

Relative Entropy and Proximity of QFTs

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hep-th/14???.???? w/ V. Balasubramanian and A. Maloney

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Motivation

Given a string compactification, we get:

- A collection of fields $\{\phi\}$
- An effective action $S_{eff}[\phi]$

Question: How well can we determine:

A list of “nearby” string compactifications

Related: What about the spacetime?

JJH '13, Hebecker '13

Related Motivation

String theory yields a landscape of EFTs

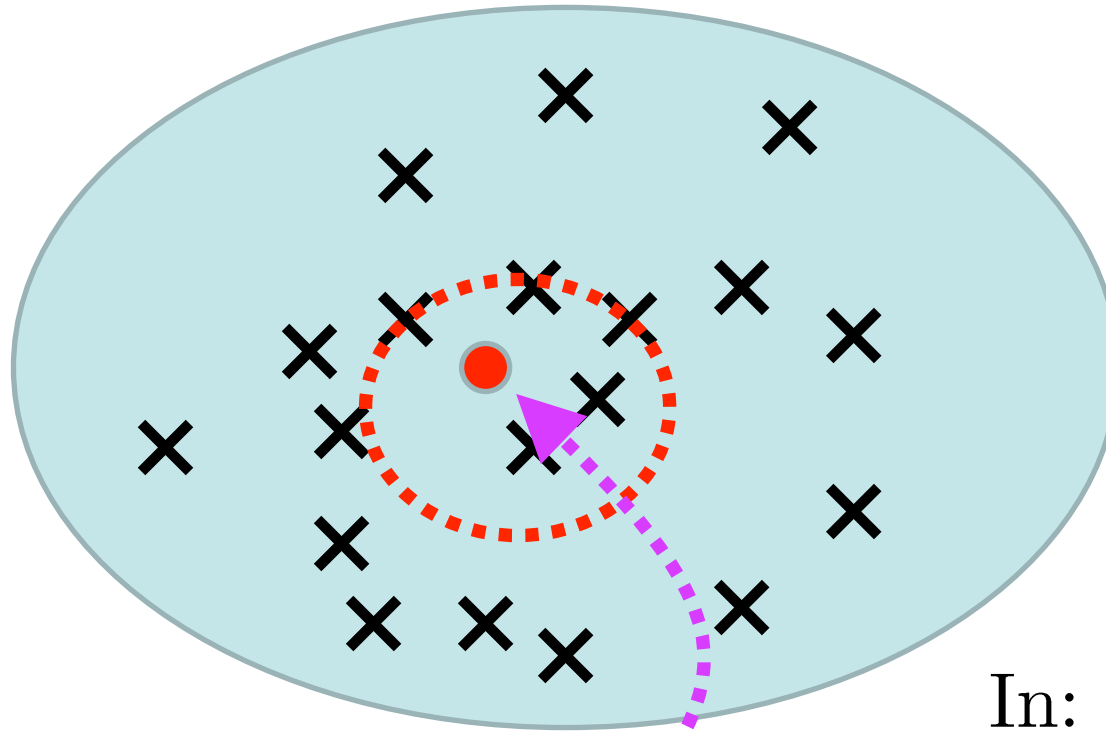
$$EFT_1, \dots, EFT_{10^{500}}$$

Question: Is there a notion of distance:

$$Distance(EFT_i, EFT_j) \geq 0?$$

Douglas, “Spaces of Quantum Field Theories” 1005.2779

Why a Distance Would Help



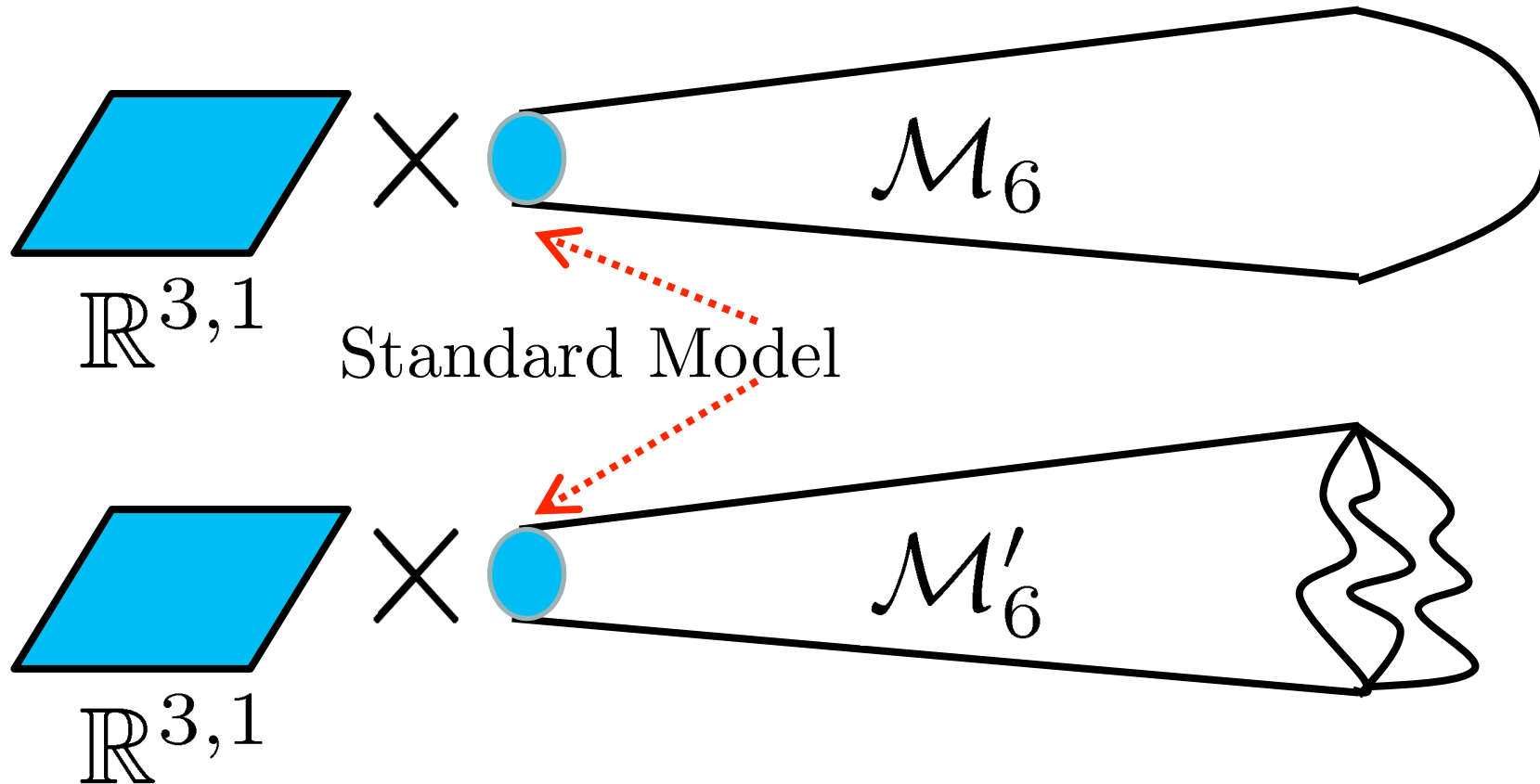
We are here

In: "Close by"

Out: "Far Away"

