \phi_4} \rangle$  simply as  $\langle 1234 \rangle$ . Only  $\langle 1234 \rangle, \langle 1324 \rangle, \langle 1243 \rangle$  are independent equations. If  $\phi_1 = \phi_2, \phi_3 = \phi_4$ , only  $\langle 1144 \rangle, \langle 1414 \rangle$  are independent equations.

Consider CFT with  $\mathbb{Z}_2$  symmetry. Assuming  $\phi_1$  is  $\mathbb{Z}_2$  odd,  $\phi_2$  is  $\mathbb{Z}_2$  even.

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## Bootstrap equations for $\mathbb{Z}_2$ even/odd scalars

Consider CFT with  $\mathbb{Z}_2$  symmetry. Assuming  $\phi_1$  is  $\mathbb{Z}_2$  odd,  $\phi_2$  is  $\mathbb{Z}_2$  even. Independent bootstrap equations :

$$(1), \langle \phi_1 \phi_1 \phi_1 \phi_1 \rangle : \\ \sum_o \lambda_{11o}^2 F_{-\Delta, \ell}^{11,11} = 0$$

$$(2), \langle \phi_2 \phi_2 \phi_2 \phi_2 \rangle : \\ \sum_o \lambda_{22o}^2 F_{-\Delta, \ell}^{22,22} = 0$$

$$(3), \langle \phi_1 \phi_2 \phi_1 \phi_2 \rangle : \\ \sum_o \lambda_{12o}^2 F_{-\Delta, \ell}^{12,12} = 0$$

$$(4)(5), \langle \phi_1 \phi_1 \phi_2 \phi_2 \rangle : \\ \sum_o \lambda_{21o} \lambda_{12o} F_{\mp, \Delta, \ell}^{21,12} \pm \lambda_{11o} \lambda_{22o} F_{\mp, \Delta, \ell}^{11,22} = 0$$

$$(\lambda_{21o} = (-1)^{\ell} \lambda_{12o})$$

$$V_{\text{even}} = \begin{pmatrix} \begin{pmatrix} F_{-\Delta, \ell}^{11,11} & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & F_{-\Delta, \ell}^{22,22} \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} F_{-\Delta, \ell}^{11,22} \\ \frac{1}{2} F_{-\Delta, \ell}^{11,22} & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} F_{+\Delta, \ell}^{11,22} \\ \frac{1}{2} F_{+\Delta, \ell}^{11,22} & 0 \end{pmatrix} \end{pmatrix}, \quad V_{\text{odd}} = \begin{pmatrix} 0 \\ 0 \\ F_{-\Delta, \ell}^{12,12} \\ (-1)^{\ell} F_{-\Delta, \ell}^{21,12} \\ -(-1)^{\ell} F_{-\Delta, \ell}^{21,12} \end{pmatrix}$$

$$\sum_{o^+} (\lambda_{11o} \quad \lambda_{22o}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{11o} \\ \lambda_{22o} \end{pmatrix} + \sum_{o^-} \lambda_{12o}^2 V_{\text{odd}} = 0$$

## Bootstrap equations for $Z_2$ even/odd scalars

The feasibility test for  $\sum_{O^+} (\lambda_{11O} \ \lambda_{22O}) \cdot V_{\text{even}, \Delta_{O^+}, \ell_{O^+}} \cdot \begin{pmatrix} \lambda_{11O} \\ \lambda_{22O} \end{pmatrix} + \sum_{O^-} \lambda_{12O}^2 V_{\text{odd}, \Delta_{O^-}, \ell_{O^-}} = 0$

Let's make some assumptions about spectrum (thus  $V_{\text{even}}$ ,  $V_{\text{odd}}$  vectors), and test whether our assumptions is true.

Positivity test : find a linear functional  $\vec{\alpha} = (\alpha_1, \alpha_2, \alpha_3, \alpha_4, \alpha_5)$  such that

$$\alpha \cdot V_{\text{even}} \geq 0$$

$$\alpha \cdot V_{\text{odd}} \geq 0$$

Note:  $\alpha \cdot V_{\text{even}}$  is a  $2 \times 2$  matrix.  $\alpha \cdot V_{\text{even}} \geq 0$  means the matrix is positive semi-definite, i.e.

$$(\lambda_{11O} \ \lambda_{22O}) \alpha \cdot V_{\text{even}} \begin{pmatrix} \lambda_{11O} \\ \lambda_{22O} \end{pmatrix} \geq 0 \text{ for arbitrary } \lambda_{11O}, \lambda_{22O}$$

