# L.C.G. ROGERS Coupling and the tail $\sigma$ -field of a onedimensional diffusion

## **1. INTRODUCTION**

The question of the triviality of the tail  $\sigma$ -field of a one-dimensional diffusion has been completely decided by work of Rösler, Fristedt and Orey, and Küchler and Lunze; the purpose of this note is essentially pedagogical, in that we shall show how the ideas of these earlier papers can be drawn together, illuminated and simplified using a little stochastic calculus.

We shall work throughout with a regular one-dimensional diffusion X with the interval  $I \subset \mathbb{R}$  as statespace. Our diffusion will always be assumed to be in natural scale, with speed measure m; we refer the reader to Breiman [1], Freedman [2], Mandl [7], Rogers and Williams [8] for definitions and properties. The sample space is the canonical space  $\Omega = C(\mathbb{R}^+, I)$  and we define for  $t \ge 0$ 

 $G_t \equiv \sigma(\{X_s : s \ge t\}) ,$ 

where X is the canonical process. The *tail*  $\sigma$ -*field* is defined to be

$$\mathcal{G} \equiv \bigcap_{t \ge 0} \mathcal{G}_t \; .$$

Informally, an event is in G if it is determined by the *ultimate* behaviour of the path. In view of the Kolmogorov 0-1 law, it seems reasonable (and is true, as we shall see) that if X is recurrent, the tail  $\sigma$ -field is trivial. If X is transient, is it possible that the tail  $\sigma$ -field is non-trivial, and if so, how? The most illuminating explanation of the fact that G can be non-trivial appears in the paper of Fristedt and Orey [3]. Consider the stochastic differential equation

$$Y_t = \int_0^t \sigma(Y_s) dB_s + t \tag{1}$$

where  $\sigma \in C_{\kappa}^{\infty}$ . Then the solution Y will diffuse for a while, and will ultimately leave the support of  $\sigma$  for ever, and follow the trajectory  $Y_t = \eta + t$ , where the random variable  $\eta$  is tail measurable. One expects that if  $\sigma$  were everywhere positive but tended to zero sufficiently rapidly at infinity, then the qualitative behaviour of Y would not change much, and G would be non-trivial. This is exactly what happens; and, as Fristedt and Orey prove, this is all that happens.

Let us now explain the principal results. It turns out that the only interesting case is (reducible to) the case where I = (0,1], and 1 is not absorbing. In this case, the diffusion tends to zero as  $t \to \infty$ . Defining for  $x \in I$ 

$$c(x) \equiv \operatorname{I\!E}^1(H_x)$$

(where  $H_x \equiv \inf\{t > 0 : X_t = x\}$ , and  $\mathbb{E}^y$  denotes expectation with respect to  $\mathbb{P}^y$ , the law of the diffusion started at y), the function c is finite-valued, and

$$M_t \equiv t - c(X_t)$$

is a local martingale. Here is the main result.

THEOREM 3. The following are equivalent:

- (i) *G* is not trivial;
- (ii)  $E^{x} < M >_{\infty} < \infty$  for some (all)  $x \in (0,1]$ ;
- (iii) for some (all)  $x \in (0,1]$ ,

$$\lim_{y \to 0} \operatorname{var}_x(H_y) < \infty;$$

(iv) 
$$\int_{0}^{1} y(m[y, 1])^{2} dy < \infty$$

- (v)  $(M_t)_{t \ge 0}$  is bounded in  $L^2(\mathbb{P}^x)$  for some (all)  $x \in (0,1]$ ;
- (vi)  $(M_t)_{t \ge 0}$  is convergent a.s.  $\mathbb{P}^x$  for some (all)  $x \in (0,1]$ .

**Remarks.** Rösler [1] proved (i)  $\Leftrightarrow$  (iii). Fristedt and Orey added (iv) and (vi). By var<sub>x</sub> we mean the variance under  $\mathbb{P}^x$ . The third condition is illuminating;  $\mathbb{E}^x(H_y) \to \infty$  as  $y \downarrow 0$ , yet the variances of  $H_y$  remain bounded. Compare this with (1).