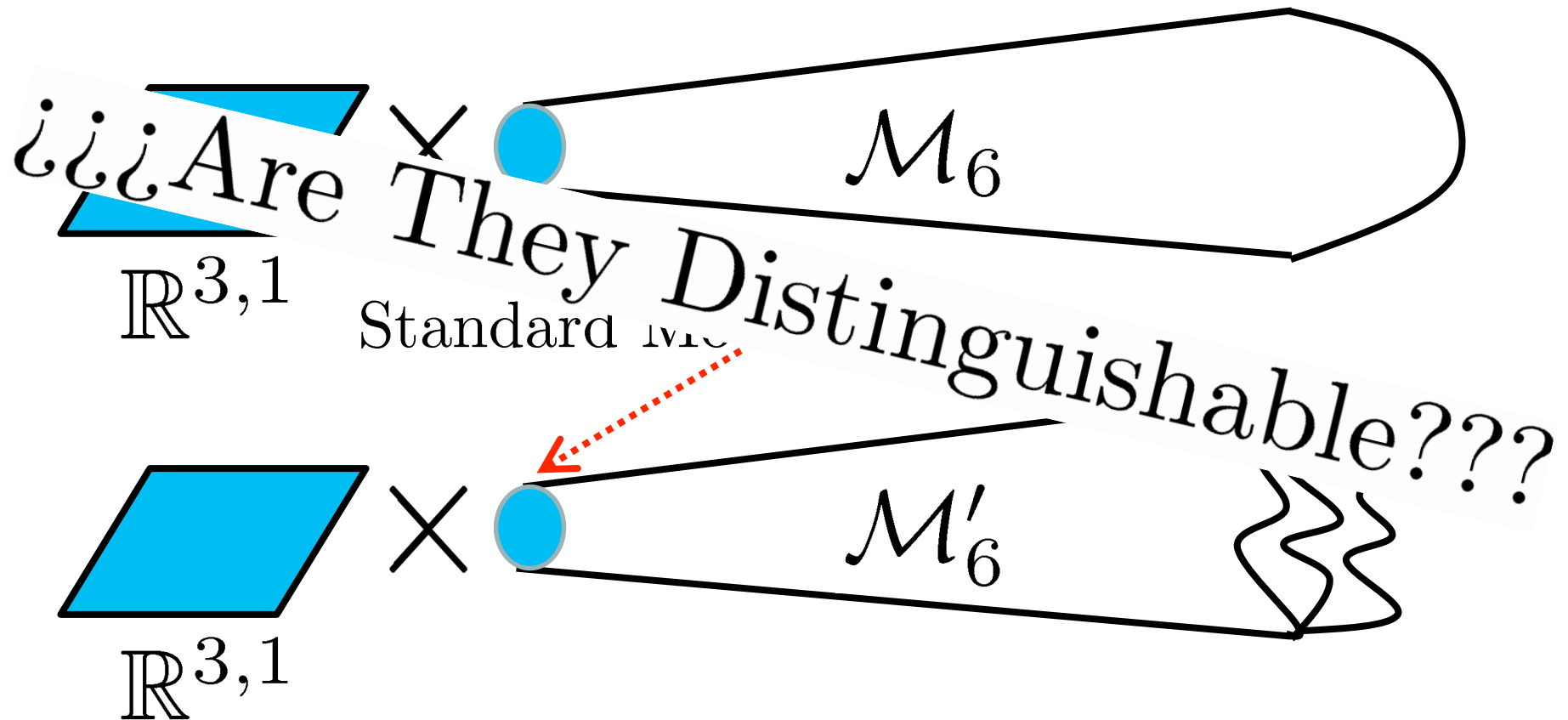
Further Motivation

A local model may have different global embeddings



Further Motivation

A local model may have different global embeddings



Basic Idea

We ask a basic question:

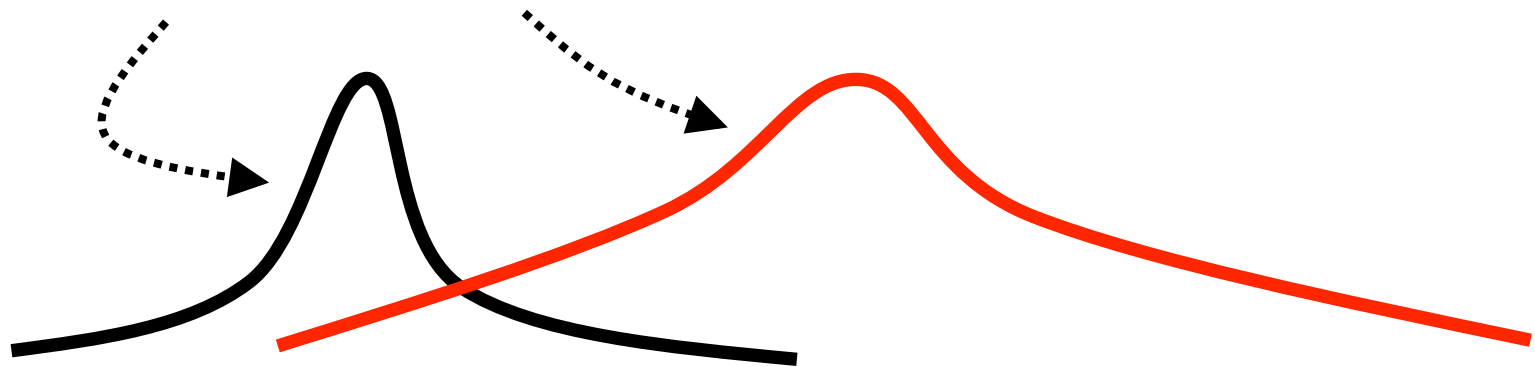
How many measurements would it take
to tell the difference between
 $S^{(1)}[\phi]$ and $S^{(2)}[\phi]$?

If you can't tell the difference,
compactifications are indistinguishable
 \Rightarrow Coarse graining on the landscape

Info and Distinguishability

This issue has been studied in the context of statistical inference and information theory

Suppose $p(z)$ and $q(z)$ two prob. distributions:



Odds of thinking we sampled from p rather than q ?

