A Companion Paper to the Momentum Space RG Article by *Knops et al.*

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1 Length Rescaling

To begin momentum space renormalization one must rescale the Hamiltonian via a length rescale in real space (and equivalently rescale the system in wave number space):

$$r' = \frac{r}{1+\varepsilon}$$

= $k(1-\varepsilon+\varepsilon^2-...)$
 $\approx r(1-\varepsilon)$ (1)

$$k' = k(1+\varepsilon) \tag{2}$$

Accordingly, our Gaussian field also rescales:

$$X(\mathbf{k'}) = \frac{1}{(1+\varepsilon)^2} X(\mathbf{k})$$

= $X(\mathbf{k})(1-2\varepsilon+3\varepsilon^2-4\varepsilon^3+...)$
 $\approx X(\mathbf{k})(1-2\varepsilon)$ (3)

2 Hamiltonian for Gaussian field X(k)

We are given a rescaled Hamiltonian:

$$H'(J, y') = H_0(X(\mathbf{k})) + H_1(Y(\mathbf{k})) + \frac{2y'}{a^2} \int d^2r \cos(X(\mathbf{r}) + Y(\mathbf{r}))$$
(4)

We want to use this model in a partition function, which in general is:

$$Z(J,y) = \int DX(\mathbf{r})e^{H(J,y)}$$
(5)

where

$$\int DX(\mathbf{r}) = \text{"sum over all states"}$$
(6)

Our partition function for our model corresponding to (4) is simply:

$$Z(J,y') = \int DX(\mathbf{r}) \int DY(\mathbf{r})e^{H'(J,y')}$$

=
$$\int DX(\mathbf{r}) \exp\left[H_0(X(\mathbf{k}))\right] \int DY(\mathbf{r}) \exp\left[H_1(Y(\mathbf{k})) + \frac{2y'}{a^2} \int d^2r \cos(X(\mathbf{r}) + Y(\mathbf{r}))\right]$$

=
$$\int DX(\mathbf{r})e^{H_0(X(\mathbf{k}))} \int DY(\mathbf{r})e^{H_1(Y(\mathbf{k}))} \exp\left[\frac{2y'}{a^2}I\right]$$
(7)

where I is the integral given in (4). We understand $\exp \{H_1(Y(\mathbf{k}))\}$ to be the probability density of $Y(\mathbf{k})$ states. Therefore we conclude that the following relationship holds:

$$\int DY(\mathbf{r})e^{H_1(Y(\mathbf{k}))} \exp\left[\frac{2y'}{a^2}I\right] = \left\langle \exp\left[\frac{2y'}{a^2}I\right] \right\rangle_1 \tag{8}$$

where $\langle \rangle_1$ denotes the average with respect to the states modeled by $H_1(Y(\mathbf{k}))$.

The average of the exponential function is known as the moment-generating function due to its use in deriving the moments of a distribution. It can also be used to find the cumulants of a distribution:

$$\langle e^{kx} \rangle = \sum_{n=0}^{\infty} k^n \frac{\langle x^n \rangle}{n!} = \exp\left[\sum_{n=1}^{\infty} k^n \frac{\langle \langle x^n \rangle \rangle}{n!}\right]$$
(9)

Note that $\langle x^n \rangle$ and $\langle \langle x^n \rangle \rangle$ correspond to the *n*th moment and *n*th cumulant of the distribution respectively. With this knowledge we will expand (8) in terms of the cumulants of $Y(\mathbf{r})$:

$$\left\langle \exp\left[\frac{2y'}{a^2}I\right] \right\rangle_1 = \exp\left[\sum_{n=1}^{\infty} \left(\frac{2y'}{a^2}\right)^n \frac{\langle\langle I^n \rangle\rangle_1}{n!}\right] \\ \approx \exp\left[\left(\frac{2y'}{a^2}\right) \langle\langle I \rangle\rangle_1 + \frac{1}{2} \left(\frac{2y'}{a^2}\right)^2 \langle\langle I^2 \rangle\rangle_1\right]$$
(10)

Recall:

$$I = \int d^2 r \cos(X(\boldsymbol{r}) + Y(\boldsymbol{r}))$$
(11)

We can redefine our original (total) Gaussian field to make our next few equations neater:

$$Z(\mathbf{r}) = X(\mathbf{r}) + Y(\mathbf{r}) \tag{12}$$

such that:

$$Z(\boldsymbol{r}) \sim \mathcal{N}(0, \sigma_X^2 + \sigma_Y^2) \tag{13}$$

Expanding the argument in (10):

$$\left(\frac{2y'}{a^2}\right)\langle\langle I\rangle\rangle_1 + \frac{1}{2}\left(\frac{2y'}{a^2}\right)^2\langle\langle I^2\rangle\rangle_1 = \left(\frac{2y'}{a^2}\right)\int d^2r\langle\cos Z(\boldsymbol{r})\rangle_1 + \frac{1}{2}\left(\frac{2y'}{a^2}\right)^2\int d^2r d^2r' \left[\langle\cos Z(\boldsymbol{r})\cos Z(\boldsymbol{r'})\rangle_1 - \langle\cos Z(\boldsymbol{r})\rangle_1\langle\cos Z(\boldsymbol{r'})\rangle_1\right]$$

$$(14)$$

We return to the partition function with (10) in mind:

$$Z = \int DX(\mathbf{r}) e^{H_0(X(\mathbf{k}))} \left\langle \exp\left[\frac{2y'}{a^2}I\right] \right\rangle_1$$

=
$$\int DX(\mathbf{r}) e^{H_0(X(\mathbf{k}))} \exp\left[\left(\frac{2y'}{a^2}\right) \langle \langle I \rangle \rangle_1 + \frac{1}{2} \left(\frac{2y'}{a^2}\right)^2 \langle \langle I^2 \rangle \rangle_1\right]$$