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## Bootstrap 3D Ising CFT $\{\sigma, \epsilon\}$

$$\sum_{O^+} (\lambda_{11O} \quad \lambda_{22O}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{11O} \\ \lambda_{22O} \end{pmatrix} + \sum_{O^-} \lambda_{12O}^2 V_{\text{odd}} = 0$$

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isolate identity,  $\sigma, \epsilon$



$$(1 \quad 1) \cdot V_{\text{even}, \Delta=0, \ell=0} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix} + (\lambda_{\sigma\sigma\epsilon} \quad \lambda_{\epsilon\epsilon\epsilon}) \cdot V_{\text{even}, \Delta=\Delta_\epsilon, \ell=0} \cdot \begin{pmatrix} \lambda_{\sigma\sigma\epsilon} \\ \lambda_{\epsilon\epsilon\epsilon} \end{pmatrix} + \lambda_{\sigma\epsilon\sigma}^2 V_{\text{odd}, \Delta=\Delta_\sigma, \ell=0}$$

$$\sum_{O^+ \neq 1, \epsilon} (\lambda_{\sigma\sigma O} \quad \lambda_{\epsilon\epsilon O}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{\sigma\sigma O} \\ \lambda_{\epsilon\epsilon O} \end{pmatrix} + \sum_{O^- \neq \sigma} \lambda_{12O}^2 V_{\text{odd}} = 0$$

Assume  $\lambda_{\sigma\epsilon\sigma} = \lambda_{\sigma\sigma\epsilon}$



$$V_{\text{identity}} + (\lambda_{\sigma\sigma\epsilon} \quad \lambda_{\epsilon\epsilon\epsilon}) \cdot V_\theta \cdot \begin{pmatrix} \lambda_{\sigma\sigma\epsilon} \\ \lambda_{\epsilon\epsilon\epsilon} \end{pmatrix} + \sum_{O^+ \neq 1, \epsilon} (\lambda_{\sigma\sigma O} \quad \lambda_{\epsilon\epsilon O}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{\sigma\sigma O} \\ \lambda_{\epsilon\epsilon O} \end{pmatrix} + \sum_{O^- \neq \sigma} \lambda_{12O}^2 V_{\text{odd}} = 0$$

$$V_{\text{identity}} \equiv (1 \quad 1) \cdot V_{\text{even}, \Delta=0, \ell=0} \cdot \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad V_\theta \equiv V_{\text{even}, \Delta=\Delta_\epsilon, \ell=0} + \begin{pmatrix} V_{\text{odd}, \Delta=\Delta_\sigma, \ell=0} & 0 \\ 0 & 0 \end{pmatrix}$$

# Unitarity

**Unitarity:** all the states in Hilbert space have positive norm.

**Consequences 1 :**  $\Delta \geq \Delta_{\text{unitary}}$  ,  $\Delta_{\text{unitary}} = \begin{cases} d/2 - 1 & \text{for spin} = 0 \\ d - 2 + \ell & \text{for spin} = \ell \end{cases}$

by demanding all descendant state has positive norm as well.

Example :  $\|P_\mu |O\rangle\|^2 = 2\Delta \| |O\rangle\|^2$  for scalar state  $|O\rangle$  , thus  $\Delta \geq 0$

**Consequences 2 :**  $\lambda_{\phi_1 \phi_2 O}$  is real

For Hermitian operators, the OPE coefficients are always real.

Proof: take Hermitian conjugation on  $\langle \phi(x) \phi(y) O^{\mu_1 \dots \mu_\ell}(z) \rangle = \lambda_{\phi\phi O} \frac{Z^{\mu_1} \dots Z^{\mu_\ell} \text{-trace}}{|x-y|^{\Delta_i + \Delta_j - (\Delta_k - \ell)} |y-z|^{\Delta_j + (\Delta_k - \ell) - \Delta_i} |z-x|^{(\Delta_k - \ell) + \Delta_i - \Delta_j}}$

If some states have negative norm :  $\langle \phi(x) \phi(y) \rangle = -\frac{1}{|x-y|^{2\Delta_\phi}}$  . If we insist to normalize it as  $\langle \phi(x) \phi(y) \rangle = \frac{1}{|x-y|^{2\Delta_\phi}}$  (as we did in defining conformal block), then  $\phi \rightarrow i\phi$  makes  $\phi$  anti-Hermitian and some OPE coefficients are imaginary.