The key to the proof of Theorem 3 is the following. If X and X' are independent diffusions in I, the first with law  $\mathbb{P}^{x}$ , the second with law  $\mathbb{P}^{x'}$ ;  $x \neq x'$ , and if  $T = \inf\{t > 0 : X_t = X'_t\}$ , then

 $G \text{ is trivial } \Leftrightarrow T < \infty \quad \text{a.s.} \tag{2}$ 

This coupling characterisation of the triviality of G is due to Küchler and Lunze [5]; we prove it below.

In \$2, we set out briefly some notation and basic ideas, and go on to prove (2). Then in \$3 we prove Theorem 3. The final section, \$4, deals with the remarkable result of Fristedt and Orey that  $G = \sigma(M_{\infty})$  in the case where G is non-trivial. We shall make use in what follows of some standard facts about one-dimensional diffusions:

- (3) the semigroup  $\{P_t : t \ge 0\}$  is strong Feller;
- (4) there is a strictly positive continuous p : (0,∞) × int(I) × int(I) → ℝ such that for f supported in int(I)

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$$P_t f(x) = \int p_t(x, y) f(y) m(dy)$$

(see Itô-McKean [4], §4.11);

(5) the  $\mathbb{P}^x$ -distribution of  $H_y$  has a uniformly continuous density (which is even unimodal - see Rösler [10]).

### 2. COUPLING AND TRIVIALITY OF THE TAIL $\sigma$ -field G

Any bounded function  $f : \mathbb{R}^+ \times \mathbb{I} \to \mathbb{R}$  which is non-constant and such that for all  $t, s \ge 0$ 

$$f(t,\cdot) = P_s f(t+s,\cdot) \tag{6}$$

will be called a *tail function*. Here,  $(P_t)$  is the transition semigroup of the diffusion X; since it is strong Feller, it can easily be deduced that any tail function must be jointly continuous. The terminology is explained by the following result.

**Proposition**. *The following are equivalent:* 

- (i) There exists a tail function;
- (ii) G is non-trivial under  $\mathbb{P}^x$  for each  $x \in int(I)$ ;
- (iii) G is non-trivial under  $\mathbb{P}^x$  for some  $x \in I$ .

There is a 1-1 correspondence between tail functions f and bounded tail-measurable random variables Y given by

$$f(t,X_t) = \mathbb{E}^{\mathcal{Y}}(\mathcal{Y} \mid \mathcal{F}_t), \quad \mathcal{Y} = \lim_{t \to \infty} f(t,X_t) .$$
(7)

*Proof.* (i) => (ii). The process  $Y_t \equiv f(t, X_t)$  is, from (6), a bounded martingale, so that  $\mathbb{P}^x$ -a.s. the limit Y exists. If Y were constant, so would  $f(t, X_t)$  be, which contradicts the existence of a strictly positive transition density (4).

(iii) => (i). Let  $Y \in L^{\infty}(G)$  be non-constant. Define f by  $f(t,X_t) = \mathbb{E}^{x}(Y | \mathcal{F}_t)$ ; the right-hand side is a function only of  $X_t$  since  $Y \in L^{\infty}(G) \subset L^{\infty}(G_t)$ . Since Y is non-constant, and  $Y = \lim_{t \to \infty} f(t,X_t)$ , f cannot be constant. The fact that (7) holds is immediate from the Markov property.

Now let  $X_0 = x$ , and let X' be an independent copy of X but with starting point x'. Define  $G'_t \equiv \sigma(\{X'_s : s \ge t\})$  and let  $\mathcal{A}_t \equiv G_t \lor G'_t$ . We define  $\mathcal{A} \equiv \bigcap_t \mathcal{A}_t$ . Because X and X' are independent, we have that

$$\mathcal{A} = \mathcal{G} \lor \mathcal{G} ;$$

with Lemma 2 of Lindvall and Rogers [6]. Without independence, this result is false in general. Let  $\tilde{\mathbb{P}}^{(x,x')}$  denote the law of the bivariate process (X,X').

Here is the key result of Küchler and Lunze.

THEOREM 1. The following are equivalent:

- (i) for all  $x, x' \in I$ ,  $\tilde{\mathbb{P}}^{(x,x')}(T < \infty) = 1$ ;
- (ii) for all  $x \in I$ , G trivial under  $\mathbb{P}^x$ .

