

e.g.2

观测值 λ_R = λ_R + $\delta\lambda$ (一阶抵消项) + λ_R (二阶正带项) = 0

抵消项也抵消高阶项

$$\lambda(\Lambda) = \lambda_R(\mu) + \delta\lambda(\Lambda)$$

$$m^2(\Lambda) = m_R^2(\mu) + \delta m^2(\Lambda)$$

$$m_R^2(\mu) = m^2(\Lambda) - \delta m^2(\Lambda)$$

有限 = $\infty - \infty$
(观测值)

Wilson Renormalized Group Theory

- ① Λ 大 \rightarrow 小
- ② 处理与相变有关的问题
- ③ 不要抵消项

3.21

重整化: 目的: 发散抵消发现
 区分: m_R, m_0
 系统的一套方法

$$Z = \text{Tr}(e^{-\beta H})$$

$$= \sum_{\{S_i\}} e^{-K \sum_i S_i^2} \quad S_i = \pm 1$$



FUTURE PEOPLE

$$= \sum_{\text{odd}} \sum_{\text{even}} e^{k \sum_i s_i s_{i+1}}$$

$$= \sum_{\text{odd}} e^{\sum_i k' s_{2i} s_{2i+1}}$$

$$\sum_{s_2=\pm 1} e^{k s_1 s_2 + k_2 s_2 s_3} = e^{k(s_1+s_3)} + e^{-k(s_1+s_3)} \cong A e^{k' s_1 s_3}$$

even 的求和掉 $\cong A e^{k' s_1 s_3}$

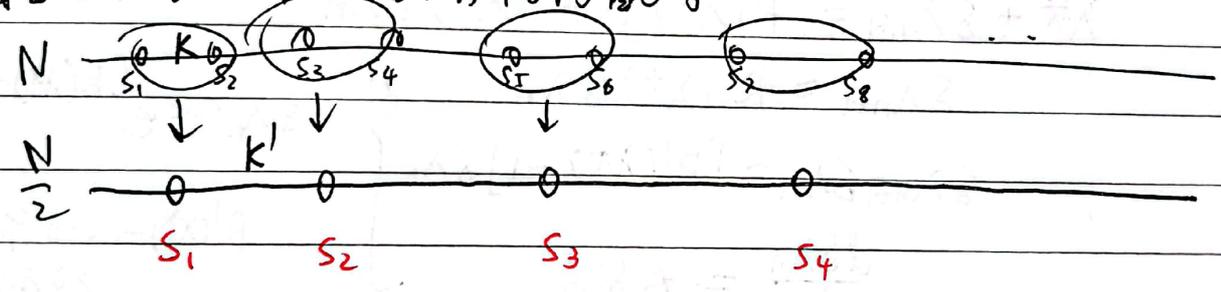
$$\begin{aligned} (s_1, s_3) = & \begin{cases} (1, 1) \\ (-1, -1) \end{cases} \} e^{2k} + e^{-2k} \triangleq A e^{k'} \\ & \begin{cases} (1, -1) \\ (-1, 1) \end{cases} \} 2 \triangleq A e^{-k'} \end{aligned}$$

即有 $A e^{k'} = 2 \cosh(k) \quad A e^{-k'} = 2$

$$\Leftrightarrow e^{2k'} = \cosh(2k) \quad \text{or} \quad 2k' = \ln \cosh(2k)$$

$$\Leftrightarrow k' = \frac{1}{2} \ln \cosh(2k)$$

相当于把链上节点(各 s_i) 两两配对



重新命名, 对无穷大 N 来说, 事实上没有任何变化, 唯一的变化是耦合常数 $k \rightarrow k'$. 如你所无穷做下去

总结: ① 两个 Block 看作一个整体

② re-scaling: $2a \rightarrow a$

$$\begin{cases} k' = \frac{1}{2} \ln \cosh(2k) \\ k_0 = \beta J = \frac{J}{k_B T} \end{cases} \quad \text{函数} \downarrow \quad \text{变量 } \lambda_{n+1} = R(\lambda_n)$$

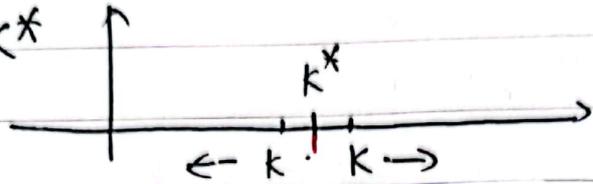
极限: $k^* = \frac{1}{2} \ln \cosh(2k^*) \Rightarrow k^* = 0$

且不论 k_0 是多少, 最后都会收敛到 $k_n \rightarrow 0$



$$2k \leq \alpha \ln \cosh(2k)$$

若 $k_0 < k^*$



k 会演化到 0

对应 $T \rightarrow \infty$: 无序(完全)

$k_0 > k^*$

$k \rightarrow \infty$

对应 $T \rightarrow 0$

$$k^* - 0^+ \Rightarrow T = \infty$$

$$k^* + 0^+ \Rightarrow T = 0$$

这是一个典型的相变行为

1dim: $\lambda_{n+1} = R(\lambda_n)$ 令 $\lambda^* = R(\lambda^*)$

$$\lambda_n = \lambda^* + \delta\lambda_n$$

$$\lambda_n = \lambda^* + \delta\lambda_n$$

$$\lambda^* + \delta\lambda_{n+1} = R(\lambda^*) + R'(\lambda^*) \delta\lambda_n$$

$$\delta\lambda_{n+1} = R'(\lambda^*) \delta\lambda_n$$

$$\delta\lambda_{n+1} \delta\lambda_n = [R'(\lambda^*) - 1] \delta\lambda_n$$

$$\frac{d\delta\lambda_n}{dn} = A \delta\lambda_n$$

$$\delta\lambda_n \propto e^{[R'(\lambda^*) - 1]n}$$

$R'(\lambda^*) - 1 < 0$ 稳
 > 0 不稳

2dim: 略

以正合并过程, 转到 k -space 去, 实际上是能标 Λ 下降的过程 (Fourier trans: 高频项(高阶项)代表更细的细节, 舍去高频项, 即 $\Lambda \downarrow$)