## Assumption for 3D Ising spectrum

We choose two numbers  $\Delta_\sigma, \Delta_\epsilon$

Assumption on the CFT data:

1,  $\sigma$  has scaling dimension  $\Delta_\sigma$

2,  $\epsilon$  has scaling dimension  $\Delta_\epsilon$

3, all other spin 0 operators has dimension larger than 3

4, all spinning operators has dimension  $\Delta \geq \Delta_{\text{unitary}}$  ,  $\Delta_{\text{unitary}} = \begin{cases} d/2 - 1 & \text{for spin} = 0 \\ d - 2 + \ell & \text{for spin} = \ell \end{cases}$

5, all OPE coefficients are real

Problem to solve : does the choice of two numbers  $\Delta_\sigma, \Delta_\epsilon$  + assumptions compatible with

$$V_{\text{identity}} + (\lambda_{\sigma\sigma\epsilon} \ \lambda_{\epsilon\epsilon\epsilon}) \cdot V_\theta \cdot \begin{pmatrix} \lambda_{\sigma\sigma\epsilon} \\ \lambda_{\epsilon\epsilon\epsilon} \end{pmatrix} + \sum_{O^+ \neq 1, \epsilon} (\lambda_{\sigma\sigma O} \ \lambda_{\epsilon\epsilon O}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{\sigma\sigma O} \\ \lambda_{\epsilon\epsilon O} \end{pmatrix} + \sum_{O^- \neq \sigma} \lambda_{12} o^2 V_{\text{odd}} = 0 ?$$

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## Numerical details

$$V_{\text{identity}} + (\lambda_{\sigma\sigma\epsilon} \ \lambda_{\epsilon\epsilon\epsilon}) \cdot V_{\theta} \cdot \begin{pmatrix} \lambda_{\sigma\sigma\epsilon} \\ \lambda_{\epsilon\epsilon\epsilon} \end{pmatrix} + \sum_{O^+} (\lambda_{\sigma\sigma O} \ \lambda_{\epsilon\epsilon O}) \cdot V_{\text{even}} \cdot \begin{pmatrix} \lambda_{\sigma\sigma O} \\ \lambda_{\epsilon\epsilon O} \end{pmatrix} + \sum_{O^-} \lambda_{12O}^2 V_{\text{odd}} = 0$$

Find a linear functional  $\vec{\alpha} = (\alpha_1, \alpha_2, \alpha_3, \alpha_4, \alpha_5)$  such that

$$\alpha \cdot (V_{\text{identity}}) = 1$$

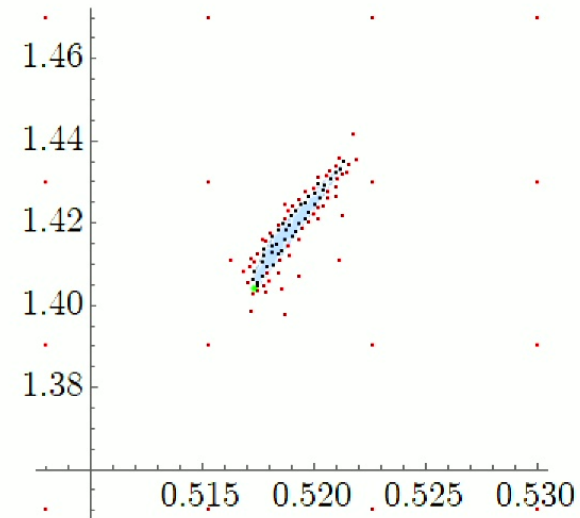
$$\alpha \cdot (V_{\theta}) \geq 0$$

$$\alpha \cdot V_{\text{even}}(\Delta, \ell) \geq 0 \text{ for } \Delta \geq 3, \ell = 0$$

$$\alpha \cdot V_{\text{odd}}(\Delta, \ell) \geq 0 \text{ for } \Delta \geq 3, \ell = 0$$

$$\alpha \cdot V_{\text{even}}(\Delta, \ell) \geq 0 \text{ for } \Delta \geq \Delta_{\text{unitary}}, \ell = 2, 4, 6, \dots$$

$$\alpha \cdot V_{\text{odd}}(\Delta, \ell) \geq 0 \text{ for } \Delta \geq \Delta_{\text{unitary}}, \ell = 1, 2, 3, 4, \dots$$



