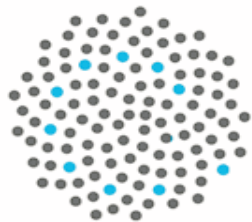


# Reversing time as an analytical tool: Isn't that just Radon-Nikodym?

Phil Pollett

<http://www.maths.uq.edu.au/~pkp>



AUSTRALIAN RESEARCH COUNCIL  
Centre of Excellence for Mathematics  
and Statistics of Complex Systems

# *Menuetto al rovesci*

Musical score for measures 1 through 10 of the piano part. The score is written in treble and bass clefs, with a key signature of three sharps (F#, C#, G#) and a 3/4 time signature. Measures 1-5 are on the top system, and measures 6-10 are on the bottom system. The music features a simple, rhythmic melody in the right hand and a supporting bass line in the left hand. Measure numbers 1 through 10 are indicated in small boxes above the notes.

Musical score for measures 11 through 20 of the piano part. The score is written in treble and bass clefs, with a key signature of three sharps (F#, C#, G#) and a 3/4 time signature. Measures 11-15 are on the top system, and measures 16-20 are on the bottom system. The music continues the simple, rhythmic melody and supporting bass line. Measure numbers 11 through 20 are indicated in small boxes above the notes.

Joseph Haydn's Sonata No. 4 for Violin and Piano (piano part only) *Menuetto al rovescio*

# Motet *Diliges Dominum*

The image displays two systems of musical notation for a piano accompaniment. Each system consists of a grand staff with a treble clef on the upper staff and a bass clef on the lower staff. The key signature is three flats (B-flat, E-flat, A-flat), and the time signature is 3/2. The first system begins with a left-pointing arrow above the treble staff, and the second system begins with a right-pointing arrow above the treble staff. The music is written in a style characteristic of the English Renaissance, featuring a steady harmonic accompaniment with a mix of chords and moving lines.

William Byrd's motet *Diliges Dominum*

# Hammerklavier Sonata

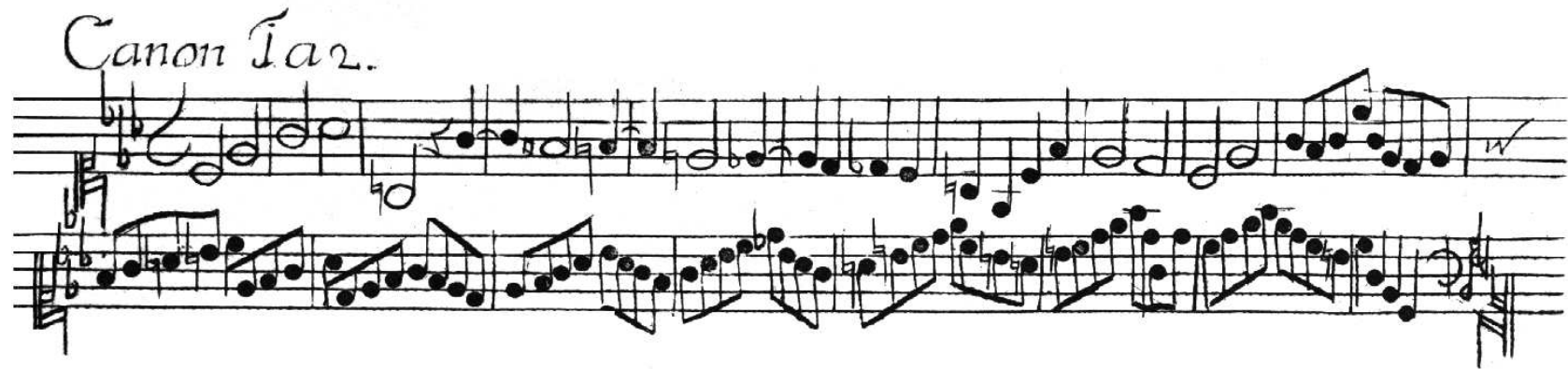
16 *tr*

21 153

*tr*

Beethoven's Piano Sonata No. 29 in B flat, Op. 106  
("Hammerklavier"), Last movement (fugue) *Allegro risoluto*

# Canon *cancrizans*



J.S. Bach's *Das Musikalische Opfer* (The Musical Offering),  
BWV 1079, Canon 1. a 2. *cancrizans*

# Canon *cancrizans*

The first system of musical notation for 'Canon cancrizans' consists of two staves. The upper staff is in treble clef, and the lower staff is in bass clef. The key signature has one flat (B-flat), and the time signature is common time (C). The music begins with a quarter rest in the upper staff, followed by a half note G4, a quarter note A4, and a half note Bb4. The lower staff features a continuous eighth-note accompaniment starting on G3, with various accidentals including sharps and naturals.

