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Kinetics of Contact Processes under Segregation

Tomás Aquino^{1, 2, *} and Marco Dentz¹

¹Spanish National Research Council (IDAEA–CSIC), 08034 Barcelona, Spain ²Geosciences Rennes, UMR 6118, CNRS, Université de Rennes 1, Rennes, France

Abstract

The kinetics of contact processes are determined by the interplay among local mass transfer mechanisms, spatial heterogeneity, and segregation. Determining the macroscopic behavior of a wide variety of phenomena across the disciplines requires linking reaction times to the statistical properties of spatially fluctuating quantities. We formulate the dynamics of advected agents interacting with segregated immobile components in terms of a chemical continuous time random walk. The inter-reaction times result from the first-passage times of mobile species to and across reactive regions, and available immobile reactants undergo a restart procedure. Segregation leads to memory effects and enhances the role of concentration fluctuations in the large-scale dynamics.

^{*} e-mail: tomas.decamposaquino@univ-rennes1.fr

6 I. INTRODUCTION

⁷ Mass-action reactions find applicability as population dynamics models of contact pro-⁸ cesses between agents, spanning biological processes [1], epidemiology [2], ecology [3], quan-⁹ tum molecular dynamics [4], and chemical reactions in geological media [5]. The large-scale ¹⁰ dynamics of such processes are determined by the interplay between local mechanisms, ¹¹ spatial heterogeneity, and segregation. Transport limitations under segregation and spa-¹² tial heterogeneity lead to inter-reaction times which are related to the first encounter time ¹³ between reactants [6–10]. Broad reaction times result in mesoscopic dynamical coupling ¹⁴ between transport and reaction due to memory effects [11, 12]. In turn, the large-scale ¹⁵ behavior may involve non-classical, fluctuation-driven kinetics [13–15], broad reaction rate ¹⁶ distributions [16], time-dependent reaction rates [17], and time-nonlocal kinetics [12]. The ¹⁷ concept of residence time and its role in reactive transport have received much attention, ¹⁸ in particular in hydrogeological applications [18, 19]. This has led to the formulation of ¹⁹ large-scale dynamics in terms of so-called stochastic-convective streamtube models, where ²⁰ reactive transport in the presence of physical and chemical heterogeneity is represented in ²¹ terms of an ensemble of streamtubes [20–23].

The classical picture for stochastic reactions, corresponding to the Gillespie algorithm [24], assumes complete reactant mixing. It leads to exponential inter-reaction times and predicts the classical mass-action rate laws for large reactant numbers [25–27]. The recently proposed chemical continuous time random walk (chCTRW) framework allows for more general interreaction times, leading to a broader class of large-scale rate laws including time-nonlocal kinetics [28]. However, linking disorder properties, mass transfer, and inter-reaction times, a fundamental step towards the understanding and quantification of the emergence of largescale kinetics in the presence of spatial heterogeneity and reactant segregation, remains in general an open problem.

This work develops this link for advective transport under spatial chemical disorder. We consider segregated immobile reactants, which react with mobile components. Inter-reaction times relate directly to first-passage times (FPTs) of mobile reactants to and across reactive regions. Upon encountering a reactive region, mobile components come into contact with the locally-available reactants. As we will see, this corresponds to a restart of the immobile reactants according to the original resident copy numbers. Processes under restart have



Figure 1. Illustration of the model segregation structure. The velocity V of mobile species (proportional to black bar length) is constant along each separate trajectory and distributed across trajectories. Mass evolves according to reaction in reactive regions (blue), interspersed with nonreactive regions (yellow), both of distributed length.

³⁷ received considerable attention in the context of chemical reactions and in particular as a ³⁸ framework for optimizing search strategies [29–32].

We derive a generalized master equation [33] for the chemical kinetics using the chCTRW framework, which we generalize to account for restart and parameterize in terms of the heterogeneity. The corresponding large-scale kinetics exhibit memory effects, and fluctuations about the average concentration play a role at large scales.

43 II. MODEL

We consider *mobile* reactants advected along separate trajectories. Advective velocity V45 is constant in each trajectory, and independent and identically distributed (i.i.d.) accord-46 ing to a probability density function (PDF) $\xi(\cdot)$ across trajectories. Additional reactants 47 are confined to certain regions, see Fig. 1. We term these reactants *immobile* and these 48 regions *reactive*. The latter are treated as well-mixed batch reactors, such that the chemical 49 dynamics within proceed according to the classical Gillespie algorithm [24].