*Proof.* (i) => (ii). Recall the definition of the coupling time T given in the first section. We use the fundamental coupling inequality for the total-variation norm of  $P_t(x, \cdot) - P_t(x', \cdot)$ ,

$$||P_t(x,\cdot) - P_t(x',\cdot)|| \le 2\tilde{\mathbb{P}}^{(x,x')}(T > t);$$

see, for example, Rogers and Williams [8], V.54.2. Suppose then that f is some tail function. Then for any  $s, t \ge 0$ ,  $x, x' \in I$ 

$$|f(s,x) - f(s,x')| \leq ||P_t(x,\cdot) - P_t(x',\cdot)|| \cdot ||f(s+t,\cdot)||_{\infty}$$
  
$$\to 0 \quad \text{as} \quad t \to \infty$$

by hypothesis. Hence immediately f is constant, and so G is  $\mathbb{P}^x$ -trivial for all x.

(ii) => (i). Since  $\mathcal{A} = \mathcal{G} \vee \mathcal{G}$ , it follows that  $\mathcal{A}$  is  $\tilde{\mathbb{P}}^{(x,x')}$ -trivial for all x, x'. Now define

 $A^{+} \equiv \{X_{t} - X'_{t} > 0 \text{ for all large enough } t\},$  $A^{-} \equiv \{X_{t} - X'_{t} < 0 \text{ for all large enough } t\},$ 

$$\phi^{\pm}(x,x') \equiv \widetilde{\mathbb{P}}^{(x,x')}(A^{\pm}) .$$

The events  $A^+$  and  $A^-$  are tail events; indeed, they are even invariant events. By hypothesis then,  $\phi^{\pm}$  take only the values 0 and 1. We shall show that  $\phi^{\pm}$  are both zero throughout  $\operatorname{int}(I) \times \operatorname{int}(I)$ , leaving the extension to the boundary as an easy exercise. Suppose that  $x_0, x'_0 \in \operatorname{int}(I)$  and that  $\phi^+(x_0, x'_0) = 1$ . Then by (4) it must be that  $\phi^+$  is equal to 1,  $m \times m$ -a.e. in  $\operatorname{int}(I) \times \operatorname{int}(I)$ ; and, since  $\phi^+$  is invariant, it follows immediately that  $\phi^+$  is 1 everywhere in  $\operatorname{int}(I) \times \operatorname{int}(I)$ . But  $\phi^+(x, x') = \phi^-(x', x)$ , and so  $\phi^-$  is 1 everywhere in  $\operatorname{int}(I) \times \operatorname{int}(I)$  – which is a contradiction because the events  $A^+$  and  $A^$ are disjoint. Thus  $\mathbb{IP}^{(x,x')}(A^{\pm}) = 0$  for all x, x', and the two independent diffusions X and X' keep on crossing over, so, in particular, the coupling time T must be finite.  $\diamond$ 

**Terminology.** We say that coupling is certain if for all  $x, x' \in I$ ,  $\tilde{\mathbb{P}}^{(x,x')}(T < \infty) = 1$ , and we say that G is trivial if G is  $\mathbb{P}^x$ -trivial for all x.

Here is a simple consequence of Theorem 1:

### If X is recurrent, then G is trivial.

*Proof.* (i) If  $I = \mathbb{R}$ , then X and X' are independent continuous local martingales, so  $\langle X - X' \rangle = \langle X \rangle + \langle X' \rangle$ . Moreover,  $\langle X \rangle_{\infty} = +\infty$  a.s. since X does not converge. Thus  $\langle X - X' \rangle_{\infty} = +\infty$  a.s., and (since X - X' is a time change of Brownian motion)  $\sup(X_t - X'_t) = \sup(X'_t - X_t) = +\infty$ . Thus coupling is certain.

(ii) If  $I = [0, \infty)$ , say, with 0 reflecting, extend the speed measure *m* into  $(-\infty, 0)$  by reflection and make up independent diffusions Y, Y' on  $\mathbb{R}$  with this speed; by (i), Y and Y' are certain to couple and, since |Y| has the same law as X, X and X' are also certain to couple.

 $\Diamond$ 

(iii) The case of compact I is similar to (ii).

## 3. CHARACTERISING THE CASES WHERE G IS NON-TRIVIAL

We now consider what happens when the diffusion X is transient. One of the following cases must apply (after shifting and rescaling I if necessary).