The second system of musical notation starts at measure 5. The upper staff continues with a half note Bb4, a quarter note C5, a half note Bb4, and a quarter note A4. The lower staff continues with the eighth-note accompaniment, showing a sequence of notes including G3, A3, Bb3, and C4.

The third system of musical notation starts at measure 9. The upper staff continues with a half note A4, a quarter note G4, a half note F4, and a quarter note E4. The lower staff continues with the eighth-note accompaniment, showing a sequence of notes including G3, A3, Bb3, and C4.

The fourth system of musical notation starts at measure 14. The upper staff continues with a half note D4, a quarter note C4, a half note Bb4, and a quarter note A4. The lower staff continues with the eighth-note accompaniment, showing a sequence of notes including G3, A3, Bb3, and C4.

# Setting

$(\Omega, \mathcal{F}, \mathbb{P})$  is our carrier triple.

$(X_t, t \in T)$  will denote a *stochastic process* with (ordered) *parameter set*  $T$  and *state space*  $(E, \mathcal{E})$ . ( $T$  would usually be “time”:  $\mathbb{Z}$  or  $\mathbb{Z}_+$ , or,  $\mathbb{R}$  or  $\mathbb{R}_+$ .)

The “elementary picture” is: for each  $t \in T$ ,

$$X_t : \Omega \rightarrow E \quad \text{and} \quad X_t^{-1} : \mathcal{E} \rightarrow \mathcal{F},$$

with  $X_t$  with  $\mathcal{F}$ -measurable.

We shall assume that  $\mathcal{E}$  includes all point sets of  $E$ , that is, for all  $x \in E$ ,  $\{x\} \in \mathcal{E}$ . At this stage, we make no further topological assumptions about the measurable space  $(E, \mathcal{E})$ .

# The time reverse process

**Definition.** Let  $(X_t, t \in T)$  and  $(X_t^*, t \in T)$  be two stochastic processes with the same parameter set  $T$  and the same state space  $(E, \mathcal{E})$ . We say that  $X^*$  is a *time reverse* of  $X$  if, for any finite sequence  $t_1 < t_2 < \dots < t_n$  in  $T$  such that

$$t_n - t_{n-1} = t_2 - t_1, t_{n-2} - t_{n-1} = t_3 - t_2, \dots,$$

and for any  $A_1, A_2, \dots, A_n \in \mathcal{E}$ ,

$$\mathbb{P}(X_{t_1} \in A_1, \dots, X_{t_n} \in A_n) = \mathbb{P}(X_{t_1}^* \in A_n, \dots, X_{t_n}^* \in A_1).$$

We say that  $X$  is *time reversible* if

$$\mathbb{P}(X_{t_1} \in A_1, \dots, X_{t_n} \in A_n) = \mathbb{P}(X_{t_1} \in A_n, \dots, X_{t_n} \in A_1).$$



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In particular, for all  $A, B \in \mathcal{E}$  and  $t, u \in T$ ,

$$\mathbb{P}(X_t \in A, X_u \in B) = \mathbb{P}(X_t^* \in B, X_u^* \in A).$$

On taking  $B = E$ , we see that  $\mathbb{P}(X_t \in A) = \mathbb{P}(X_u^* \in A)$ , which implies  $\pi(A) := \mathbb{P}(X_t \in A) = \mathbb{P}(X_t^* \in A)$  (*the same for all  $t$* ).

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**Conclusion.** The above definition only makes sense if  $X^*$  and  $X$  are *stationary* with the same *stationary law*  $\pi$ .

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**Conclusion.** The above definition only makes sense if  $X^*$  and  $X$  are *stationary* with the same *stationary law*  $\pi$ .

**Example.** Brownian motion has no time reverse.

**Exercise.** Think of a diffusion that *does* have a time reverse.

# Transition function

$$p_t(x, A) = \mathbb{P}(X_{s+t} \in A | X_s = x)$$

# The time reverse of a Markov process

**Definition.** If  $(E, \mathcal{E})$  is a measurable space, then a *transition function*  $p = (p_t, t \geq 0)$  on  $(E, \mathcal{E})$  is a family of mappings  $p_t : E \times \mathcal{E} \rightarrow \mathbb{R}_+$  with the following properties:

- (1) for all  $A \in \mathcal{E}$ ,  $p_t(\cdot, A)$  is an  $\mathcal{E}$ -measurable function,
- (2) for all  $x \in E$ ,  $p_t(x, \cdot)$  is a subprobability measure on  $(E, \mathcal{E})$  (that is, a measure on  $(E, \mathcal{E})$  with  $p_t(x, E) \leq 1$ ),
- (3) the *Chapman-Kolmogorov equation* holds, that is, for all  $x \in E$  and  $A \in \mathcal{E}$ ,  $p_{s+t}(x, A) = \int_E p_s(x, dy)p_t(y, A)$ ,  $s, t \geq 0$ , and

The transition function  $p$  is called *honest* if, for all  $x \in E$  and  $t \geq 0$ ,  $p_t(x, \cdot)$  is a probability measure ( $p_t(x, E) = 1$ ).

