

Thm 2.8 (Bakry - Emery) Let  $\nu$ :

$$\mathbb{E}_\nu [F(\varphi)] = \frac{1}{Z} \int e^{-H(\varphi)} F(\varphi) d\varphi$$

be the prob. measure on  $X = \mathbb{R}^N$ , where  $\text{Hess } H(\varphi) \geq \lambda \cdot I \quad \forall \varphi \in X$ . Then  $\nu$  satisfies the LSI( $\lambda$ ) with  $\lambda \geq 1$ .

Proof.  $\nu$  can be characterized as the invariant measure for a Markov process with the generator

$$\Delta^H = \Delta - (\nabla H, \nabla), \quad \text{where}$$

$$\Delta = \sum_{x \in \Lambda} \frac{\partial^2}{\partial \varphi_x^2}, \quad (\nabla H, \nabla) = \sum_{x \in \Lambda} \frac{\partial H}{\partial \varphi_x} \frac{\partial}{\partial \varphi_x}$$

$|\Lambda| = N$

The corresponding semigroup  $P_t$  is defined by

$$P_t F(\varphi) := \mathbb{E}_{\varphi_0 = \varphi} [F(\varphi_t)]$$

If  $F_t := P_t F$ , then  $\frac{\partial F_t}{\partial t} = \Delta^H F_t$   
(backward Kolmogorov)

If  $m_t$  is the distribution at time  $t$ ,  $\mathbb{E}_{m_t} [F_0] = \mathbb{E}_{m_0} [F_t]$   $F_0 \equiv F$   
 $m_t \xrightarrow{t \rightarrow \infty} \nu$ .

Then  $\text{Ent}_\nu(F) := \mathbb{E}_\nu [\phi(F)] - \phi(\mathbb{E}_\nu [F])$

$\phi(x) = x \log x$ .

Vis inv.

Ergodicity

(Some analysis needed, as  $\phi$  isn't bounded on  $(0, \infty)$ )

$$= \mathbb{E}_\nu [\phi(F_0)] - \phi(\mathbb{E}_\nu [F_0])$$

$$= \mathbb{E}_\nu [\phi(F_0)] - \mathbb{E}_\nu [\phi(F_\infty)]$$

Ergodicity:  $\forall F_0 \in L^2(\nu)$ ,  $F_t \rightarrow \mathbb{E}_\nu [F_0]$

in  $L^2(\nu) \Rightarrow$  For any bounded smooth  $F_0: X \rightarrow \mathbb{R}$ ,  $g: \mathbb{R} \rightarrow \mathbb{R}$ ,  $\lim_{t \rightarrow \infty} \mathbb{E}_\nu [g(F_t)] = g(\mathbb{E}_\nu [F_0])$

Therefore,

$$\text{Ent}_V(F) = - \int_0^\infty dt \frac{\partial}{\partial t} \mathbb{E}_V[\phi(F_t)]$$

Prop. 2.2

$$\int_0^\infty dt I_V(F_t)$$

where  $I_V(F_t) := \mathbb{E}_V \left[ \frac{(\nabla F_t)^2}{F_t} \right]$

is the Fisher information.

Claim.  $I_V(F_t) \leq e^{-2\lambda t} I_V(F_0)$

The claim implies that

$$\begin{aligned} \text{Ent}_V(F) &\leq \int_0^\infty dt e^{-2\lambda t} I_V(F_0) = \frac{1}{2\lambda} I_V(F_0) \\ &= \frac{2}{\lambda} D_V(\sqrt{F}) \Rightarrow \text{log-Sobolev} \\ &\text{with } \mu = \lambda. \end{aligned}$$

Proof of the claim: use Gronwall.

$$\frac{\partial}{\partial t} I_V(F_t) = \frac{\partial}{\partial t} \mathbb{E}_V \left[ \frac{(\nabla F_t)^2}{F_t} \right] \stackrel{\text{and.}}{=} \mathbb{E}_V \left[ \frac{\partial}{\partial t} \frac{(\nabla F_t)^2}{F_t} \right]$$

$\nabla$  is invariant,

$\nabla \cdot \nabla = \Delta^H$

so  $\mathbb{E}_V[\Delta^H G] = 0$ ,

provided  $G$  is "nice"

$$\mathbb{E}_V \left[ \left( \frac{\partial}{\partial t} - \Delta^H \right) \frac{(\nabla F_t)^2}{F_t} \right]$$