- (i) I = [0,1], 0 and 1 both absorbing. Then G is not trivial.
- (ii) I = [0,1], 0 absorbing, 1 not absorbing. In this case, if f is a tail function,  $f(s, 0) = f(t, 0) \forall s, t \ge 0$  and so, since the diffusion is certain to be absorbed in 0, f is constant. Hence G is trivial.

(iii) I = (0,1], 1 is absorbing. Again, G is non-trivial.

(iv) I = (0,1], 1 is not absorbing. This is the interesting case.

(v)  $I = [0, \infty), 0$  is absorbing. As in (ii), G is trivial.

(vi)  $I = (0, \infty)$ . We shall see later that this can be reduced to case (iv).

Thus we shall until further notice assume that

I = (0,1] and 1 is not absorbing.(8)

In particular, this implies that for  $0 < y \le x \le 1$ , all moments of  $H_y$  are finite under  $\mathbb{P}^x$  (see, for example, Rogers and Williams [8], V.46.1) and, with  $c(x) = \mathbb{E}^1(H_x)$  as before,

 $M_t = t - c(X_t)$  is a continuous local martingale under each  $\mathbb{P}^x$ .

Indeed, for y < x,

$$M(t \wedge H_{y}) = \mathbb{E}^{x}[H_{y} \mid \mathcal{F}_{t}] - c(y) .$$
<sup>(9)</sup>

It is well known that c is strictly decreasing and convex.

We are now in a position to prove the main result.

THEOREM 3. The following are equivalent:

# (i) G is not trivial; (ii) $\mathbb{E}^{x} < M >_{\infty} < \infty$ for some (all) $x \in (0,1]$ ; (iii) for some (all) $x \in (0,1]$ ; $\lim_{y \to 0} \operatorname{var}_{x}(H_{y}) < \infty$ ; (iv) $\int_{0}^{1} y(m[y,1])^{2} dy < \infty$ ; (v) $(M_{t})_{t \geq 0}$ is bounded in $L^{2}(\mathbb{P}^{x})$ for some (all) $x \in (0,1]$ ;

(vi)  $(M_t)_{t\geq 0}$  is  $\mathbb{P}^x$ -a.s. convergent for some (all)  $x \in (0,1]$ .

Proof. (ii) 
$$\Leftrightarrow$$
 (iii). From (9), for  $0 < y \le x \le 1$   
 $\mathbb{E}^x < M >_{H_y} = \mathbb{E}^x (M(H_y) - M(0))^2$   
 $= \mathbb{E}^x (H_y - \mathbb{E}^x (H_y))^2$   
 $= \operatorname{var}_x (H_y)$ .

Moreover, by the strong Markov property at  $H_x$ ,

$$\operatorname{var}_1(H_{\gamma}) = \operatorname{var}_1(H_{\chi}) + \operatorname{var}_{\chi}(H_{\gamma}) ,$$

and  $\operatorname{var}_1(H_x) < \infty$  since all moments of  $H_x$  are finite under  $\mathbb{P}^1$ . (ii)  $\Leftrightarrow$  (v) is immediate, as is (v) => (vi).

To prove the other equivalences, we shall assume we have the diffusion X represented as a time change of a Brownian motion, as we may. Thus if B is a Brownian motion started at  $x \in I$ , with local time process  $\{l(t,x) : t \ge 0, x \in \mathbb{R}\}$ ,

$$A_t \equiv \int_{I} m(da)l(t,a) ,$$
  

$$\sigma_t \equiv \inf\{u : A_u > t\} ,$$
  

$$X_t \equiv B(\sigma_t) .$$

Let  $\tau_y \equiv \inf\{t : B_t = y\}, \zeta \equiv \tau_0$ . Then  $H_y = A(\tau_y)$  for y < x. Hence, defining c(x) = 0 for  $x \ge 1$ ,

$$M(A_{t \wedge \tau_y}) = A_{t \wedge \tau_y} - c(B_{t \wedge \tau_y})$$
$$= \mathbb{E}^x[A_{\tau_y} | \mathcal{B}_t] - c(y)$$

is a  $(\mathcal{B}_t)$ -martingale (where  $(\mathcal{B}_t)$  is the filtration of B) and hence  $M \circ A$  is a  $(\mathcal{B}_t)$ -local martingale on  $[0, \zeta)$ . Since c is convex, we may apply the generalised Itô formula (see Rogers and Williams [8], IV.45.1) to deduce that for  $t < \zeta$ 