# The time reverse of a Markov process

It is “usual” to have  $p_0(x, A) = I_A(x)$  ( $x \in E, A \in \mathcal{E}$ ), *but we certainly do not require*  $\lim_{t \downarrow 0} p_t(x, A) = I_A(x)$ .

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**Interpretation.** For an honest transition function  $p$  there always exists a (time-homogeneous) *Markov process*  $(X_t, t \geq 0)$  with  $p_t(x, A) = \mathbb{P}(X_{s+t} \in A | X_s = x)$  ( $s, t \geq 0, A \in \mathcal{E}$ ).

(If  $p$  is dishonest, then we can append a coffin state  $\partial$  making  $p$  honest over  $(E^\partial, \mathcal{E}^\partial)$ , where  $E^\partial = E \cup \partial$ .)



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If  $X$  has *stationary law*  $\pi$ , that is,  $\mathbb{P}(X_s \in A) = \pi(A)$ ,  $s \geq 0$ , then, by Total Probability,

$$\mathbb{P}(X_s \in A, X_{t+s} \in B) = \int_A \pi(dx) p_t(x, B) \quad (s, t \geq 0).$$

# The time reverse of a Markov process

**Theorem 1.** Let  $(X_t, t \geq 0)$  and  $(X_t^*, t \geq 0)$  be two Markov processes on the same state space  $(E, \mathcal{E})$  with transition functions  $p$  and  $p^*$ , respectively. Then,  $X^*$  is the time reverse of  $X$  *if and only if*

- (1)  $X$  and  $X^*$  are stationary with the same stationary law  $\pi$ .
- (2)  $p^*$  is the reverse of  $p$  with respect to  $\pi$ .

In particular (**corollary!**),  $X$  is time reversible *if and only if*  $X$  is stationary with stationary law  $\pi$  and  $p$  is reversible with respect to  $\pi$ .

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So, what do I mean by “ $p^*$  is the reverse of  $p$  with respect to  $\pi$ ” and “ $p$  is reversible with respect to  $\pi$ ”?

# The reverse transition function

**Definition.** Let  $p$  and  $p^*$  transition functions on the same measurable space  $(E, \mathcal{E})$ , and let  $m$  be a measure on  $(E, \mathcal{E})$ . Then,  $p^*$  is the *reverse of  $p$  with respect to  $m$*  if

$$\int_B m(dx) p_t(x, A) = \int_A m(dx) p_t^*(x, B) \quad (A, B \in \mathcal{E}, t \geq 0).$$

If  $p$  is its own reverse with respect to  $m$ , that is,

$$\int_B m(dx) p_t(x, A) = \int_A m(dx) p_t(x, B) \quad (A, B \in \mathcal{E}, t \geq 0),$$

then  $p$  is said to be *reversible with respect to  $m$* .

# The reverse transition function

**Some implications.** Putting  $B = E$  we get

$$\int_E m(dx) p_t(x, A) = \int_A m(dx) p_t^*(x, E) \quad (A \in \mathcal{E}, t \geq 0).$$

Since  $p^*$  is a transition function,  $p_t^*(x, \cdot)$  is a subprobability measure, we get

$$\int_E m(dx) p_t(x, A) \leq \int_A m(dx) = m(A) \quad (A \in \mathcal{E}, t \geq 0).$$

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We say that  $m$  is a *subinvariant measure* for  $p$ . Moreover,  $m$  is an *invariant measure* for  $p$ , that is equality holds,

$$\int_E m(dx) p_t(x, A) = m(A) \quad (A \in \mathcal{E}, t \geq 0),$$

if  $p^*$  is honest.

# The reverse transition function

Conversely, if  $m$  is invariant for  $p$ , then

$$\int_A m(dx) = m(A) = \int_E m(dx) p_t(x, A) = \int_A m(dx) p_t^*(x, E)$$

for all  $A \in \mathcal{E}$  and  $t \geq 0$ , that is,

$$\int_A m(dx) (1 - p_t^*(x, E)) \geq 0 \quad (A \in \mathcal{E}, t \geq 0),$$

So, if  $m$  is, additionally, a  $\sigma$ -finite measure, we may apply Radon-Nikodym to show<sup>a</sup> that  $p^*$  is  $m - a.e.$  *honest*, that is, for all  $t > 0$ ,  $p_t^*(x, E) = 1$  for  $m$ -*almost all*  $x \in E$ .