The length L_r of reactive regions is an i.i.d. random variable distributed according to ⁵¹ the PDF $\rho_r(\cdot)$. The well-mixed assumption is difficult to justify if ρ_r has infinite moments. ⁵² Instead, we assume L_r is characterized by a finite mean ℓ_r , and for simplicity we set $\rho_r(\ell) =$ ⁵³ $e^{-\ell/\ell_r}/\ell_r$. Segregation of immobile reactants is characterized by the PDF $\rho_c(\cdot)$ of non-⁵⁴ reactive region lengths L_c between reactive regions, which are also i.i.d. We distinguish the ⁵⁵ cases of *mild segregation*, corresponding to finite inter-reactive-region mean distance ℓ_c , and ⁵⁶ strong segregation, corresponding to infinite mean distance, $\rho_c(\ell) \approx (\ell/\ell_c)^{-1-\beta}/[\ell_c|\Gamma(-\beta)|]$ ⁵⁷ for $\ell \gg \ell_c$, where $\Gamma(\cdot)$ is the gamma function, ℓ_c is a characteristic length, and $0 < \beta < 1$ [34]. ⁵⁸ We assume in what follows that mobile reactants begin within a reactive region to simplify ⁵⁹ the exposition. The general case of start in an arbitrary region, as illustrated in Fig. 1, can ⁶⁰ readily be treated in the framework developed below by including an additional waiting time ⁶¹ until the first reaction, corresponding to the time to reach the first reactive region.

The idealized model set forth here retains the key features of segregation and will allow us 62 ⁶³ to develop a quantitative model of its impact on chemical reactions. We consider in general $_{64}$ a set of m_s species S_j , $j = 1, \ldots, m_s$, which are subject to m_r reactions, $\sum_{j=1}^{m_s} r_{ij}S_j \rightarrow_{\kappa_i}$ $\sum_{j=1}^{m_s} p_{ij} S_j$, $i = 1, \ldots, m_r$. Each reaction is characterized by its stoichiometry, with r_{ij} (p_{ij}) ⁶⁶ denoting the number of reactants (products) of species j consumed (produced) by reaction 67 *i*. Furthermore, each reaction is associated with an intrinsic (microscopic) reaction rate 68 κ_i which fully determines its behavior under well-mixed conditions. All reactions involve ⁶⁹ immobile reactants, so that no reaction occurs outside reactive regions. The chemical state $_{70} N(t)$ describes the number of mobile and immobile reactants of each species at time t. The $_{71}$ net change in chemical state due to reaction *i* is described by the stoichiometry vectors 72 $s_i = p_i - r_i$. Sub- or superscripts M, I mark quantities relating to mobile and immobile ⁷³ components, respectively. For example, the chemical state is decomposed into mobile and ⁷⁴ immobile components as $N = (N_M, N_I)$ and the stoichiometry vectors as $s_i = (s_i^M, s_i^I)$. ⁷⁵ Throughout, we denote the Laplace transform by a tilde, the Laplace variable by λ , and ⁷⁶ ensemble averages (across trajectories) by angled brackets. A vertical bar is used to denote 77 conditioning.

We introduce also some key quantities governing the dynamics. The mean time spent row in a reactive region at velocity v is $\mu(v) = \ell_r/v$, so that $1/\mu(v)$ is the probability per unit time for mobile reactants to exit a given reactive region and experience the delay induced by the subsequent non-reactive region. Reactive patches are associated with a Damköhler number $Da(v) = \mu(v)a(\overline{n})/\overline{n}$, where \overline{n} are characteristic copy numbers of each species (e.g., the initial state) and \overline{n} is an overall characteristic copy number (e.g., the average of \overline{n} across components or a component of interest). This dimensionless number is the ratio of the characteristic transport and reaction times in a single reactive region. Finally, the ratio of characteristic non-reactive and reactive region lengths is denoted by $\alpha = \ell_c/\ell_r$.

87 III. QUALITATIVE DYNAMICS OF REACTION UNDER SEGREGATION

In order to motivate the general theory developed below, we first illustrate the im-⁸⁹ pact of segregation on the large-scale kinetics of the mobile–immobile degradation reaction ⁹⁰ $S_M + S_I \rightarrow_{\kappa} \emptyset$, where S_M is mobile and S_I is immobile. We consider both mild and ⁹¹ strong segregation, specifically with exponential and Lévy-stable [34] inter-reactive region ⁹² lengths. For simplicity, we set a fixed velocity V = v and a fixed number of initial immobile ⁹³ components $n_{0,I}$ in each reactive region.