$$M \circ A_t = \int_0^t c'(B_s) dB_s$$

and that  $\frac{1}{2}c'' = dm$  as measures. This tells us that for each  $x \in I$ , c'(x) = -2m([x, 1]), and that

$$_t = \int_0^t c'(B_s)^2 ds = _{A_t}$$

Now  $\langle M \circ A \rangle_t$  is an additive functional of *B*, and the criterion for this to converge as  $t \uparrow \zeta$  is well known from the study of boundary behaviour of diffusions; we have

and if this condition fails,  $\langle M \rangle_{\infty} = +\infty$ ,  $\mathbb{P}^{x}$ -a.s.. See, for example, Rogers and 84

Williams [8], V.51.2. Equivalence of (ii) and (iv) is now immediate, and, since  $M_t = W(\langle M \rangle_t)$  for some Brownian motion W, (vi) implies that  $\mathbb{P}^x$ -a.s.  $\langle M \rangle_{\infty} < \infty$  for some x so that, from (10),  $\mathbb{E}^x \langle M \rangle_{\infty} < \infty$ , which is (ii).

All that now remains is to prove that (i) is equivalent to all the other statements. If (vi) holds, then  $M_{\infty}$  is a non-degenerate tail-measurable random variable (*M* cannot be constant); thus *G* is not trivial. Conversely, if (iv) is false,  $\langle M \rangle_{\infty} = +\infty$ ,  $\mathbb{P}^{x}$ -a.s. for each *x*, and so if *X*, *X'* are independent copies of the diffusion with different starting points then  $\langle M - M' \rangle_{\infty} = +\infty$  a.s., where  $M'_{t} \equiv t - c (X'_{t})$ . Hence, as before, for some *t* 

$$c(X_t) - c(X'_t) = 0 ,$$

and so X and X' couple, since c is strictly decreasing.

Finally, to dispose of the case where  $I = (0, \infty)$ , notice that if G is non-trivial there is some tail function, which is a tail function for the diffusion obtained by time-changing out the time spent in  $(1, \infty)$ . Hence for this new diffusion we can apply Theorem 3 and conclude that

$$\int_{0}^{1} y^{2} (m[y,1])^{2} dy < \infty .$$
<sup>(11)</sup>

Conversely, if this integral is finite,  $N_{\infty}$  is a non-trivial *G*-measurable random variable, where N is the martingale

$$N_t = A_t - \bar{c}(X_t) ,$$

with  $A_t \equiv \int_0^t I_{(0,1]}(X_s) ds$ ,  $\bar{c}(x) \equiv \mathbb{E}^1[A(H_x)]$ . In summary then, when  $I = (0, \infty)$ , non-triviality of *G* is equivalent to (11).

#### 4. THE STRUCTURE OF G IN THE CASES OF NON-TRIVIALITY

We shall suppose still that we are dealing with the interesting case where I = (0,1] and 1 is reflecting; all others can be reduced to this.

If G is non-trivial, and  $\Lambda \in G$ ,  $\mathbb{P}^1(\Lambda) \in (0,1)$ , we let f denote the corresponding tail function, and note that for each x

$$\Lambda = \{ f(t, X_t) \to 1 \} = \{ f(H_{1/n}, n^{-1}) \to 1 \} \quad \mathbb{P}^x - a.s..$$

Thus if  $Y_k \equiv H_{1/k} - c(1/k)$ , we have that

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$$\Lambda \in \bigcap_{n} \mathcal{A}_{n} \equiv \bigcap_{n} \sigma(\{Y_{k} : k \ge n\}) .$$

Now since  $Z_k \equiv Y_k - Y_{k-1}$  defines a sequence of independent random variables, and since  $Y_k \to Y \equiv \lim_{t \to \infty} M_t \equiv M_{\infty}$ , we have that

$$\Lambda \in \bigcap_{n} \mathcal{A}_{n} = \bigcap_{n} \sigma(\{Y, Z_{k} : k > n\}) \equiv \mathcal{A} .$$

In view of Kolmogorov's 0-1 law, we expect that  $\mathcal{A} = \sigma(Y)$ ; as an example below will show, this is not correct. However, a special feature of the current situation saves us.