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<sup>a</sup>I will write out the argument carefully later



# Markov chains

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**Exercise.** Let  $P = (p(i, j), i, j \in E)$  be a transition matrix and let  $m = (m(j), j \in E)$  be a collection of positive numbers. Define  $P^* = (p^*(i, j), i, j \in E)$  by  $p^*(i, j) = m(j)p(j, i)/m(i)$  ( $i, j \in E$ ). Show that  $P^*$  is a transition matrix whenever  $m$  is invariant for  $P$ , that is,

$$\sum_{j \in E} m(j)p(j, i) = m(i) \quad (i \in E).$$

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**Exercise.** Agree that the  $n$ -step transition matrices bear the same relationship:  $m(i)p_n^*(i, j) = m(j)p_n(j, i)$  ( $i, j \in E$ ).

# The reverse transition function

**Question.** Given a transition function  $p$  and a subinvariant measure  $m$  on  $(E, \mathcal{E})$ , can we always find a transition function  $p^*$  on  $(E, \mathcal{E})$  that is the reverse of  $p$  with respect to  $m$ ? That is,

$$\int_B m(dx) p_t(x, A) = \int_A m(dx) p_t^*(x, B) \quad (A, B \in \mathcal{E}, t \geq 0).$$

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**Exercise.** (Hint!) Let  $m$  be a  $\sigma$ -finite measure on  $(E, \mathcal{E})$ . Show that, for every  $B \in \mathcal{E}$  with  $m(B) < \infty$ ,  $\mu_B(\cdot) := \int_B m(dx) p_t(x, \cdot)$  is a finite measure on  $(E, \mathcal{E})$ .

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**Exercise.** (Bigger hint!) Let  $m$  be a  $\sigma$ -finite measure on  $(E, \mathcal{E})$  that is subinvariant for  $p$ . Show that, for every  $B \in \mathcal{E}$ ,  $\mu_B$  is absolutely continuous with respect to  $m$ .

# The time reverse of a Markov process

**Theorem 1.** Let  $(X_t, t \geq 0)$  and  $(X_t^*, t \geq 0)$  be two Markov processes on the same state space  $(E, \mathcal{E})$  with transition functions  $p$  and  $p^*$ , respectively. Then,  $X^*$  is the time reverse of  $X$  *if and only if*

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**Proof.** Suppose  $X^*$  is the time reverse of  $X$ .



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**Proof.** Suppose  $X^*$  is the time reverse of  $X$ . We have already seen that  $X$  and  $X^*$  are necessarily stationary with the same stationary law  $\pi$ , and that

$$\mathbb{P}(X_s \in A, X_{t+s} \in B) = \int_A \pi(dx) p_t(x, B) \quad (A, B \in \mathcal{E}, s, t \geq 0).$$

# The time reverse of a Markov process

Thus, for all  $A, B \in \mathcal{E}$ ,  $s, t \geq 0$ ,

$$\begin{aligned}\int_A \pi(dx) p_t(x, B) &= \mathbb{P}(X_s \in A, X_{t+s} \in B) \\ &= \mathbb{P}(X_s^* \in B, X_{t+s}^* \in A) \\ &= \int_B \pi(dx) p_t^*(x, A).\end{aligned}$$

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Conversely, if (1) and (2) hold, then, as we have already seen,  $\pi$  is an invariant measure for  $p$  (and for  $p^*$ ):

$$\int_E \pi(dx) p_t(x, B) = \pi(B), \quad \pi(A) = \int_E \pi(dx) p_t^*(x, A).$$

# The time reverse of a Markov process

Therefore, for any  $t_1 < t_2 < \dots < t_n$  in  $T$  such that  $t_n - t_{n-1} = t_2 - t_1$ ,  $t_{n-2} - t_{n-1} = t_3 - t_2, \dots$ , and for any  $A_1, A_2, \dots, A_n \in \mathcal{E}$ ,

$$\mathbb{P}(X_{t_1} \in A_1, \dots, X_{t_n} \in A_n) = \int_E \pi(dx) \int_{A_1} p_{t_1}(x, dx_1) \int_{A_2} p_{t_2-t_1}(x_1, dx_2) \dots \int_{A_n} p_{t_n-t_{n-1}}(x_{n-1}, dx_n).$$

Since  $\pi$  is invariant for  $p$ , this becomes

$$\begin{aligned} & \int_{A_1} \pi(dx_1) \int_{A_2} p_{t_2-t_1}(x_1, dx_2) \dots \int_{A_n} p_{t_n-t_{n-1}}(x_{n-1}, dx_n) \\ &= \int_{A_n} \pi(dx_n) \int_{A_{n-1}} p_{t_n-t_{n-1}}^*(x_n, dx_{n-1}) \dots \int_{A_1} p_{t_2-t_1}^*(x_2, dx_1), \end{aligned}$$

repeatedly applying  $\int_A \pi(dx) p_t(x, B) = \int_B \pi(dx) p_t^*(x, A)$ .

