

# How to get Stable distributions using RG?

①

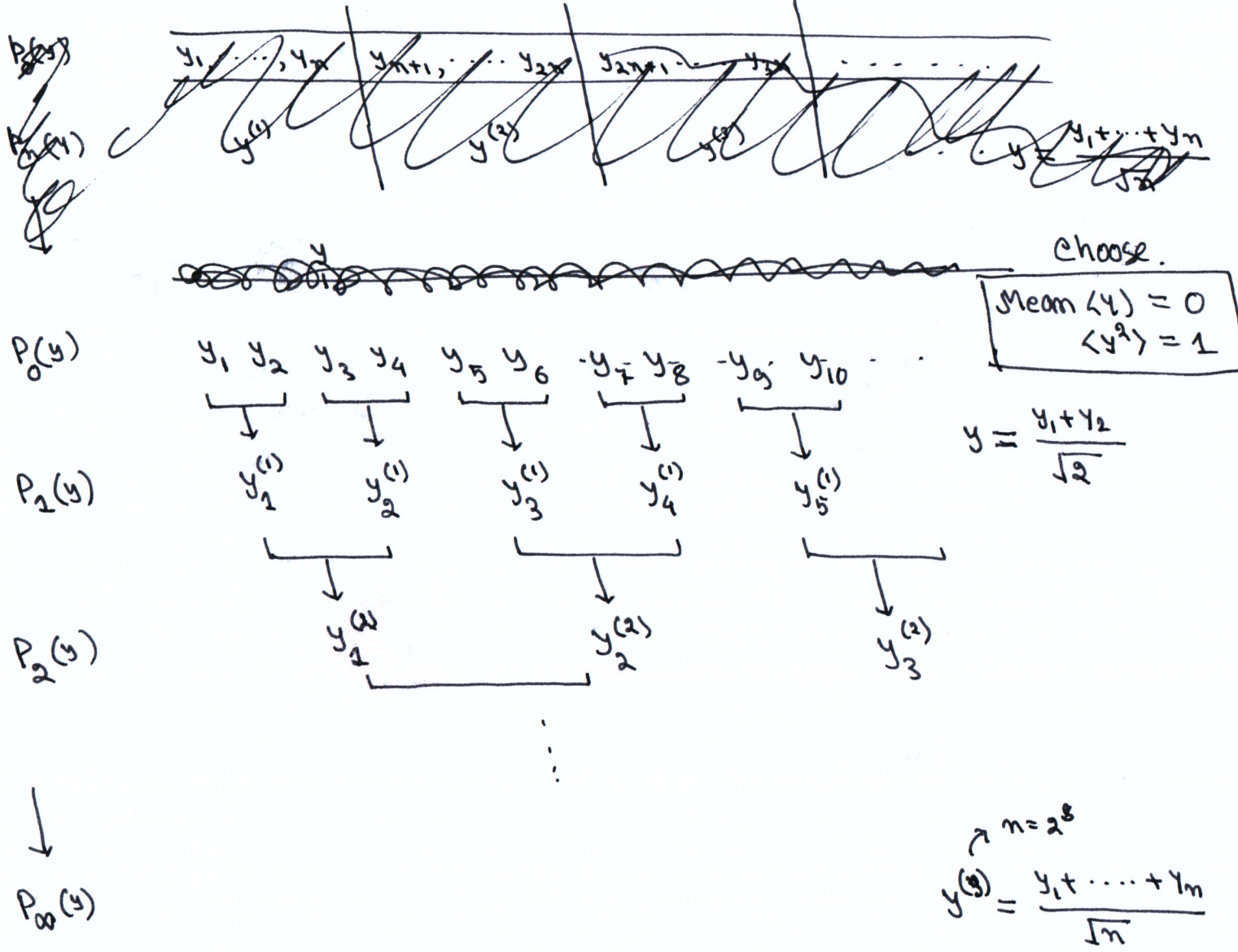
Ref 1. Exercise 12.11 in Book of Sethna.

↳ [functional RG]

Term paper

- Ref 2. "RG and probability theory", Jona-Lasinio, Phys. Rep. 352(2001) 439.
- Ref 3. "An elementary renormalization-group approach to ...". Ariel Amir, J Stat mech (2020) 013214.

Basic idea: For CLT case,



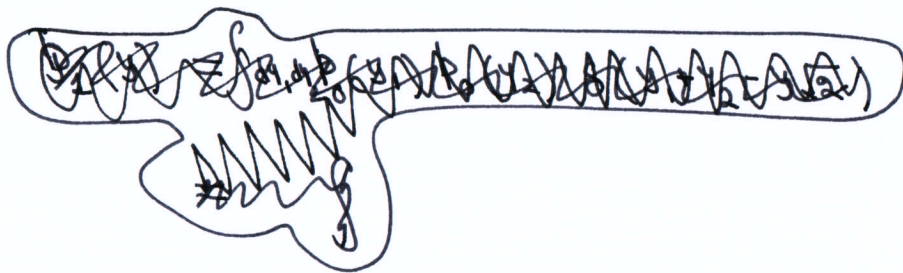
Show that  
 (For CLT-case)  $P_\infty(y) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{y^2}{2\sigma^2}}$  (take  $\sigma^2 = 1$  for simplicity)