## Numerical details

1, We truncate  $\alpha$  as  $\alpha = \sum_{n+m \leq \Lambda} \alpha_{m,n} \partial_z^m \partial_{\bar{z}}^n \Big|_{z=\bar{z}=\frac{1}{2}}$

2,  $F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}$  can be approximated as  $\partial_z^m \partial_{\bar{z}}^n F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}(z, \bar{z}) \Big|_{z=\bar{z}=\frac{1}{2}} \approx \chi_\ell(\Delta) \times P_{\Delta,\ell}^{m,n}(\Delta)$  ,  $\chi_\ell(\Delta) = \frac{(3-2\sqrt{2})^\Delta}{(\Delta-r_1)(\Delta-r_2)\dots(\Delta-r_K)}$

$P_{\Delta,\ell}^{m,n}(\Delta)$  : polynomial in  $\Delta$  , can be computed using **scalar\_blocks**  $\mathbb{I}$

$\chi_\ell(\Delta)$  are the same for all  $F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}$  terms in  $V_{\text{even}}$ ,  $V_{\text{odd}}$  . All poles are below unitary bound.

$$\alpha \cdot V_{\text{even},\Delta,\ell} = \chi_\ell(\Delta) \sum_{m,n} \alpha_{mn} \cdot \begin{pmatrix} \begin{pmatrix} (P_{-\Delta,\ell}^{11,11})^{mn} & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & (P_{-\Delta,\ell}^{22,22})^{mn} \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} (P_{-\Delta,\ell}^{11,22})^{mn} \\ \frac{1}{2} (P_{-\Delta,\ell}^{11,22})^{mn} & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} (F_{+\Delta,\ell}^{11,22})^{mn} \\ \frac{1}{2} (P_{+\Delta,\ell}^{11,22})^{mn} & 0 \end{pmatrix} \end{pmatrix} \rightarrow \alpha \cdot V_{\text{even},\Delta,\ell}^{(\text{poly})} \geq 0$$



## Constraints for conformal field theory

2, Operator Product Expansion :  $\phi_i(x) \phi_j(y) = \sum_k \lambda_{ijk} C_k(x-y, \partial_y) \mathcal{O}_k(y)$

↓  
Operator Product Expansion (OPE) coefficients

conformal symmetry fix  $C_k(x-y, \partial_y)$  :  $\langle \phi_i(x) \phi_j(y) \mathcal{O}_k(z) \rangle = \lambda_{ijk} C_k(x-y, \partial_y) \langle \mathcal{O}_k(y) \mathcal{O}_k(z) \rangle$

$$\begin{aligned} \langle \overline{\phi(x_1)} \phi(x_2) \overline{\phi(x_3)} \phi(x_4) \rangle &= \sum \lambda_{12} \lambda_{34} C_a(x_1-x_2, \partial_2) C_b(x_3-x_4, \partial_4) \langle \mathcal{O}^a(x_2) \mathcal{O}^b(x_4) \rangle \\ &= \sum \lambda_{12} \lambda_{34} x_{12}^{-\Delta_a} x_{34}^{-\Delta_b} g_{\Delta, \ell}(u, v) \end{aligned}$$

$$u = z \bar{z} = \frac{x_{12}^2 x_{34}^2}{x_{13}^2 x_{24}^2}, \quad v = (1-z)(1-\bar{z}) = \frac{x_{14}^2 x_{23}^2}{x_{13}^2 x_{24}^2}$$

## Numerical details

1, We truncate  $\alpha$  as  $\alpha = \sum_{n+m \leq \Lambda} \alpha_{m,n} \partial_z^m \partial_{\bar{z}}^n \Big|_{z=\bar{z}=\frac{1}{2}}$

2,  $F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}$  can be approximated as  $\partial_z^m \partial_{\bar{z}}^n F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}(z, \bar{z}) \Big|_{z=\bar{z}=\frac{1}{2}} \approx \chi_\ell(\Delta) \times P_{\Delta,\ell}^{m,n}(\Delta)$  ,  $\chi_\ell(\Delta) = \frac{(3-2\sqrt{2})^\Delta}{(\Delta-r_1)(\Delta-r_2)\dots(\Delta-r_K)}$

$P_{\Delta,\ell}^{m,n}(\Delta)$  : polynomial in  $\Delta$  , can be computed using **scalar\_blocks**