# The time reverse of a Markov process

Now use  $t_n - t_{n-1} = t_2 - t_1$ ,  $t_{n-2} - t_{n-1} = t_3 - t_2, \dots$ , together with the fact that  $\pi$  is invariant for  $p^*$ , to complete the proof:

$$\begin{aligned} & \mathbb{P}(X_{t_1} \in A_1, \dots, X_{t_n} \in A_n) \\ &= \int_{A_n} \pi(dx_n) \int_{A_{n-1}} p_{t_n - t_{n-1}}^*(x_n, dx_{n-1}) \cdots \int_{A_1} p_{t_2 - t_1}^*(x_2, dx_1) \\ &= \int_{A_n} \pi(dx_1) \int_{A_{n-1}} p_{t_n - t_{n-1}}^*(x_1, dx_2) \cdots \int_{A_1} p_{t_2 - t_1}^*(x_{n-1}, dx_n) \\ &= \int_{A_n} \pi(dx_1) \int_{A_{n-1}} p_{t_2 - t_1}^*(x_1, dx_2) \cdots \int_{A_1} p_{t_n - t_{n-1}}^*(x_{n-1}, dx_n) \\ &= \mathbb{P}(X_{t_1}^* \in A_n, \dots, X_{t_n}^* \in A_1). \end{aligned}$$

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**Example.** The *Ornstein-Uhlenbeck (OU) process*.

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**Example.** The *Ornstein-Uhlenbeck (OU) process*. It is a (Gaussian) diffusion process  $(Z_t, t \geq 0)$  on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  that satisfies

$$dZ_t = -\beta Z_t dt + \sigma dB_t \quad (t \geq 0),$$

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where  $(B_t, t \geq 0)$  is standard Brownian motion on  $(\mathbb{R}, \mathcal{B}(\mathbb{R}))$  and  $\beta$  and  $\sigma$  are positive constants.

Its transition function  $p$  is absolutely continuous with respect to Lebesgue measure<sup>a</sup> in that  $p_t(x, A) = \int_A p_t(x, y) dy$ , where  $p_t(x, y)$  (the *transition density*) is the Gaussian density with mean  $x e^{-\beta t}$  and variance  $\sigma^2(1 - e^{-2\beta t})/(2\beta)$ .

---

<sup>a</sup>True for all diffusions!



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**Exercise.** Show that  $\pi$  given by  $\pi(A) = \int_A \phi(y) dy$ , where  $\phi$  is the Gaussian density with mean 0 and variance  $\sigma^2/(2\beta)$ , is an invariant probability measure for  $p$ .

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You will need to verify that

$$\int_{\mathbb{R}} \pi(dx) p_t(x, A) = \pi(A),$$

or, equivalently, for the transition density,

$$\int_{-\infty}^{\infty} \phi(x) p_t(x, y) dx = \phi(y) \quad (t \geq 0, y \in \mathbb{R}).$$

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Also, by Theorem 1, our process  $Z$  (the stationary OU process) is *time reversible*.

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$$p_t(x, y) = (2\pi t)^{-n/2} \exp(-|y - x|^2/2t) \quad (x, y \in \mathbb{R}^n)$$

satisfies  $p_t(x, y) = p_t(y, x)$ , and so its transition function  $p$  is *reversible with respect to Lebesgue measure* (and hence invariant for  $p$ ).

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For the complete story on *reversible diffusions* (not necessarily time-reversible!), see John Kent's 1978 paper\*.

\*Kent, J. (1978) Time-reversible diffusions, Adv. Appl. Probab. 10, 819–835.

# The reverse transition function

## The reverse transition function as an analytical tool.

Suppose that we are given a transition function  $p$  and measure  $m$  on  $(E, \mathcal{E})$ . If we can determine the reverse transition function, that is, a transition function  $p^*$  on  $(E, \mathcal{E})$  satisfying  $\int_B m(dx) p_t(x, A) = \int_A m(dx) p_t^*(x, B)$ ,  $A, B \in \mathcal{E}$ ,  $t \geq 0$ , then, as already remarked,  $m$  will be subinvariant for  $p$  and invariant for  $p$  if  $p^*$  is honest (and only if  $p^*$  is  $m - a.e.$  honest).

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Of course, in most cases, we would be given more fundamental information, such as the *diffusion coefficients* (diffusions), or the *transition rates* (chains). We might hope to be able to establish the existence, and then the honesty of  $p^*$  *without actually exhibiting  $p^*$  explicitly*.