Notice:  $\frac{\partial}{\partial t} F_t = \Delta^H F_t$

$$\# \frac{\partial}{\partial t} \frac{(\nabla F)^2}{F} = - \frac{1}{F^2} (\nabla F)^2 \Delta^H F + 2 \frac{(\nabla F, \nabla \Delta^H F)}{F}$$

$$= - \frac{1}{F^2} (\nabla F)^2 (\Delta F - (\nabla H, \nabla F)) + 2 \frac{(\nabla F, \nabla \Delta F)}{F} - 2 \frac{(\nabla F, \nabla (\nabla H, \nabla F))}{F}$$

$$\textcircled{A} = \sum_j \partial_i F \cdot \partial_i (\partial_j H \partial_j F) = (\nabla F, \text{Hess } H \nabla F) + (\nabla F, \text{Hess } F \nabla H)$$

$$\frac{\partial}{\partial t} \frac{(\nabla F)^2}{F} = - \frac{1}{F^2} (\nabla F)^2 (\Delta F - (\nabla H, \nabla F)) + 2 \left( \frac{(\nabla F, \nabla \Delta F)}{F} \right)$$

$$- 2 \frac{1}{F} \left( (\nabla F, \text{Hess } H \nabla F) + (\nabla F, \text{Hess } F \nabla H) \right)$$

$$\Delta^H \frac{(\nabla F)^2}{F} = \Delta \frac{(\nabla F)^2}{F} - (\nabla H, \nabla) \frac{(\nabla F)^2}{F}$$

$$= \Delta \frac{(\nabla F)^2}{F} + \frac{1}{F^2} (\nabla F)^2 (\nabla H, \nabla F) - 2 \frac{(\nabla F, \text{Hess } F \nabla H)}{F}$$

$$\left( \frac{\partial}{\partial t} - \Delta^H \right) \frac{(\nabla F)^2}{F} = - \frac{1}{F^2} (\nabla F)^2 \Delta F + 2 \frac{(\nabla F, \nabla \Delta F)}{F}$$

$$- 2 \frac{1}{F} (\nabla F, \text{Hess } H \nabla F) - \Delta \frac{(\nabla F)^2}{F}$$

$$= - 2F (\nabla \log F, \text{Hess } H \nabla \log F) + R$$

$$R = - \frac{1}{F^2} (\nabla F)^2 \Delta F + \frac{2}{F} (\nabla F, \nabla \Delta F) - \Delta \frac{(\nabla F)^2}{F}$$

$$1X = \sum_i \partial_i \left( \frac{2(\nabla F, \nabla \partial_i F)}{F} - \frac{(\nabla F)^2}{F^2} \partial_i F \right)$$

$$= \frac{2}{F} (\nabla F, \nabla \Delta F) + \frac{2}{F} |\text{Hess } F|^2 - \frac{2^H}{F^2} \sum_i (\nabla F, \nabla \partial_i F) \partial_i F$$

$$- \frac{(\nabla F)^2}{F^2} \Delta F - \frac{2(\nabla F, \nabla \partial_i F) \partial_i F}{F^2} + 2 \frac{(\nabla F)^2}{F^3} (\nabla F, \nabla F)$$

$$R = -\frac{2}{F} |\text{Hess } F|_2^2 + \frac{4}{F^2} (\nabla F, \text{Hess } F \nabla F) - 2 \frac{(\nabla F, \nabla F)^2}{F^3} \quad \textcircled{2}$$

On the other hand,

$$2F |\text{Hess } \log F|_2^2 = 2F \sum_{ij} (\partial_i \partial_j \log F)^2$$

$$= 2F \sum_{ij} \left( \partial_i \left( \frac{\partial_j F}{F} \right) \right)^2 = 2F \sum_{ij} \left( \frac{1}{F} \partial_i \partial_j F - \frac{1}{F^2} \partial_i F \partial_j F \right)^2$$

$$= 2F \sum_{ij} \frac{1}{F^2} (\partial_i \partial_j F) (\partial_i \partial_j F)$$

$$- 2 \frac{1}{F^3} (\partial_i \partial_j F) \partial_i F \partial_j F + \frac{1}{F^4} \partial_i F \partial_j F \partial_i F \partial_j F$$

$$= \frac{2}{F} |\text{Hess } F|_2^2 - \frac{4}{F^2} (\nabla F, \text{Hess } F \nabla F) + \frac{2}{F^3} (\nabla F, \nabla F)^2$$