$\chi_\ell(\Delta)$  are the same for all  $F_{\Delta,\ell}^{\Delta_{12},\Delta_{34}}$  terms in  $V_{\text{even}}$ ,  $V_{\text{odd}}$  . All poles are below unitary bound.

$$\alpha \cdot V_{\text{even},\Delta,\ell} = \chi_\ell(\Delta) \sum_{m,n} \alpha_{mn} \cdot \begin{pmatrix} \begin{pmatrix} (P_{-,\Delta,\ell}^{11,11})^{mn} & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & (P_{-,\Delta,\ell}^{22,22})^{mn} \end{pmatrix} \\ \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} (P_{-,\Delta,\ell}^{11,22})^{mn} \\ \frac{1}{2} (P_{-,\Delta,\ell}^{11,22})^{mn} & 0 \end{pmatrix} \\ \begin{pmatrix} 0 & \frac{1}{2} (F_{+,\Delta,\ell}^{11,22})^{mn} \\ \frac{1}{2} (P_{+,\Delta,\ell}^{11,22})^{mn} & 0 \end{pmatrix} \end{pmatrix} \longrightarrow \alpha \cdot V_{\text{even},\Delta,\ell}^{(\text{poly})} \geq 0$$

## Numerical details

3, For both  $\alpha \cdot V_{\text{even}, \Delta, \ell}^{(\text{poly})} \geq 0$  and  $\alpha \cdot V_{\text{odd}, \Delta, \ell}^{(\text{poly})} \geq 0$ , we demand the condition for  $\ell = 0, \dots, \ell_{\max}$

The problem become polynomial programming, PMP

Find  $\alpha$  such that

$$\sum_k \alpha_k M_0^k(x) = 1$$

$$\sum_k \alpha_k M_j^k(x) > 0 \text{ for } x > 0 \text{ for } j = 1, \dots, j_{\max}$$

$$M_j^k(x) \leftrightarrow V_{\text{even/odd}, \Delta = \Delta_{\min} + \mathbf{1}, \ell}$$

# SDPB : input

$sdp \equiv \text{SDP}[\langle \text{objective} \rangle, \langle \text{normalization} \rangle, \langle \text{positive matrices with prefactors} \rangle]$

$\text{objective} \equiv \{a_0, \dots, a_N\}$

$\text{normalization} \equiv \{n_0, \dots, n_N\}$

$\text{positive matrices with prefactors} \equiv \{$   
 $\langle \text{positive matrix with prefactor } 1 \rangle,$   
 $\dots$   
 $\langle \text{positive matrix with prefactor } J \rangle,$   
 $\}$

$\text{positive matrix with prefactor } j \equiv$   
 $\text{PositiveMatrixWithPrefactor}[\langle \text{prefactor} \rangle,$   
 $\{$   
 $\{$   
 $\{Q_{j,11}^0(x), \dots, Q_{j,11}^N(x)\}, \dots, \{Q_{j,m_j 1}^0(x), \dots, Q_{j,m_j 1}^N(x)\}$   
 $\},$   
 $\dots$   
 $\{$   
 $\{Q_{j,1m_j}^0(x), \dots, Q_{j,1m_j}^N(x)\}, \dots, \{Q_{j,m_j m_j}^0(x), \dots, Q_{j,m_j m_j}^N(x)\}$   
 $\},$   
 $\}$   
 $\}$

Maximize  $\alpha \cdot \text{objective}$ , such that

$\alpha \cdot \text{normalization} = 1$

$\alpha \cdot M_j(x) \geq 0$  for  $x \geq 0$  for  $j = 1, 2, \dots, J$

$$M_j = \begin{pmatrix} M_j^0 \\ M_j^1 \\ \vdots \\ M_j^N \end{pmatrix}$$

$$M_j^n = \begin{pmatrix} (M_j^n)_{11} & \dots & (M_j^n)_{1m_j} \\ \vdots & \ddots & \vdots \\ (M_j^n)_{m_j 1} & \dots & (M_j^n)_{m_j m_j} \end{pmatrix}$$

$$(M_j^n)_{a b} = Q_{j,a b}^n(x)$$

## SDPB in feasibility mode

Set objective = 0 . Ask whether  $\alpha$  exist for the condition

$$\sum_k \alpha_k M_0^k(x) = 1$$

$$\sum_k \alpha_k M_j^k(x) > 0 \text{ for } x > 0 \text{ for } j = 1, \dots, j_{\max}$$