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Show that, for every  $B \in \mathcal{E}$  with  $m(B) < \infty$ ,

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**Solution.** We are told that  $\int_E m(dx) p_t(x, A) \leq m(A)$  ( $A \in \mathcal{E}$ ). So,  $\mu_B(A) \leq m(A)$ , and hence  $m(N) = 0 \Rightarrow \mu_B(N) = 0$ .

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*We're in business!* Radon-Nikodym provides us with an  $\mathcal{E}$ -measurable ( $m$ -integrable) non-negative function  $f$  define uniquely  $m - a.e.$  by  $\mu_B(A) = \int_A f(x)m(dx)$  ( $A \in \mathcal{E}$ ).

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Writing this out carefully: if  $m$  is a  $\sigma$ -finite measure that is subinvariant for  $p$ , then, for all  $t \geq 0$ , and for every  $B \in \mathcal{E}$  with  $m(B) < \infty$ , there is an  $\mathcal{E}$ -measurable  $f_t(\cdot, B)$  that satisfies

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**Why?**

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Second, we need to show

(2) that  $f_t(x, \cdot)$  is a subprobability measure on  $(E, \mathcal{E})$ , and

(3) that  $f$  satisfies the Chapman-Kolmogorov equation

$$f_{s+t}(x, A) = \int_E f_s(x, dy) f_t(y, A),$$

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And third, we need a refinement of the usual definition of a transition function (because our reverse transition function can at best be known  $m - a.e.$  uniquely), which was ...

# The reverse transition function

**Definition.** If  $(E, \mathcal{E})$  is a measurable space, then a transition function  $p = (p_t, t \geq 0)$  on  $(E, \mathcal{E})$  is a family of mappings  $p_t : E \times \mathcal{E} \rightarrow \mathbb{R}_+$  with the following properties:

(1) for all  $A \in \mathcal{E}$ ,  $p_t(\cdot, A)$  is an  $\mathcal{E}$ -measurable function,

(2) for all  $x \in E$ ,  $p_t(x, \cdot)$  is a subprobability measure on  $(E, \mathcal{E})$  (that is, a measure on  $(E, \mathcal{E})$  with  $p_t(x, E) \leq 1$ ),

(3) the Chapman-Kolmogorov equation holds, that is, for all  $x \in E$  and  $A \in \mathcal{E}$ ,  $p_{s+t}(x, A) = \int_E p_s(x, dy)p_t(y, A)$ ,  $s, t \geq 0$  (unmarked sums shall be over  $E$ ), and

The transition function  $p$  is called honest if, for all  $x \in E$  and  $t \geq 0$ ,  $p_t(x, \cdot)$  is a probability measure ( $p_t(x, E) = 1$ ).

# The reverse transition function

**Definition.** If  $(E, \mathcal{E})$  be a measurable space and let  $m$  be a  $\sigma$ -finite measure on  $(E, \mathcal{E})$ . Then, an  $m$ -a.e. *transition function*  $p = (p_t, t \geq 0)$  on  $(E, \mathcal{E})$  is a family of mappings  $p_t : E \times \mathcal{E} \rightarrow \mathbb{R}_+$  with the usual properties (1)–(3), *but (2) and (3) are required to hold for  $m$ -almost all  $x \in E$ .*

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Next we will show that our candidate reverse transition function  $f$  satisfies (2) and (3) assuming that our subinvariant measure  $m$  is a *finite measure*—thus avoiding technicalities.

# The reverse transition function

Recall that  $f$  is determined  $m - a.e.$  uniquely: for all  $t \geq 0$ , and for every  $B \in \mathcal{E}$ , there is an  $\mathcal{E}$ -measurable  $f_t(\cdot, B)$  that satisfies

$$\int_B m(dx) p_t(x, A) = \int_A m(dx) f_t(x, B) \quad (A \in \mathcal{E}).$$

*Remember we are assuming that  $m(B) < \infty$  for all  $B \in \mathcal{E}$ .*

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First we prove that for  $m$ -almost all  $x$ ,  $f_t(x, \cdot)$  is a subprobability measure on  $(E, \mathcal{E})$ , that is, a measure on  $(E, \mathcal{E})$  with  $f_t(x, E) \leq 1$ .