= R

Conclusion:

$$\left( \frac{\partial}{\partial t} - \Delta^H \right) \frac{(\nabla F)^2}{F} = -2F \underbrace{|\text{Hess } \log F|_2^2}_{\geq 0} - 2F \underbrace{(\nabla \log F, \text{Hess } H \log F)}_{\geq \lambda \text{Id}}$$

$$\leq -2\lambda F (\nabla \log F)^2 = -2\lambda \frac{(\nabla F)^2}{F}$$

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Conclusion:

$$\left( \frac{\partial}{\partial t} - \Delta^H \right) I_V(F_t) \leq -2\lambda \mathbb{E}_V \left[ \frac{(\nabla F_t)^2}{F_t} \right] = -2\lambda I_V(F_t)$$

$$\Rightarrow I_V(F_t) \leq e^{-2\lambda t} I_V(F_0)$$

The claim is proved.

The theorem is proved.

# Intro to Polchinski equation.

(1)

## 1. Gaussian integration.

Let  $C \in \mathbb{R}^{N \times N}$  be positive semi-definite, let  $P_C$  be the corresp. centred Gaussian measure with covariance  $C$ ,  $\text{supp}(P_C) = \text{Im}(C)$

$$(1) \quad E_C[F] \propto \int_{\mathbb{R}^N} e^{-\frac{1}{2}(\xi, C^{-1}\xi)} F(\xi) d\xi$$

Otherwise replace with  $\text{Im}(C)$

**Remark:** it is useful to work with semi-definite covariance matrices. Then one can consider (potentially) large- $N$  limits of random matrices.

**Semigroup property:** If  $C = C_1 + C_2$ ,

$$(2) \quad E_C[F(\xi)] = E_{C_2} [ E_{C_1} [ F(\xi_1 + \xi_2) ] ],$$

Positive s-definite

Independent

where  $\xi_i \sim N(0, C_i)$

# It is enough to check (2) for  $\xi \mapsto F(\xi) = e^{(\lambda, \xi)}$  for any  $\lambda \in \mathbb{R}^N$

$$\begin{aligned} \text{Then: } & E_{C_2} [ E_{C_1} [ F(\xi_1 + \xi_2) ] ] \\ &= E_{C_2} [ E_{C_1} [ e^{(\lambda, \xi_1 + \xi_2)} ] ] = E_{C_2} [ e^{(\lambda, \xi_2)} E_{C_1} [ e^{(\lambda, \xi_1)} ] ] \\ &= e^{\frac{1}{2}(\lambda, C_2 \lambda)} e^{\frac{1}{2}(\lambda, C_1 \lambda)} = e^{\frac{1}{2}(\lambda, (C_1 + C_2) \lambda)} \end{aligned}$$

$$= e^{\frac{1}{2}(A, eA)} = E_c [e^{(A, \beta)}]$$

(2)

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Notations. •  $\Delta_c := \sum_{i,j} c_{ij} \frac{\partial^2}{\partial x_i \partial x_j}$ ; (3)

(4) •  $(\cdot, \cdot)_c : \mathbb{R}^N \times \mathbb{R}^N \rightarrow \mathbb{R}$   
 $(u, v) \mapsto (u, v)_c = \sum_{i,j} c_{ij} u_i v_j$

•  $|u|_c^2 = (u)_c^2 = (u, u)_c$ .

•  $(u, v) = \sum_i u_i v_i$

any closed interval can be used

Principal object:  $t \in [0, +\infty] \mapsto C_t$  - continuous increasing function into positive semidefinite matrices on  $\mathbb{R}^N$  (increasing in the sense of quadratic forms). More precisely,

$$C_t = \int_0^t C_s ds,$$

where  $s \mapsto C_s$  is cadlag taking values in positive semidefinite matrices.

Say that  $C_\infty = \int_0^\infty C_s ds$  is a covariance decomposition of  $C_\infty$ .

Let  $X \subset \mathbb{R}^N$  be the image of  $C_\infty$ .