Output : I

“dual feasible jump detected”, “find dual feasible solution” : find  $\alpha$

“primal feasible jump detected”, “find primal feasible solution” : can't find  $\alpha$

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“primal feasible jump detected”, “find primal feasible solution” : can't find  $\alpha$

## SDPB in optimality mode : bound OPE coefficient

$$\sum_{O \in \phi \times \phi} \lambda_{\phi\phi O}^2 F_{\Delta, \ell}(u, v) = 0 \quad \longrightarrow \quad \lambda_{O_0}^2 F_{\Delta_0, \ell_0}(u, v) = -F_{0,0}(u, v) - \sum_O \lambda_O^2 F_{\Delta, \ell}(u, v)$$

Find a linear functional  $\alpha$  such that

$$\alpha(F_{\Delta_0, \ell_0}(u, v)) = 1 \quad (1)$$

$$\alpha(F_{\Delta, \ell}(u, v)) \geq 0 \text{ for } \Delta > \Delta_{\min} \quad (2)$$

If a  $\alpha$  exist, we find an inequality

$$\lambda_{O_0}^2 = -\alpha(F_{0,0}(u, v)) - \sum_O \lambda_O^2 \alpha(F_{\Delta, \ell}(u, v)) \leq -\alpha(F_{0,0}(u, v))$$

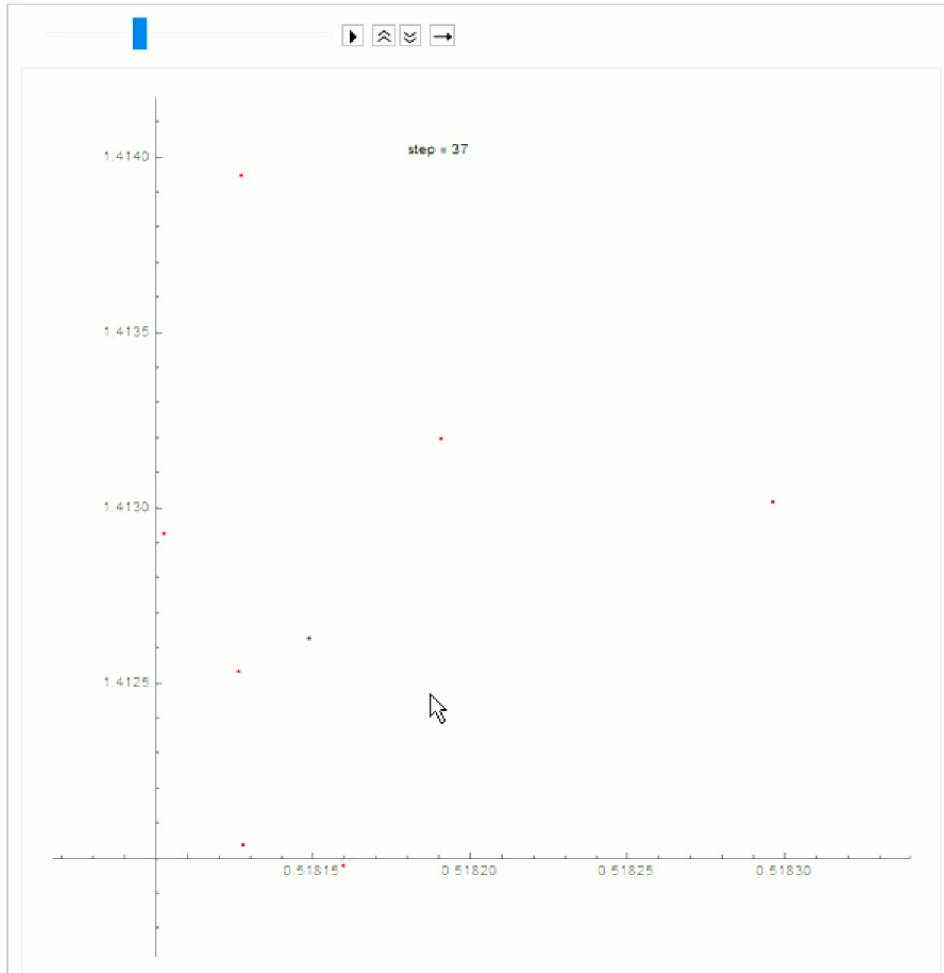
To the most restrictive bound, we need  $\alpha$  that minimizing  $-\alpha(F_{0,0}(u, v))$  subject to (1), (2)