We consider an instantaneous, point initial injection of mobile components, and we sim-⁹⁴ We consider an instantaneous, point initial injection of mobile components, and we sim-⁹⁵ ulate their average degradation under segregation for large particle numbers. When char-⁹⁶ acteristic reactant numbers $\overline{n} \to \infty$, we may define continuous (number) concentrations ⁹⁷ $C(t) = N(t)/\overline{n}$. We take \overline{n} as the average initial number of reactants. The reaction in reac-⁹⁸ tive patches then proceeds according to the well-mixed rate laws. For $n_{M,I}$ mobile/immobile ⁹⁹ particles under well-mixed conditions, there are $n_M n_I$ mobile–immobile particle pairs avail-¹⁰⁰ able for reaction. The reaction between each pair proceeds at the microscopic reaction rate ¹⁰¹ κ . The reaction between *some* pair thus proceeds at a rate $a(\mathbf{n}) = \kappa n_M n_I$. In terms of ¹⁰² concentrations, the well-mixed rate laws are thus given by

$$\frac{dc_{M,I}^{wm}(t)}{dt} = -\kappa^C c_M^{wm}(t) c_I^{wm}(t), \qquad (1)$$

¹⁰³ where $\kappa^C = \overline{n}\kappa$ is a macroscopic reaction rate associated with well-mixed concentration de-¹⁰⁴ cay. The analytical solution of this equation, along with details on the numerical simulations ¹⁰⁵ under segregation, is given in Appendix A.

The solid lines in Fig. 2 show the ensemble-averaged mobile concentrations $\bar{c}_M(t)$, nor-106 malized by the initial concentration $c_{0,M}$, as a function of time t for different values of 107 Damköhler number $Da = \kappa^{C} \mu c_{0,I}$, where $c_{0,I}$ is the initial immobile concentration in each 108 reactive region. The evolution of concentration presents qualitative differences in the func-109 tional form of the decay with varying Damköhler number. For *mild sequence* (left panel), 110 the decay at low Da is always exponential, but slower-than-exponential decay is present for 111 longer times as Da increases. Under strong segregation (right panel), we observe power-law 112 decay for all finite Da, characterized by the exponent β associated with the inter-reactive-113 ¹¹⁴ region lengths. The solutions for a single well-mixed patch, corresponding to the Da $\rightarrow \infty$ ¹¹⁵ limit, are also shown. In this case, the concentration initially decays faster than in the ¹¹⁶ segregated systems. However, due to depletion of the initial immobile concentration, which



Figure 2. Temporal evolution of concentration under the full degradation reaction $S_M + S_I \rightarrow_{\kappa} \varnothing$ (solid lines) and the catalytic degradation reaction $S_M + S_I \rightarrow_{\kappa} S_I$ (markers), for different values of Damköhler number Da. The initial conditions for mobile and immobile concentrations are $c_{0,M} = 4/3$ and $c_{0,I} = 2/3$. Fixed parameters are $\alpha = 2$, v = 4, and $\kappa^C = 2$, and Da is set by varying ℓ_r . Concentrations are averaged over 10⁵ realizations. Left: Mild segregation, with exponentiallydistributed non-reactive region lengths. Right: Strong segregation, with Lévy-stable-distributed non-reactive region lengths with exponent $\beta = 0.7$.

¹¹⁷ in this example is lower than the initial mobile concentration, reaction then slows down and ¹¹⁸ the mobile concentration approaches equilibrium.

The impact of segregation on reaction dynamics, illustrated here for this simple degra-119 ¹²⁰ dation reaction, is in general due to two factors. First, reaction is punctuated by times spent in non-reactive regions, which leads to an effective reaction slowdown. Second, mo-121 bile components react only with the locally-available immobile reactants in a given region. 122 When a reactive region is left and a subsequent one is entered, there is a restart of available 123 immobile reactants, according to their initially-available concentration. In order to highlight 124 the role of immobile reactant depletion and restart, the markers in Fig. 2 show the evolution 125 of mobile concentrations for the catalytic degradation reaction $\mathcal{S}_M + \mathcal{S}_I \rightarrow_{\kappa} \mathcal{S}_I$, using the 126 same parameter values as before. In this case, the immobile component acts as a catalyst, 127 and it is thus neither consumed nor subject to the effect of restart. For a single well-mixed 128 patch, in the Da $\rightarrow \infty$ limit, the mobile species' decay is then purely exponential and always 129 130 faster than under segregation. However, as $Da \rightarrow 0$, corresponding to negligible reaction ¹³¹ within each single region, the decay of mobile concentration becomes identical under the 132 two reactions.

133 IV. THEORY

In what follows, we will discuss and quantify the mechanisms discussed in the previous section. To this end, we develop a general framework for arbitrary reaction dynamics under spatial segregation.