(Useful remark:  $(\text{image } C_{t_1}) \subset (\text{image } C_{t_2}) \subset X$   
for  $t_1 < t_2$ )

Let  $X_t$  be the image of  $C_t, t \geq 0$

Proposition 3.1 For  $F \in C^2(X, \mathbb{R})$ , define

$$F_t = P_{C_t} * F, \text{ i.e.}$$

$$F_t(\varphi) := E_{C_t} [F(\varphi + \xi)]$$

Then for all  $t$  which are not disc points

of  $C_t$ ,  
(5) or  
(6)

$$\frac{\partial F_t}{\partial t} = \frac{1}{2} \Delta_{C_t} F_t, F_0 = F$$

$$F_0(\varphi) = E_{C_0} [F(\varphi + \xi)] \stackrel{\xi=0 \text{ a.s.}}{=} F(\varphi) \text{ o.k.}$$

#

It  $t$  is not a disc point,  $\exists$

$\exists$  s.t.  $X_S$  is indep. of  $S \in [t-\epsilon, t+\epsilon]$ ,  
 $C_S$  is strictly positive definite for  
 and  $S \in [t-\epsilon, t+\epsilon]$ . Then (1) is true

on  $X_S$  for any such  $S$ , and (5) follows  
 from the defn of  $E_Q$

Alternatively, can use Ito's calculus:

Define

$$\xi_t = \varphi + \int_0^t \sqrt{C_s} dB_s \in \mathbb{R}^n, t \geq 0$$

Then  $F_t(\varphi) = E[F(\xi_t)]$  and

$$dF(\xi_t) = \Delta F(\xi_t) dt + \frac{1}{2} \Delta \otimes \Delta F(\xi_t) dt \otimes dt$$

$$= \Delta F(\xi_t) dt + \frac{1}{2} \Delta_{C_t} F(\xi_t) dt \Rightarrow$$

$\varphi$ -derivatives

$$F_t(\varphi) = \frac{1}{2} \Delta_{C_t} F_t(\varphi)$$

$$F_0(\varphi) = E[F(\xi_0)] = F(\varphi) \quad \#$$

## 2. Renormalized potential and the Polchinski equation.

Let  $\nu_0$  be a prob. measure on  $X \subset \mathbb{R}^N$

Let  $C_\infty = \int_0^\infty C_t dt$  be a covariance <sup>Linear</sup> decomp.

Suppose that the  $\nu_0$  expectation can be presented in the form of a Gaussian

(10) expectation: 
$$\mathbb{E}_{\nu_0}[F] = \frac{1}{Z} \mathbb{E}_{C_\infty} [e^{-V_0(\xi)} F(\xi)]$$

where  $V_0: X \rightarrow \mathbb{R}$  is a potential assumed to be bounded below

$$(Z = \mathbb{E}_{C_\infty} [e^{-V_0(\xi)}])$$

Definition 2. For  $t > s > 0$  and bounded  $F: X \rightarrow \mathbb{R}$  and  $\varphi \in X$ , define

• the renormalized potential  $V_t$ :

(11) 
$$V_t(\varphi) = -\log \mathbb{E}_{C_t} [e^{-V_0(\varphi+\xi)}]$$

• the Polchinski semigroup  $P_{s,t}$ :

(12) 
$$P_{s,t} F[\varphi] = e^{V_t(\varphi)} \mathbb{E}_{C_t - C_s} [e^{-V_s(\varphi+\xi)} F(\varphi+\xi)]$$

• the renormalized measure  $\nu_t$ :

(13) 
$$\begin{aligned} \mathbb{E}_{\nu_t}[F] &= P_{t,\infty} F(0) \\ &= e^{V_\infty(0)} \mathbb{E}_{C_\infty - C_t} [e^{-V_t(\xi)} F(\xi)] \end{aligned}$$



Remark. Consistency of (13):

$$\begin{aligned}
1 &= \mathbb{E}_{V_t} [1] = e^{V_{\infty}(0)} \mathbb{E}_{C_{\infty}-C_t} [e^{-V_t(\xi)}] \\
&= e^{V_{\infty}(0)} \mathbb{E}_{C_{\infty}-C_t} [\mathbb{E}_{C_t} [e^{-V_0(\xi_1+\xi_2)}]] \\
&\stackrel{\text{Defn of ren. pot}}{=} e^{V_{\infty}(0)} \underbrace{\mathbb{E}_{C_{\infty}} [e^{-V_{\infty}(\xi)}]}_{e^{-V_{\infty}(0)}} = 1 \quad \text{o.k.}
\end{aligned}$$