Such that  $\alpha$  should satisfy  $\alpha(F_{\Delta, \ell}(u, v)) = 0$  for  $\Delta, \ell$  of each  $O$  in the  $\sum_{O \in \phi \times \phi} \dots$

Zeros in  $\alpha(F_{\Delta, \ell}(u, v)) \longrightarrow$  physical spectrum  $\Delta$  (**Extremal Functional Method, EFM**)

✕

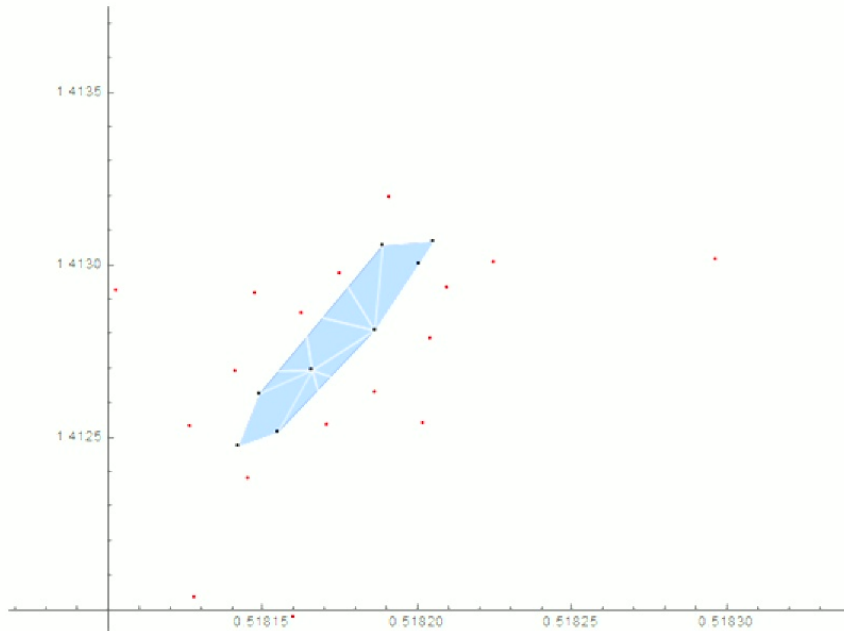
# Delaunay search





# Delaunay search

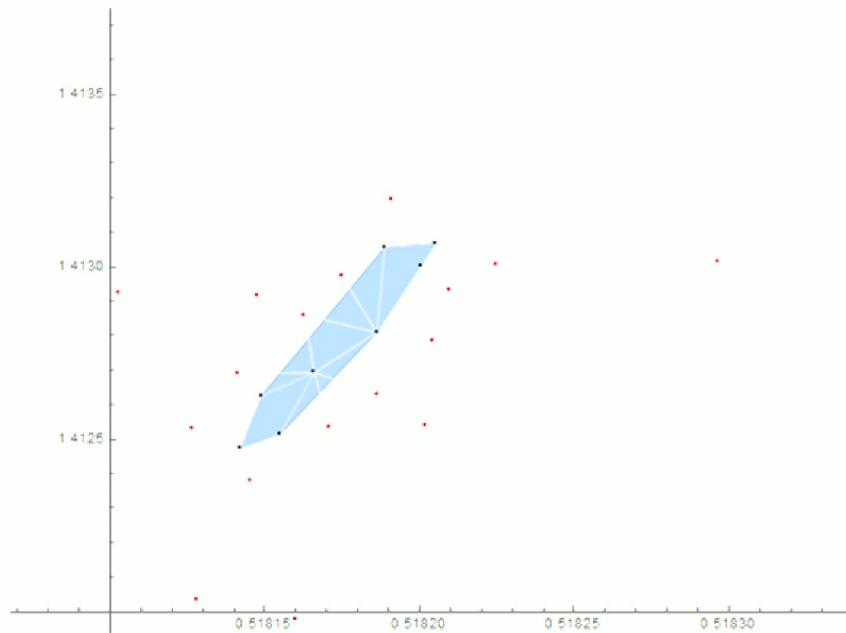
Delaunay search : an adaptive method to scan the boundary of a region  
Assume we have a set allowed points and a set of disallowed points



- 1, Find a triangulation connecting all points
- 2, Find boundary triangles
- 3, Rank the triangles by areas;  
find the largest  $n$  triangles;  
scan their middle points.

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