Relative Entropy

Suppose $p(z)$ and $q(z)$ two prob. distributions:

Shannon Entropy: $-\int dz p(z) \log p(z)$

Kullback-Leibler Divergence / Relative Entropy:

$$D_{KL}(p||q) \equiv \int dz p(z) \log \frac{p(z)}{q(z)}$$

Properties of D_{KL}

- $D_{KL}(p||q) \geq 0$
- $D_{KL}(p||q) = 0$ iff $p = q$ almost surely
- measured in “nats” rather than “bits”

Interpretations of D_{KL}

Learning:

If $p(z)$ is the “true” dist, but we think it’s $q(z)$

$D_{KL}(p||q)$ = info we’d gain from learning $p(z)$

Chernoff Bound:

Sample N times from $q(z)$

$\Pr[\text{gen}^{ed} \text{ from } p(z)] \leq \exp(-ND_{KL}(p||q))$

Proximity

$D_{KL}(p||q)$ says how “close” q and p are

But it’s not a metric... (not symmetric)

Infinitesimal Version: $p(z) = q(z|\lambda) + \frac{\partial q}{\partial \lambda^i} \delta \lambda^i$

$$D_{KL}(p||q) = G_{ij} \delta \lambda^i \delta \lambda^j$$

$$\text{Info Metric: } G_{ij} = \int dz q \frac{\partial \log q}{\partial \lambda^i} \frac{\partial \log q}{\partial \lambda^j}$$

The Proposal (I / II)

V. Balasubramanian, JJH, A. Maloney, to appear

Consider Euclidean Signature theory

Suppose we know $S[\phi]$

This defines a probability distribution:

$p[\phi] = \frac{\exp(-S[\phi])}{Z}$, i.e. a Boltzmann factor

Partition Function: $Z = \int \mathcal{D} \phi \exp(-S[\phi])$

The Proposal (II / II)

Suppose we have two theories, i.e. two distns:

$$p[\phi] = \frac{\exp(-S_p[\phi])}{Z_p} \quad \text{and} \quad q[\phi] = \frac{\exp(-S_q[\phi])}{Z_q}$$

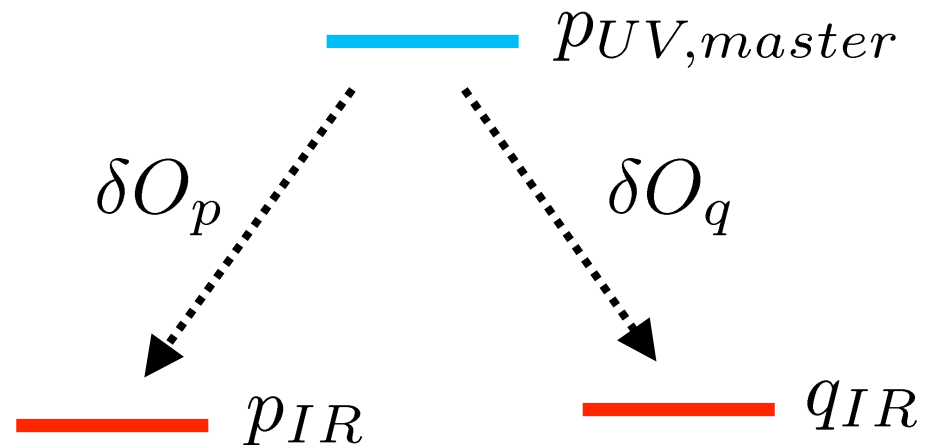
$$\text{Proximity} = D_{KL}(p||q) = \int \mathcal{D}\phi p[\phi] \log \left(\frac{p[\phi]}{q[\phi]} \right)$$

“Sample” = Field configuration in spacetime

Note...

p and q could even have different field content!

RG as motion on space of couplings



\Rightarrow Just need a “master theory” (e.g. strings)

Perturbation Theory

Perturbation Theory

Suppose $S_p - S_q = \int d^D x \delta\lambda^i O_i(x) \equiv \Theta$

To leading order in perturbation theory, we get:

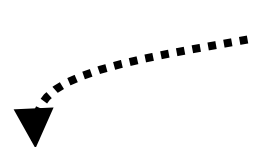
$$D_{KL}(p||q) = \langle (\Theta - \langle \Theta \rangle_p)(\Theta - \langle \Theta \rangle_p) \rangle_p$$

Regulators...

$$\begin{aligned} D_{KL}(p||q) &= \int d^D x \int d^D y \langle O_i(x) O_j(y) \rangle_p \delta\lambda^i \delta\lambda^j \\ &= \text{Vol}(\mathcal{M}_D) \int d^D x \langle O_i(x) O_j(0) \rangle_p \delta\lambda^i \delta\lambda^j \end{aligned}$$

Two Divergences:

Info Density

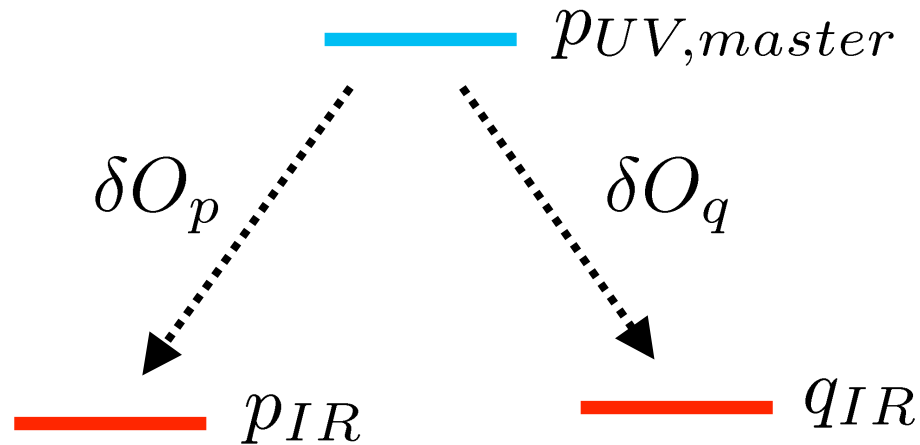


i) IR (just use $\mathcal{D}_{KL} \equiv D_{KL}/\text{Vol}(\mathcal{M}_D)$)

ii) UV from contact terms (i.e. $x \sim 0$ region)

Scheme Dependence

So, $D_{KL}(p||q)$ depends on a scheme
specify it once for master theory...

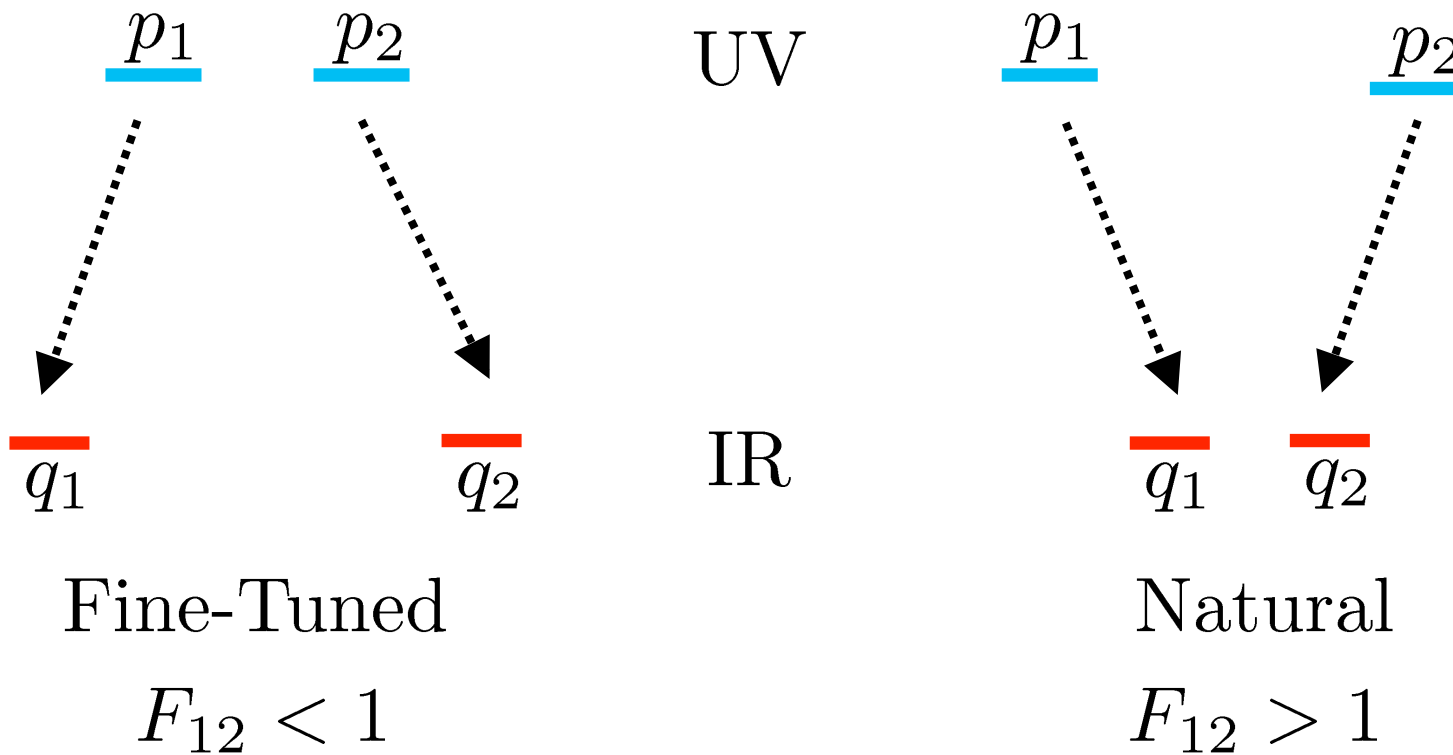


RG: $\frac{\partial D_{KL}(p||q)}{\partial(\log \mu)^m}$ tells us info as a f^n of scale