$\xi_1 \perp \xi_2$

Remark. The renormalized potential and moment generating function:

$$\begin{aligned}
V_{\infty}(C_{\infty}h) &\stackrel{(\#)}{=} -\log \mathbb{E}_{C_{\infty}} [e^{-V_0(C_{\infty}h+\xi)}] \\
&\stackrel{\text{Gaussian}}{\downarrow} -\log \mathbb{E}_{C_{\infty}} [e^{-V_0(\xi)} e^{(h,\xi)} e^{-\frac{1}{2}(h,C_{\infty}h)}] \\
&\stackrel{\text{change of variables}}{\xi \rightarrow \xi - C_{\infty}h} = \frac{1}{2}(h,C_{\infty}h) - \log \mathbb{E}_{C_{\infty}} [e^{-V_0(\xi)} e^{(h,\xi)}] \\
&= \frac{1}{2}(h,C_{\infty}h) - \log \underbrace{\mathbb{E}_{C_{\infty}} [e^{-V_0(\xi)}]}_{e^{-V_{\infty}(0)}} \mathbb{E}_{V_0} [e^{(h,\xi)}] \\
&= V_{\infty}(0) + \frac{1}{2}(h,C_{\infty}h) - \log \mathbb{E}_{V_0} [e^{(h,\xi)}]
\end{aligned}$$

Proposition 3. For  $t \geq 0$  and any  $F: X \rightarrow \mathbb{R}$  s. t. the following quantities make sense,

$$\mathbb{E}_{V_0} [F] = \mathbb{E}_{V_t} [P_{0,t} F]$$

Proof.

$$\begin{aligned}
E_{V_0}[F] &= \frac{1}{E_{C_{\infty}}[e^{-V_0(\xi)}]} E_{C_{\infty}}[e^{-V_0(\xi)} F(\xi)] \\
&= \frac{1}{E_{C_{\infty}}[e^{-V_0(\xi)}]} E_{C_{\infty}-C_t} \left[ E_{C_t} \left[ e^{-V_0(\xi+\varphi)} F(\xi+\varphi) \right] \right] \\
&= \frac{1}{E_{C_{\infty}}[e^{-V_0(\xi)}]} E_{C_{\infty}-C_t} \left[ e^{-V_t(\varphi)} P_{0,t} F(\varphi) \right] \equiv
\end{aligned}$$

$$\# P_{0,t} F(\varphi) = e^{V_t(\varphi)} E_{C_t} \left[ e^{-V_0(\xi+\varphi)} F(\xi+\varphi) \right] \#$$

$$\equiv \frac{e^{-V_{\infty}(0)}}{E_{C_{\infty}}[e^{-V_0(\xi)}]} E_{V_t}[F]$$

= 1 (defn of the RP)

QED

Remarks:

① (3.12)  $\Rightarrow P_{0,t} F(\varphi) = e^{V_t(\varphi)} E_{C_t} [e^{-V_0(\varphi+\xi)} F(\varphi+\xi)]$   
 $\stackrel{(3.11)}{=} E_{C_t} [e^{-V_0(\varphi+\xi)} F(\varphi+\xi)]$   
 Cond. expectation.  $\frac{E_{C_t} [e^{-V_0(\varphi+\xi)} F(\varphi+\xi)]}{E_{C_t} [e^{-V_0(\varphi+\xi)}]} := E_{\mu_t^\varphi} [F(\xi)]$

$\mu_t^\varphi$  is called the fluctuation measure.  
 Equivalently,

$$\# \frac{1}{Z} \int d\xi e^{-\frac{1}{2} \langle \xi, C_t^{-1} \xi \rangle - V_0(\varphi+\xi)} F(\varphi+\xi)$$

$$= \frac{1}{Z} \int d\xi e^{-\frac{1}{2} \langle \xi-\varphi, C_t^{-1} (\xi-\varphi) \rangle - V_0(\xi)} F(\xi)$$

$$= \frac{1}{Z} \int d\xi e^{-\frac{1}{2} \langle \xi, C_t^{-1} \xi \rangle - \frac{1}{2} \langle \varphi, C_t^{-1} \varphi \rangle - \langle \xi, C_t^{-1} \varphi \rangle - V_0(\xi)} F(\xi)$$

$$\mu_t^\varphi(d\xi) \sim e^{V_t(\varphi) - \frac{1}{2} \langle \varphi, C_t^{-1} \varphi \rangle} \# e^{-\frac{1}{2} \langle \xi, C_t^{-1} \xi \rangle + \langle \xi, C_t^{-1} \varphi \rangle - V_0(\xi)} d\xi$$