# The reverse transition function

**Claim.**  $f_t(x, E) \leq 1 \dots\dots$



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Take  $B = E$  in the definition of  $f$ :

$$\int_E m(dx) p_t(x, A) = \int_A m(dx) f_t(x, E) \quad (A \in \mathcal{E}, t \geq 0).$$

So, for all  $A \in \mathcal{E}$ ,

$$\int_A m(dx) f_t(x, E) = \int_E m(dx) p_t(x, A) \leq m(A) = \int_A m(dx),$$

and hence

$$\int_A m(dx) (1 - f_t(x, E)) = \int_A m(dx) - \int_A m(dx) f_t(x, E) \geq 0.$$

# The reverse transition function

It follows that  $\int_{(\cdot)} m(dx) (1 - f_t(x, E))$  is a totally finite positive measure on  $(E, \mathcal{E})$ .

It is absolutely continuous with respect to  $m$ , because if  $N$  is an  $m$ -null set in  $\mathcal{E}$ , then

$$(0 \leq) \int_N m(dx) (1 - f_t(x, E)) \leq \int_N m(dx) = m(N) = 0,$$

and hence it has the  $m - a.e.$  uniquely determined Radon-Nikodym derivative  $1 - f_t(x, E)$ .

Since the latter is  $m - a.e.$  unique, it is therefore  $m - a.e.$  positive, that is,  $f_t(x, E) \leq 1$ , for  $m$ -almost all  $x \in E$ .

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**Claim.**  $f_t(x, \cdot)$  is a measure for  $m$ -almost all  $x$  . . . . .

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First,

$$0 = \int_{\emptyset} m(dx) p_t(x, A) = \int_A m(dx) f_t(x, \emptyset) \quad (A \in \mathcal{E}).$$

and hence, by the Radon-Nikodym Theorem,  $f_t(x, \emptyset) = 0$  for  $m$ -almost all  $x$ .

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We already have  $f_t(\cdot, B) \geq 0$ ,  $m - a.e.$ , so we only need to check  $\sigma$ -additivity.

# The reverse transition function

Let  $B_1, B_2, \dots$  be a sequence of disjoint sets in  $\mathcal{E}$ . Using the definition of  $f$  and Fubini, we have, for all  $A \in \mathcal{E}$ ,

$$\begin{aligned}\int_A m(dx) f_t(x, \cup_j B_j) &= \int_{(\cup_j B_j)} m(dx) p_t(x, A) \\ &= \sum_j \int_{B_j} m(dx) p_t(x, A) \\ &= \sum_j \int_A m(dx) f_t(x, B_j) \\ &= \int_A m(dx) \sum_j f_t(x, B_j) (< \infty),\end{aligned}$$

That is,  $\int_A m(dx) \left( f_t(x, \cup_j B_j) - \sum_j f_t(x, B_j) \right)$ , for all  $A \in \mathcal{E}$ , and hence  $f_t(\cdot, \cup_j B_j) = \sum_j f_t(\cdot, B_j)$ ,  $m - a.e.$  (again by the Radon-Nikodym Theorem).

# The reverse transition function

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Since  $\int_B m(dx) p_t(x, A) = \int_A m(dx) f_t(x, B)$  for all  $A, B \in \mathcal{E}$ , we have that

$$\begin{aligned}\int_A m(dx) f_{s+t}(x, B) &= \int_B m(dx) p_{s+t}(x, A) \\ &= \int_B m(dx) \int_E p_t(x, dy) p_s(y, A) \\ &= \int_A m(dx) \int_E f_s(x, dy) f_t(y, B).\end{aligned}$$

The Radon-Nikodym Theorem then tells us that

$$f_{s+t}(\cdot, B) = \int_E f_s(\cdot, dy) f_t(y, B), \quad m - a.e.$$



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Finally, we tackle the Chapman-Kolmogorov equation . . . . .

Since  $\int_B m(dx) p_t(x, A) = \int_A m(dx) f_t(x, B)$  for all  $A, B \in \mathcal{E}$ , we have that

$$\begin{aligned}\int_A m(dx) f_{s+t}(x, B) &= \int_B m(dx) p_{s+t}(x, A) \\ &= \int_B m(dx) \int_E p_t(x, dy) p_s(y, A) \\ &= \int_A m(dx) \int_E f_s(x, dy) f_t(y, B).\end{aligned}$$

The Radon-Nikodym Theorem then tells us that

$$f_{s+t}(\cdot, B) = \int_E f_s(\cdot, dy) f_t(y, B), \quad m - a.e.$$

We have proved the following simple result.

# The reverse transition function

**Proposition 1.** Let  $p$  be a transition function on a measurable space  $(E, \mathcal{E})$  and suppose that  $m$  is a totally finite measure on  $(E, \mathcal{E})$  that is subinvariant for  $p$ . Then there exists an  $m - a.e.$  transition function  $p^*$  which is the  $m - a.e.$  unique reverse of  $p$  with respect to  $m$ .

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I am happy to report the following pleasing result.