RG operation:

$$y = \frac{y_1 + y_2}{\sqrt{2}}$$

coarse-graining

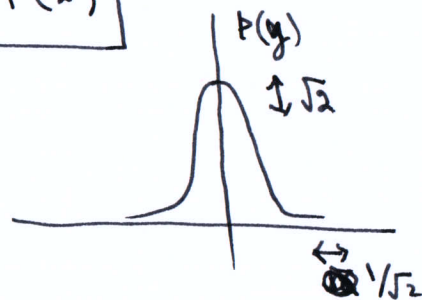
(2)



$$\begin{aligned}
 P_2(y) &= \int dy_1 dy_2 P_0(y_1) P_0(y_2) \delta\left(\frac{y_1 + y_2}{\sqrt{2}} - y\right) \\
 &= \sqrt{2} \int dy_1 dy_2 P_0(y_1) P_0(y_2) \delta(y_1 + y_2 - \sqrt{2}y) \\
 &= \sqrt{2} \int dy_2 P_0(\sqrt{2}y - y_2) P_0(y_2)
 \end{aligned}$$

$$\Rightarrow \boxed{R[P](y) = \sqrt{2} \int dx P(\sqrt{2}y - x) P(x)}$$

Rescaling



A fixed point:

$$R[P_n^*](y) = P_{n+1}^*(y) = P_n^*$$

$$\boxed{P^*(y) = \sqrt{2} \int dx P^*(\sqrt{2}y - x) P^*(x)}$$

The RG equation.

Easier to solve in Fourier-space.

$$\hat{g}(k) = \int dy e^{iky} P(y)$$

$$\Rightarrow \boxed{\hat{g}(k) = \left[ \hat{g}\left(\frac{k}{\sqrt{2}}\right) \right]^2}$$

$$\Rightarrow \hat{g}(k) = e^{-\frac{k^2}{2}} \text{ is a fixed point}$$

$\uparrow$   
 $P^*(x) = \text{Gaussian.}$

⊗ generalize:

$$y = \frac{y_1 + y_2}{b}$$

(3)

$$P^*(y) = b \int dx P^*(x) P^*(by-x)$$

b is the rescaling factor.

General:  $b = 2^{1/\alpha}$

$$\hat{g}(k) = \left[ \hat{g}\left(\frac{k}{2^{1/\alpha}}\right) \right]^2$$

$P\left(\frac{M_n}{n^{1/\alpha}} = y\right) = \text{Lévy distr.}$

$$\hat{g}(k) = e^{-|k|^\alpha}$$

Symmetric.

is a fixed point.  $\Leftrightarrow$  Lévy distr.

[ choosing b is equivalent to ~~also~~ keeping conserved quantities in  $\mathbb{R}^d$  ]

[ Information of b comes from your parent  $P_0(x)$ , knowing what is your typical fluctuations ]

If b is chosen incorrectly, the limiting distribution may be trivial ( $\delta(x)$ ) or it may not exist!

Then, choice of b comes from demanding that there is a unique limit  $P^*(y)$ .

Is it an attractive fixed point? [Relevant, marginal, irrelevant perturbations.]

Linear stability analysis:



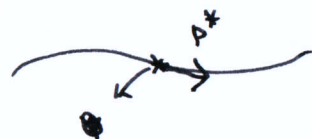
make a small perturbation around  $P^*(y)$

$$\begin{aligned} R[P^* + \epsilon \delta P](y) &= b \int dx [P^* + \epsilon \delta P](by-x) (P^*(x) + \epsilon \delta P(x)) \\ &= P^*(y) + \epsilon [L \cdot \delta P](y) \end{aligned}$$

$$[L \cdot \delta P](y) = b \int dx [P^*(by-x) \cdot \delta P(x) + \delta P(by-x) \cdot P^*(x)]$$

Eigenbasis:

$$[L \cdot \psi_\lambda](y) = \lambda \psi_\lambda(y)$$



Take perturbation along <sup>any</sup> one eigen basis.  $\delta p = \psi_\lambda$

(4)

$$R[p^* + \epsilon \psi_\lambda] = p^* + \epsilon \lambda \psi_\lambda \Rightarrow \text{we use: } \begin{matrix} \text{Repeated} \\ \psi_\lambda \end{matrix} \rightarrow p^* + \epsilon \lambda^n \psi_\lambda$$

$$\hat{R}[g^*(k) + \epsilon \hat{\psi}_\lambda(k)] = g^*(k) + \epsilon \cdot \lambda \cdot \hat{\psi}_\lambda(k)$$

⊗

⊗ What are eigenvalues and eigenvectors. Fourier basis.

$$\begin{aligned} [\alpha \cdot \psi_\lambda](y) &= \lambda \psi_\lambda(y) \\ \Rightarrow [\hat{\alpha} \cdot \hat{\psi}_\lambda](k) &= \lambda \hat{\psi}_\lambda(k) \end{aligned}$$

$$2 \cdot g^*\left(\frac{k}{b}\right) \hat{\psi}_\lambda\left(\frac{k}{b}\right) = \lambda \hat{\psi}_\lambda(k)$$

**Solution:**

$$\hat{\psi}_\lambda(k) = (ik)^{\alpha_1} g^*(k) \quad \text{with } \lambda = \frac{2}{b^{\alpha_1}}$$

What it means:

$$\lambda_0 = 2 > 1 \quad \longleftrightarrow \text{relevant}$$

← but does not conserve probability.

$$\lambda_1 = \frac{2}{b} > 1 \quad \longleftrightarrow \quad \text{"}$$

for  $b = \sqrt{2}$

$$\lambda_2 = \frac{2}{b^2} = 1 \quad \longleftrightarrow \text{marginal}$$

$$\lambda_3 = \frac{2}{b^3} < 1 \quad \longleftrightarrow \text{irrelevant.}$$

⋮

→ we don't perturb along this direction, because it breaks conservation that

$\langle y \rangle, \langle y^2 \rangle$  are fixed!