- similar to the original measure, but with  $C_\infty \rightarrow C_t < C_\infty \Rightarrow$  "More convex measure"

② Proposition 3.3  $\Rightarrow$

$$E_{V_0} [F] = E_{V_t} [P_{0,t} F]$$

$$\stackrel{\textcircled{1}}{=} E_{V_t} [E_{\mu_t^\varphi} [F(\xi)]]$$

Over  $\varphi$       Over  $\xi$  Cov.  $C_t$  "fast scales"  
 - an instance of measure decomposition, see section 2.5  
 Cov  $C_\infty - C_t$  "slow scales" nu:

Multiscale renormalization  $\rightarrow$   
 infinitesimal renormalization steps  $\rightarrow$   
 HJ equation for  $V_t$ , known as  
 Polchinski equation

Prop. 3.4. Let  $(C_t)_{t \geq 0}$  be a covariance decomposition, let  $V_0 \in C^2$ . Then  $\forall t$ :  $C_t$  is diff, the renorm. potential  $V_t$  satisfies the Polchinsky equation

$$\frac{\partial}{\partial t} V_t = \frac{1}{2} \Delta_{\dot{C}_t} V_t - \frac{1}{2} (\nabla V_t)^2_{\dot{C}_t}$$

Proof. Define  $Z_t(\varphi) = \mathbb{E}_{C_t} [e^{-V_0(\xi + \varphi)}]$

$$\text{Prop 3.1} \Rightarrow \begin{cases} \frac{\partial}{\partial t} Z_t = \frac{1}{2} \Delta_{\dot{C}_t} Z_t \\ Z_0 = e^{-V_0} \end{cases}$$

$Z_0 > 0 \Rightarrow Z_t > 0 \Rightarrow V_t = -\log Z_t$  is well defined and

$$\begin{aligned} \dot{V}_t &= -\frac{1}{Z_t} \dot{Z}_t = -\frac{1}{Z_t} \Delta_{\dot{C}_t} Z_t \\ &= -\frac{1}{2} e^{V_t} \Delta_{\dot{C}_t} e^{-V_t} \\ &= -\frac{1}{2} e^{V_t} \sum_{ij} (\dot{C}_t)_{ij} \partial_i \partial_j e^{-V_t} \\ &= \frac{1}{2} \Delta_{\dot{C}_t} V_t - \frac{1}{2} (\nabla V_t, \nabla V_t)_{\dot{C}_t} \end{aligned}$$

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**Proposition 3.5.** Let  $V_0$  be bounded below. The operators  $(P_{s,t})_{s \leq t}$  form a time-dependent Markov semigroup with generators  $(L_t)$  defined below in the sense that

$$\text{and } \begin{cases} P_{s,t} F \geq 0 & \text{if } F \geq 0, s \leq t \\ P_{s,t} 1 = 1 \end{cases} \quad \begin{aligned} P_{t,t} &= \text{Id} \\ P_{r,t} P_{s,r} &= P_{s,t}, s \leq r \leq t \end{aligned}$$

at all points  $t$  at which  $C_t$  is diff. ( $C_s$  is differentiable),

$$\begin{aligned} \frac{\partial}{\partial t} P_{s,t} F &= L_t P_{s,t} F \\ - \frac{\partial}{\partial s} P_{s,t} F &= P_{s,t} L_s F, \quad (s \leq t) \end{aligned}$$

where  $L_t F = \frac{1}{2} \Delta_{C_t} F - (\nabla V_t, \nabla F)_{C_t}$

The evolution of  $V_t$  is dual to  $(P_{s,t})$ :

$$\begin{cases} E_{V_t} [P_{s,t} F] = E_{V_s} [F] \quad (s \leq t) \\ - \frac{\partial}{\partial t} E_{V_t} [F] = E_{V_t} [L_t F] \end{cases}$$

**Proof.** Recall the definition:

$$(*) \quad P_{s,t} F(\varphi) = e^{V_t(\varphi)} E_{C_t - C_s} \left[ e^{-V_s(\varphi + \xi)} F(\varphi + \xi) \right]$$

$$\Rightarrow P_{t,t} F(\varphi) = \lim_{\substack{\text{SM} \\ \text{cont}}} e^{V_t(\varphi)} \underbrace{E_{C_t - C_s}}_{\substack{\text{Goes to} \\ \delta_0 \text{ measure}}} \left[ e^{-V_s(\varphi + \xi)} F(\varphi + \xi) \right] = e^{V_0(\varphi)} e^{-V_0(\varphi)} F(\varphi) = F(\varphi)$$