Remainder of Talk: Applications

Quantifying Fine-Tuning

$$F_{ij} \equiv \frac{D_{KL}(p_i^{UV} || p_j^{UV})}{D_{KL}(q_i^{IR} || q_j^{IR})}$$



Proximity for CFTs

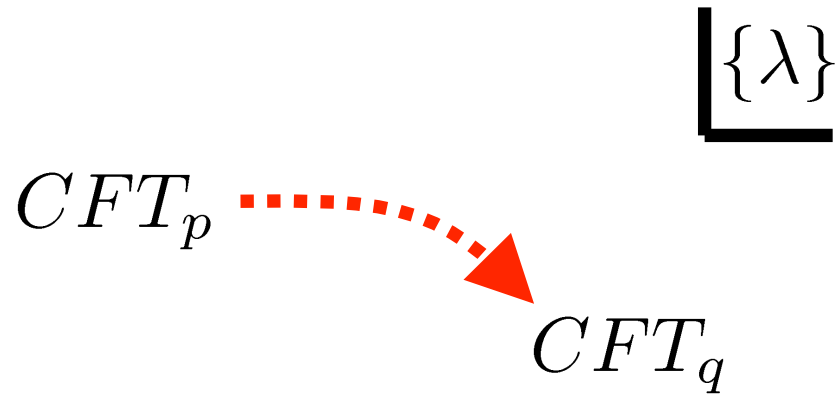
Special Case: p a CFT

$$D_{KL}(p||q) = \int d^D x \int d^D y \langle O_i(x) O_j(y) \rangle_p \delta \lambda^i \delta \lambda^j$$

$$\langle O_i(x) O_j(y) \rangle_p = \frac{G_{ij}^{Zamolodchikov}}{||x-y||^{2\Delta_O}}$$

$$\Rightarrow G_{ij}^{Information} \propto G_{ij}^{Zamolodchikov}$$

ii Zamolodchikov Metric!!



Two CFTs connected by marg. pertⁿs

Formal notion of proximity between CFTs...

?? Zamolodchikov Metric??

So why didn't Douglas like this?

Not all CFTs connected by marginal pertⁿs...

Example: Two isolated $c = 4/5$ Models:

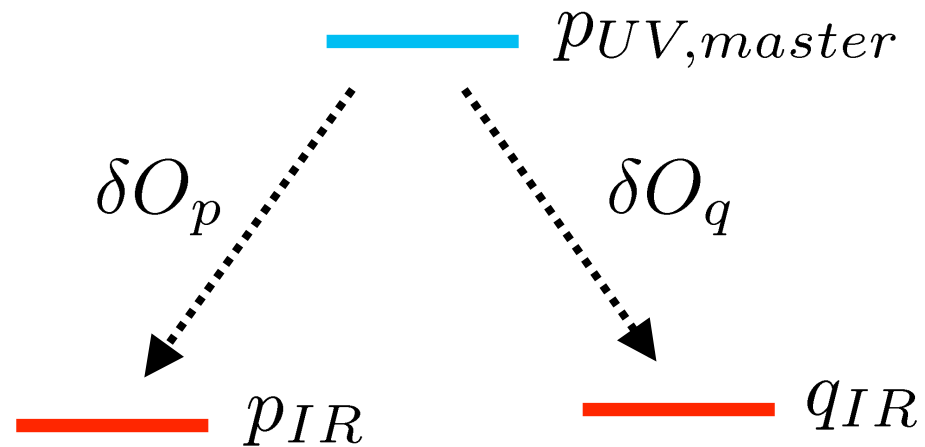
Douglas, "Spaces of Quantum Field Theories" 1005.2779

i) Tetracritical Ising

ii) 3-State Potts

But Recall...