# The reverse transition function

**Theorem 2.** Let  $(E, \mathcal{O}, \mathcal{E})$  be an inner-regular, measurable topological space with a countable basis. Let  $p$  be a transition function on  $(E, \mathcal{E})$  and suppose that  $m$  is a  $\sigma$ -finite measure on  $(E, \mathcal{E})$  that is subinvariant for  $p$ . Then, there exists an  $m - a.e.$  transition function  $p^*$  which is the  $m - a.e.$  unique reverse of  $p$  with respect to  $m$ .

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What is  $(E, \mathcal{O}, \mathcal{E})$  and why?

# An application

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**Corollary.** Let  $(E, \mathcal{O}, \mathcal{E})$  be an inner-regular, measurable topological space with a countable basis. Let  $p$  be a transition function on  $(E, \mathcal{E})$  and suppose that  $m$  is a  $\sigma$ -finite measure on  $(E, \mathcal{E})$  that is subinvariant for  $p$ . Let  $p^*$  be the  $m - a.e.$  unique reverse transition function with respect to  $m$ . Then,  $m$  is invariant for  $p$  *if and only if*  $p^*$  is  $m - a.e.$  honest.



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We extend this from  $(E, \mathcal{E}_0, m)$  to  $(E, \mathcal{E}, m)$  by approximating  $(E, \mathcal{E}, m)$  by *finite* measure spaces  $(E, \mathcal{E}_n, \mu)$ ,  $n = 1, 2, \dots$

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**The idea.** Statements (like the one immediately above) concerning a given  $\sigma$ -finite measure which hold over  $\mathcal{E}_0$  are extended to  $\mathcal{E}$ , for they are shown to hold over each of the  $\sigma$ -algebras  $\{\mathcal{E}_n\}_{n=1}^{\infty}$  defined by  $\mathcal{E}_n = \{A \cap E_n : A \in \mathcal{E}\}$ .

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For example, our elementary argument (above) gives us an  $f^{(n)}$  for each  $n$  with the right properties. We then set

$$p_t^*(x, A) = \sum_{n=1}^{\infty} f_t^{(n)}(x, A \cap E_n) \quad (x \in E, A \in \mathcal{E}, t \geq 0)$$

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and hope!! The details are quite tough, and rely on us being able to exploit the inner regularity of measures relative to compact sets that our topological structure permits.



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If  $\nu$  is a finite measure on  $(E, \mathcal{E})$ , then  $\nu$  is said to be *inner regular* if, for all  $A \in \mathcal{E}$ ,  $\nu(A) = \sup \{ \nu(K) : \text{compact } K \subset A \}$ .

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If *every* finite measure on  $(E, \mathcal{E})$  is inner regular, we say that the *space* is inner regular.

# A simple stress release model

Stress accumulates in small amounts of expected size  $\epsilon$ , and at rate  $\lambda$ , and all the stress accumulated so far is released completely (a seismic event—say a mine collapse) at points of a Poisson process with rate  $\sigma$ .

Let  $E = \mathbb{R}_+$  and  $\mathcal{E} = \mathcal{B}(\mathbb{R}_+)$ . Define “rates”

$$q(x, A) = \sigma I_A(0) + \int_A g(x, y) dy \quad (A \in \mathcal{E}, x \in E),$$

where

$$g(x, y) = \frac{\lambda}{\epsilon^2} \exp(-(y-x)/\epsilon) I_{(x, \infty)}(y).$$

# A simple stress release model

This is a version of a stochastic slip-predictable model for earthquake occurrences\*. The state of the process represents the accumulated stress on a fault. The process waits a time which is exponentially distributed with mean  $1/q(x) = 1/(\sigma + \lambda/\epsilon)$  and then either jumps to 0, which is identified as a seismic event (stress release), with probability  $\alpha = \sigma/(\sigma + \lambda/\epsilon)$  or otherwise jumps up a distance which is exponentially distributed with mean  $\epsilon$ . For small  $\epsilon$ , the later is a pure-jump analogue of a continuous constant stress increase.

\*Kiremidjian, A.S. and Anagnos, T. (1984) Stochastic slip-predictable model for earthquake occurrences, Bull. Seism. Soc. Amer. 74, 739–755.

# A simple stress release model

Define  $m = (m(x), x \in E)$  by  $m(\{0\}) = \alpha$  and, for  $x > 0$ ,  
 $m(dx) = \epsilon^{-1} \alpha(1 - \alpha) \exp(-\alpha x/\epsilon) dx$ , so that  $m((0, \infty)) = 1 - \alpha$ ,  
 $m(E) = m([0, \infty)) = 1$ , and,

$$m((0, x)) = (1 - \alpha)(1 - e^{-\alpha x/\epsilon}) \quad (x > 0)$$

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This shall remain one of life's mysteries, at least until I have the opportunity to speak again on this topic.