**Semigroup** - generalize the splitting argument used before:

$$P_{s,t} F(\varphi) = e^{V_t(\varphi)} \mathbb{E}_{c_t-c_s} \left[ e^{-V_s(\varphi+\xi)} F(\varphi+\xi) \right]$$

$$= e^{V_t(\varphi)} \mathbb{E}_{c_t-c_r+c_r-c_s} \left[ e^{-V_s(\varphi+\xi)} F(\varphi+\xi) \right]$$

$$= e^{V_t(\varphi)} \mathbb{E}_{c_t-c_r} \left[ \underbrace{\mathbb{E}_{c_r-c_s}}_{\sim \xi_1} \left[ e^{-V_s(\varphi+\xi_2+\xi_1)} F(\varphi+\xi_2+\xi_1) \right] \right]$$

$$= e^{V_t(\varphi)} \mathbb{E}_{c_t-c_r} \left[ e^{-V_r(\varphi+\xi_2)} P_{s,r} F(\varphi+\xi_2) \right]$$

$$= P_{r,t} P_{s,r} F(\varphi)$$

Positivity follows from the defn. (\*)  
Continuity: also from the definition, as

$$\| P_{s,t} F \|_\infty = \| e^{V_t(\varphi)} \mathbb{E}_{c_t-c_s} \left[ e^{-V_s(\varphi)} F(\varphi+\xi) \right] \|_\infty$$

$$= \| F \|_\infty \leq \| e^{V_t(\varphi)} \mathbb{E}_{c_t-c_s} \left[ e^{-V_s(\varphi)} \right] \|_\infty$$

$$\| F \|_\infty = \| F \|_\infty \quad (\text{Conclusion?})$$

$$\frac{\partial}{\partial t} P_{s,t} F(\varphi) = \partial_t V_t(\varphi) P_{s,t} F(\varphi)$$

$$+ e^{V_t} \Delta_{c_t} \mathbb{E}_{c_t-c_s} \left[ e^{-V_s(\xi+\varphi)} F(\varphi+\xi) \right]$$

$$= \partial_t V_t(\varphi) P_{s,t} F(\varphi) + e^{V_t} \frac{1}{2} \Delta_{c_t} e^{-V_t} (P_{s,t} F(\varphi))$$

$$\begin{aligned}
PE &= \left( \frac{1}{2} \Delta \dot{c}_t V_t - \frac{1}{2} (\nabla V_t)^2 \dot{c}_t \right) P_{s,t} F(\varphi) \\
&+ \frac{1}{2} \Delta c_t P_{s,t} F(\varphi) \\
&- \left( \nabla V_t, \nabla P_{s,t} F(\varphi) \right) \dot{c}_t \\
&- \frac{1}{2} (\Delta \dot{c}_t V_t) P_{s,t} F(\varphi) + \frac{1}{2} (\nabla V_t, \nabla V_t) \dot{c}_t P_{s,t} F(\varphi) \\
&= \underbrace{\left( \frac{1}{2} \Delta c_t - (\nabla V_t, \nabla) \dot{c}_t \right)}_{L_t} P_{s,t} F(\varphi)
\end{aligned}$$

$\frac{\partial}{\partial s}$  - derivative - similar.

(3.27) is a gen of Prop 3.3:

$$\begin{aligned}
\mathbb{E}_{V_0} [F] &= \mathbb{E}_{V_t} [P_{0,t} F] \xrightarrow{\text{Gen}} \mathbb{E}_{V_t} [P_{s,t} F] \\
&= \mathbb{E}_{V_s} [F]
\end{aligned}$$

The last statement follows by diff:

$$\begin{aligned}
\frac{\partial}{\partial s} \mathbb{E}_{V_s} [F] &= \frac{\partial}{\partial s} \mathbb{E}_{V_t} [P_{s,t} F] \\
&= \mathbb{E}_{V_t} \left[ \frac{\partial}{\partial s} P_{s,t} F \right] \\
&= - \mathbb{E}_{V_t} [P_{s,t} L_s F]
\end{aligned}$$

$$\text{Take } s=t: \frac{\partial}{\partial t} \mathbb{E}_{V_t} [F] = - \mathbb{E}_{V_t} [L_t F]$$

Done.