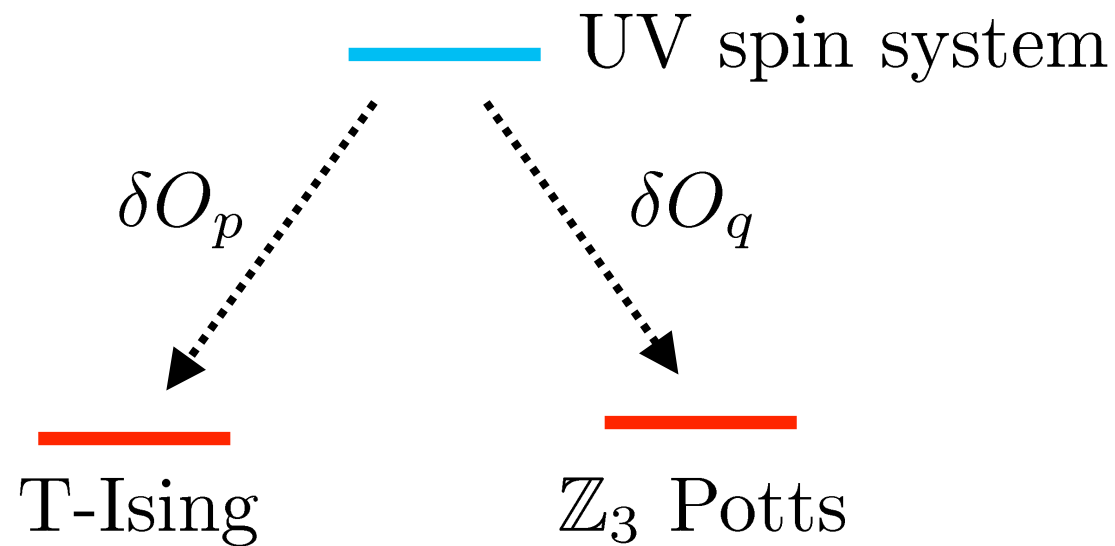
p and q could even have different field content!



\Rightarrow Just need a “master theory” (e.g. strings)

But Recall...

p and q could even have different field content!



\Rightarrow Just need a “master theory” (spin system)

Special Case: p a CFT

$$D_{KL}(p||q) = \int d^D x \int d^D y \langle O_i(x) O_j(y) \rangle_p \delta\lambda^i \delta\lambda^j$$

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What About Central Charge?

Treat $g_{\mu\nu}$ as a parameter: $p[\phi|g_{\mu\nu}]$

Take $q[\phi|g_{\mu\nu}] = p[\phi|g_{\mu\nu} + \delta g_{\mu\nu}]$

$$D_{KL}(p||q) = \int_x \int_y \langle T^{\mu\nu}(x) T^{\rho\sigma}(y) \rangle_p \delta g_{\mu\nu} \delta g_{\rho\sigma}$$

$$\text{Note } \langle T^{\mu\nu}(x) T^{\rho\sigma}(y) \rangle = C_T \frac{I^{\mu\nu, \rho\sigma}(x, y)}{||x-y||^{2D}}$$

Info $\propto C_T$: measures local degrees of freedom

Proximity for Flux Vacua

A Toy Model

Consider a theory of one real scalar ϕ

$$S[\phi] = \int d^4x \left(\frac{1}{2} (\partial\phi)^2 - V(\phi) \right)$$

Suppose we have fluxes \vec{N} and \vec{M}
with eff. potentials $V_{\vec{N}}(\phi)$ and $V_{\vec{M}}(\phi)$

Suppose further:

$$V(\phi) = \frac{1}{\Lambda^{2l-4}} (\phi - \phi_1)^2 \cdots (\phi - \phi_l)^2$$

ϕ_i depend on choice of flux vector

Question

¿What is proximity $D_{KL}(\vec{N} || \vec{M})$?

Approximation: Assume a “discretuum”
such that ϕ_i in $V(\phi)$ are parameters:

$$V(\phi | \{\phi_1, \dots, \phi_l\}) = \frac{1}{\Lambda^{2l-4}} (\phi - \phi_1)^2 \cdots (\phi - \phi_l)^2$$

a la Bousso Polchinski hep-th/0004134

So, $p[\phi | \{\phi_1, \dots, \phi_l\}]$

Computation

$$D_{KL}(\vec{N} || \vec{M}) = G_{ij} \delta\phi^i \delta\phi^j$$

$$G_{ij} = \int \int d^4x d^4y \left\langle \frac{\partial V(\phi(x))}{\partial \phi^i} \frac{\partial V(\phi(y))}{\partial \phi^j} \right\rangle$$

$$\text{Saddle Point: } p[\phi] \sim \frac{1}{l} \left(\frac{e^{-S_1}}{Z_1} + \dots + \frac{e^{-S_l}}{Z_l} \right)$$

$$S_k[\phi] = \int d^4x \left(\frac{1}{2} (\partial\phi)^2 - \frac{1}{2} m_k^2 \phi^2 \right)$$

Answer

$$D_{KL}(\vec{N} || \vec{M}) = G_{ij} \delta\phi^i \delta\phi^j$$

$$G_{ij} = \delta_{ij} \times \frac{\text{Vol}(\mathcal{M}_4)}{l} \times m_i^2$$

$$m_i^2 = V''(\phi)|_i = \frac{2}{\Lambda^{2l-4}} \prod_{k \neq i} (\phi_k - \phi_i)^2$$

Proximity for Inflatons

Suppose...