=
$$\int DX(\mathbf{r}) \exp\left[H_0(X(\mathbf{k})) + \left(\frac{2y'}{a^2}\right) \langle \langle I \rangle \rangle_1 + \frac{1}{2} \left(\frac{2y'}{a^2}\right)^2 \langle \langle I^2 \rangle \rangle_1\right] (15)$$

We retrieve our Hamiltonian from the argument in the partition function:

$$H = H_0(X(\boldsymbol{k})) + \left(\frac{2y'}{a^2}\right) \int d^2 r \langle \cos Z(\boldsymbol{r}) \rangle_1 + \frac{1}{2} \left(\frac{2y'}{a^2}\right)^2 \int d^2 r d^2 r' \left[\langle \cos Z(\boldsymbol{r}) \cos Z(\boldsymbol{r'}) \rangle_1 - \langle \cos Z(\boldsymbol{r}) \rangle_1 \langle \cos Z(\boldsymbol{r'}) \rangle_1 \right]$$
(16)

We will attempt to evaluate the first-order term in (16):

$$\langle \cos Z(\boldsymbol{r}) \rangle_1 = \langle \cos(X(\boldsymbol{r}) + Y(\boldsymbol{r})) \rangle_1$$

$$= \langle \cos X(\boldsymbol{r}) \cos Y(\boldsymbol{r}) + \sin X(\boldsymbol{r}) \sin Y(\boldsymbol{r}) \rangle_1$$

$$= \langle \cos X(\boldsymbol{r}) \cos Y(\boldsymbol{r}) \rangle_1 + \langle \sin X(\boldsymbol{r}) \sin Y(\boldsymbol{r}) \rangle_1$$

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$$(17)$$

Keep in mind the following for $X \sim \mathcal{N}(0, \sigma^2)$:

$$\langle x^n \rangle = \begin{cases} 0 & \text{if } n \text{ is odd} \\ \sigma^n (n-1)!! & \text{if } n \text{ is even} \end{cases}$$
(18)

We investigate the terms in (17):

$$\begin{aligned} \langle \cos Y \rangle_{1} &= \left\langle 1 - \frac{Y^{2}}{2!} + \frac{Y^{4}}{4!} - \dots \right\rangle_{1} \\ &= \left\langle 1 \rangle_{1} - \frac{1}{2} \langle Y^{2} \rangle_{1} + \frac{1}{24} \langle Y^{4} \rangle_{1} - \dots \right. \\ &= \left. 1 - \frac{1}{2} \langle Y^{2} \rangle_{1} + \frac{3}{24} \langle Y^{2} \rangle_{1}^{2} - \dots \right. \\ &= \left. \sum_{n=0}^{\infty} \frac{1}{n!} \left[-\frac{1}{2} \langle Y^{2} \rangle_{1} \right]^{n} \\ &= \left. \exp\left(-\frac{1}{2} \langle Y^{2} \rangle_{1} \right) \right. \end{aligned}$$
(19)

The 3rd equality holds because $Y(\mathbf{r})$ is a Gaussian with zero mean. Similarly:

$$\langle \sin Y \rangle_1 = \left\langle Y - \frac{Y^3}{3!} + \frac{Y^5}{5!} - \dots \right\rangle_1$$

$$= \langle Y \rangle_1 - \frac{1}{6} \langle Y^3 \rangle_1 + \frac{1}{120} \langle Y^5 \rangle_1 - \dots$$

$$= 0$$

$$(20)$$

Remember that odd-numbered moments vanish when the mean is zero. Now we seek the find the average of Y^2 as in (19). We know this to be equivalent to the variance of Y if $\langle Y \rangle_1$ is 0. Note Eq. 2.4 from the paper:

$$H_1(Y(\mathbf{k})) = -\frac{1}{2} \int \frac{d^2k}{(2\pi)^2} \left[\frac{Jk^2}{\psi(k)} Y(\mathbf{k}) Y(-\mathbf{k}) + E_1(k) \right]$$
(21)

We can retrieve the variance from this expression because the distribution of a Gaussian variable is of the form:

$$\rho_Y(y) \sim \exp\left[-\frac{(y-\mu)^2}{2\sigma^2}\right]$$
(22)

Therefore:

$$\langle Y^2 \rangle_1 = \int DY(\mathbf{r}) Y^2 \exp[H_1]$$

$$= \int \frac{d^2k}{(2\pi)^2} \frac{\psi(k)}{Jk^2}$$

$$= \frac{1}{(2\pi)^2 J} \int_0^{2\pi} d\phi \int_0^{\infty} dk \frac{\psi(k)}{k}$$

$$= -\frac{\varepsilon}{2\pi J} \int_0^{\infty} \frac{d\phi(k)}{dk}$$

$$= -\frac{\varepsilon}{2\pi J} \left[\phi(\infty) - \phi(0)\right]$$

$$= \frac{\varepsilon}{2\pi J}$$

$$(23)$$

where we have used Eq. 2.5 from the article to relate the cut-off functions and we know the cut-off functions must go to zero at infinity and to unity at zero.

Our result for the first order integrand is then:

$$\langle \cos(X(\boldsymbol{r}) + Y(\boldsymbol{r})) \rangle_1 = \cos X(\boldsymbol{r}) \exp\left(-\frac{1}{2} \langle Y^2 \rangle_1\right)$$

= $\cos X(\boldsymbol{r}) \exp\left(-\frac{\varepsilon}{4\pi J}\right)$
 $\approx \cos X(\boldsymbol{r}) \left[1 - \frac{\varepsilon}{4\pi J}\right]$ (24)