THEOREM 4. Let  $\{Z_k; k \ge 1\}$  be independent random variables whose partial sums  $Y_n$  converge a.s. to Y. Suppose that for some k,  $Z_k$  has a uniformly continuous density. Then the tail  $\sigma$ -field of  $\{Y_n\}$  is  $\sigma(Y)$ .

*Proof.* (i) Without loss of generality, suppose that  $Z_1$  has a uniformly continuous density f. Notice that for any probability distribution F, the convolution of F with the law of  $Z_1$  has the density  $(F * f)(x) = \int F(dt)f(x - t)$ , which is again uniformly continuous with modulus of continuity no larger than that of f. Thus if  $\phi_k$  is the density of  $Y_k$ , and  $\phi$  is the density of Y, we have for all a < b

$$\int_a^b \phi_k(x) dx \quad \to \quad \int_a^b \phi(x) dx \quad ,$$

and so by the equi-uniform continuity of  $\{\phi, \phi_k : k \ge 1\}$ , it follows that  $\phi_k(a) \to \phi(a)$  for every *a*, and hence that  $\phi_k(y_k) \to \phi(y)$  if  $y_k \to y$ .

(ii) It suffices to prove that for bounded  $X \in \mathcal{F}_n = \sigma(\{Y_k : k \le n\})$  and bounded  $\mathcal{A}$ -measurable V,

$$E(XV) = E(E(X|Y)V) .$$
<sup>(12)</sup>

Now  $E(X | \mathcal{A}_n) = E(X | Y_n) = g(Y_n)$  for some bounded Borel function g, so for  $k \ge n$ 

$$E(XV) = E(E(X | \mathcal{A}_k)V)$$
  
=  $E[E(E(X | \mathcal{A}_n) | \mathcal{A}_k)V]$   
=  $E[E(g(Y_n) | \mathcal{A}_k)V]$   
=  $E[E(g(Y_n) | Y_k)V]$ .

If  $F_{n,k}$  is the distribution function of  $Y_k - Y_n \equiv Z_{n+1} + \cdots + Z_k$ , then

$$E[g(Y_n)|Y_k] = \int F_{n,k}(dt) g(Y_k - t)/\phi_k(Y_k) .$$
 (13)

Now take bounded uniformly continuous  $\tilde{g}$  such that  $E |(g - \tilde{g})(Y_n)| < \varepsilon$ , and notice that  $E[\tilde{g}(Y_n)|Y_k] \rightarrow E[\tilde{g}(Y_n)|Y]$  a.s. as  $k \rightarrow \infty$ ; indeed, the denominator is a.s. convergent since  $Y_k \rightarrow Y$ , and the convergence of the numerator is ensured by the uniform continuity of  $\tilde{g}\phi_n$ . The equality (12) follows.

Because the first passage time  $H_{1/k}$  has a uniformly continuous density under each  $\mathbb{P}^x$ , x > 1/k, we can apply Theorem 4 to obtain the pleasing conclusion  $\Lambda \in \sigma(M_{\infty})$ ; the *only* non-trivial information in the tail  $\sigma$ -field G comes from the limit of  $t - c(X_t)$ .

Finally, we provide an example which shows that Theorem 4 fails without the assumption of a uniformly continuous density for some  $Z_k$ . Let  $p_1 < p_2 < \cdots$  be the primes in ascending order, and suppose that

$$P(Z_k = \frac{1}{p_k}) = P(Z_k = -\frac{1}{p_k}) = \frac{1}{2}$$

Since  $p_k$  is of order k log k, the martingale  $Y_k = Z_1 + \cdots + Z_k$  is almost surely and  $L^2$  convergent. Now

$$Y_n + \sum_{1}^{n} (p_k)^{-1} = 2 \sum_{1}^{n} I_{\{Z_k > 0\}} (p_k)^{-1}$$

and so from  $Y_n$  we can deduce all of  $Z_1, ..., Z_n$ , because if we take primes  $q_1, ..., q_s$ and combine  $\sum_{j=1}^{r} (q_j)^{-1}$  over a common denomnator, then that denominator is  $\prod q_j$ . However, knowing Y does not allow us to deduce the sign of  $Z_1$ , for example.

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