Suppose we have single scalar slow-roll with:
Large field range, and effective action:

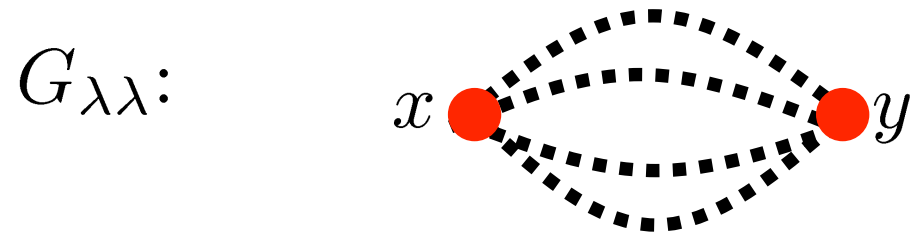
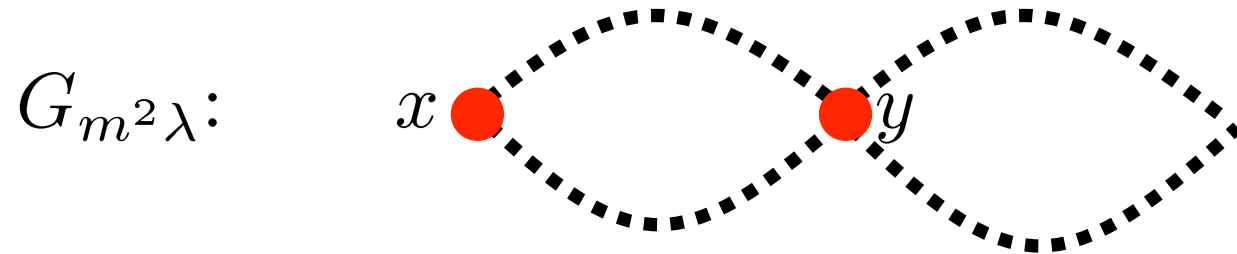
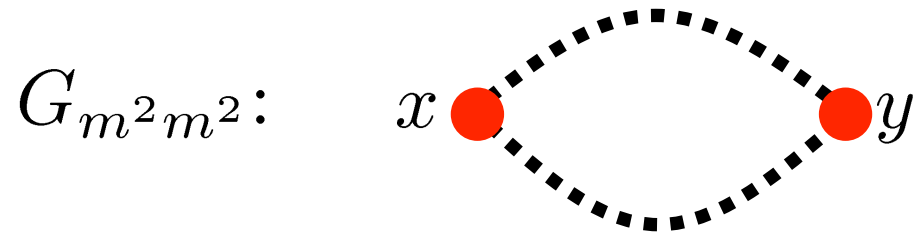
$$S_{eff}[\phi] = \int d^4x \left(\frac{1}{2} (\partial\phi)^2 - \frac{m^2}{2} \phi^2 - \frac{\lambda}{4!} \phi^4 \right)$$

Proximity of $\lambda = 0$ and $\lambda \neq 0$?

How much info do higher order ops give?

Information Metric

We have two parameters: m^2 and λ ...



Information Metric

We have two parameters: m^2 and λ ...

$$G_{m^2 m^2} \sim \text{Vol}(\mathcal{M}_4) \times \log m^2$$

$$G_{m^2 \lambda} \sim \text{Vol}(\mathcal{M}_4) \times (\Lambda_{UV}^2 + m^2 \log m^2)$$

$$G_{\lambda \lambda} \sim \text{Vol}(\mathcal{M}_4) \times (\Lambda_{UV}^4 + \log^3 m)$$

$$\lambda = 0 \text{ vs } \lambda \neq 0$$

$$D_{KL}(\lambda = 0 || \lambda \neq 0) \sim \lambda^2 \times \left(\frac{\Lambda_{UV}}{\Lambda_{IR}} \right)^4 \sim \lambda^2 \times \left(\frac{M_{pl}}{m_{inf}} \right)^4$$

But to not spoil $m^2\phi^2$ slow roll, we already need:

$$\lambda < \left(\frac{m_{inf}}{\Delta\phi} \right)^2 \quad \text{so if } \Delta\phi \sim 10M_{pl} \dots$$

$$\Rightarrow D_{KL} < \left(\frac{M_{pl}}{\Delta\phi} \right)^4 \sim 10^{-4}$$

Conclusions / Future

Conclusions

- To study landscape, would like $D(EFT_i, EFT_j)$
- Proposal from statistical inference: $D_{KL}(p||q)$
- Recovers G^{Zam} , and other intuitive measures
- Applications: Flux Vacua, Fine-Tuning, Inflatons,...

Future

- Formal: Use this to prove a / c / F-theorems?
- Pheno: SM versus MSSM w/ heavy superpartners?
- Cosmo: Apply to inflationary measure problem?