标度分析

① ϕ^4 理论

$$\mathcal{L} = \frac{1}{2}(\partial\phi)^2 - \frac{m^2}{2}\phi^2 - \frac{\lambda}{4!}\phi^4$$



eg. $\int_{-\infty}^{+\infty} f(x) dx = \int_{-\infty}^{+\infty} f(\lambda x) d(\lambda x)$
 FUTURE PEOPLE $= \int_{-\infty}^{+\infty} f(y) dy \quad y = \lambda x$

$$S = \int \mathcal{L} dx$$

$$\int dx \frac{1}{2} (\partial_t \phi)^2 = \int dx' \frac{1}{2} (\partial_{t'} \phi(x'))^2$$

$t' = \lambda t$
 $x' = \lambda x$

$$\phi(\lambda x) = \lambda^\alpha \phi(x)$$

$$= \lambda^{d-2+2\alpha} \int dx \frac{1}{2} (\partial_t \phi(x))^2$$

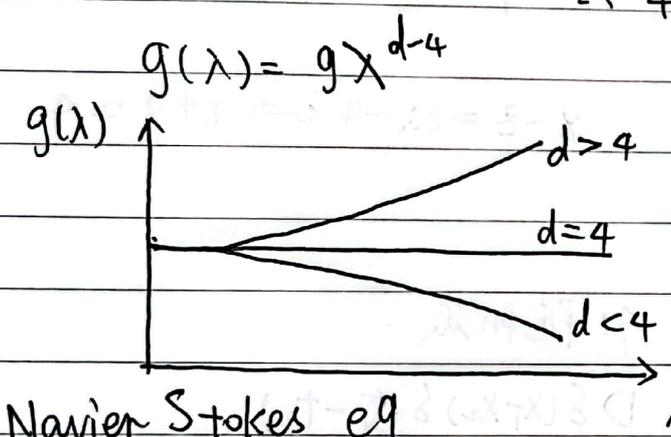
对 λ 成立 $\therefore d-2+2\alpha=0 \Rightarrow \alpha = \frac{2-d}{2}$

$$\int g \phi^4(x) dx = \int g' \phi^4(x') dx' = g \lambda^{4\alpha+d+\beta} \int \phi^4(x) dx$$

$g' = g \lambda^\beta$

$$\therefore 4\alpha + d + \beta = 0$$

$$\therefore 4 - 2d + d + \beta = 0 \Rightarrow \beta = d - 4$$



Navier Stokes eq

$$\begin{cases} u_t + u \cdot \nabla u + \nabla p - \Delta u = 0 & x \rightarrow \lambda x \\ \nabla \cdot u = 0 & t \rightarrow \lambda^2 t \\ u|_{t=0} = u_0 \end{cases}$$

$$\begin{cases} u(\lambda x, \lambda^2 t) \rightarrow \lambda^{\alpha_1} u(x, t) \\ p(\lambda x, \lambda^2 t) \rightarrow \lambda^{\alpha_2} p(x, t) \\ \alpha_1 = -1 \quad \alpha_2 = -2 \end{cases}$$

此变换下该方程不变

$$u(x, t) \Rightarrow u'(x', t') = u'(\lambda x, \lambda^2 t)$$

$t' = \lambda^2 t, x' = \lambda x$
 $u' = \lambda u$

这个例子说明很多微分方程都可以做标度分析
是在

$$\text{Kpz eq} \quad \frac{\partial h}{\partial t} = \nu \frac{\partial^2 h}{\partial x^2} + \eta$$

$$\left\{ \begin{array}{l} \frac{\partial h'}{\partial t'}(x', t') = \nu \frac{\partial^2 h'(x', t')}{\partial x'^2} + \eta'(x', t') \\ x' = bx \\ t' = b^z t \end{array} \right.$$

$$h'(x', t') = h'(bx, b^z t) = b^x h(x, t)$$

$$\frac{\partial h'}{\partial t'} = \frac{b^x}{b^z} \frac{\partial h}{\partial t}$$

$$\frac{\partial^2 h'}{\partial x'^2} = \frac{b^x}{b^z} \frac{\partial^2 h}{\partial x^2}$$

$$\left. \begin{array}{l} \frac{\partial h'}{\partial t'} = \frac{b^x}{b^z} \frac{\partial h}{\partial t} \\ \frac{\partial^2 h'}{\partial x'^2} = \frac{b^x}{b^z} \frac{\partial^2 h}{\partial x^2} \end{array} \right\} b^{x-z} \frac{\partial h}{\partial t} = \nu b^{x-2} \frac{\partial^2 h}{\partial x^2}$$

欲让此模型不变 只需令 $z=2$

$$b^{x-2} \frac{\partial h}{\partial t} = b^{2x-4} \left(\frac{\partial^2 h}{\partial x^2} \right)^2$$

$$x-2 = 2x-4 \Leftrightarrow x+z=4$$

其它例子如 δ 函数

$$\langle \eta(x, t) \rangle = 0$$

η : 随机数

$$\langle \eta(x_1, t_1) \eta(x_2, t_2) \rangle = D \delta(x_1 - x_2) \delta(t_1 - t_2)$$

$$\delta(bx) = \frac{1}{b} \delta(x)$$

下节预告: 转到动量空间: Wilson RG Theory