137 A. Catalytic reactions

We start by examining the case where the immobile components serve as catalysts, i.e., they participate in the reactions but are not consumed or produced ($s_i^I = 0$). We further assume for the moment that the immobile reactants are present in identical copy numbers in the each reactive region. In this case, the dynamics proceed identically with the classical wellmixed scenario, except the periods of availability of reactants are punctuated by waiting times due to segregation. We proceed to find the chCTRW description of this system in terms of a generalized master equation, starting from the determination of the inter-reaction times.

146 1. Inter-reaction times

The lack of memory of the exponential distribution implies that the distribution of distas tances to the end of reactive regions, starting from any point within a reactive region, has the same length distribution as the full regions [27]. Thus, it is sufficient to study inter-reaction times starting at the beginning of reactive regions.

Consider a given chemical state N(t) = n and velocity V = v. For advective transport, 151 the FPT across reactive/conservative regions is given by the crossing time $L_{r,c}/v$. Thus, 152 the corresponding PDFs are given by $\psi_{r,c}(t|v) = v\rho_{r,c}(vt)$. Following chCTRW theory, we 153 write the inter-reaction times as $\tau(\mathbf{n}, v) = \tau_r(\mathbf{n}) + \tau_c(\tau_r, v)$. The intrinsic reaction time τ_r 154 represents reaction in the absence of segregation. It depends on the current chemical state 155 and corresponds to time spent in reactive regions, where reaction proceeds according to the 156 classical well-mixed theory. Let $\phi_i^r(t|\mathbf{n}) dt$ represent the joint probability that, in the absence 157 158 of segregation, reaction $1 \leq i \leq m_r$ fires next and after a waiting time in [t, t + dt]. Then, ¹⁵⁹ $\tau_r(\boldsymbol{n})$ has PDF $\phi^r(t|\boldsymbol{n}) = \sum_i \phi^r_i(t|\boldsymbol{n})$ (see Appendix B for a brief overview of the well-mixed $_{160}$ description). The additional global delay τ_c is caused by segregation. It corresponds to ¹⁶¹ time spent in non-reactive regions, which is fully determined by the current τ_r and does ¹⁶² not depend further on the chemical state. Specifically, a given time $\tau_r = t_r$ corresponds to ¹⁶³ fully traversing a certain number $\eta(t_r, v)$ of reactive regions, and $\tau_c(t_r, v) = \sum_{k=1}^{\eta(t_r, v)} L_{c,k}/v$, ¹⁶⁴ where the $L_{c,k}$ are i.i.d. according to ρ_c . For exponential reactive lengths, $\eta(t_r, v)$ is Poisson-¹⁶⁵ distributed, as shown in Appendix C. Thus, segregation leads to a compound-Poisson inter-¹⁶⁶ reaction delay. According to [28], the full inter-reaction time density is then related to the ¹⁶⁷ intrinsic inter-reaction time density by

$$\tilde{\phi}_i(\lambda|\boldsymbol{n}, v) = \tilde{\phi}_i^r[\lambda/\tilde{K}(\lambda|v)|\boldsymbol{n}, v],$$
(2)

168 where

$$\tilde{K}(\lambda|v) = \left[1 + \frac{1 - \tilde{\psi}_c(\lambda|v)}{\mu(v)\lambda}\right]^{-1}$$
(3)

¹⁶⁹ is the Laplace transform of a memory kernel, as we will see shortly.

170 2. Generalized master equation

We first define the probabilities $\gamma_i(\boldsymbol{n}|\boldsymbol{n}')$ of reaction *i* yielding the chemical state \boldsymbol{n} , given the starting state \boldsymbol{n}' . The change in state is given by the stoichiometry vector \boldsymbol{s}_i . Thus,

$$\gamma_i(\boldsymbol{n}|\boldsymbol{n}') = \delta_{\boldsymbol{n},\boldsymbol{n}'+\boldsymbol{s}_i},\tag{4}$$

¹⁷³ where $\delta_{,,}$ is the Kronecker delta. According to the general theory developed in [28], the ¹⁷⁴ generalized master equation for the ensemble-averaged propagator of the chemical state, ¹⁷⁵ given initial state n_0 and velocity v, is then given by

$$\frac{\partial p(\boldsymbol{n}, t | \boldsymbol{n}_0, v)}{\partial t} = \sum_{\boldsymbol{n}' \ge 0} \sum_{i=1}^{m_r} \int_0^t dt' \left[\gamma_i(\boldsymbol{n} | \boldsymbol{n}') - \delta_{\boldsymbol{n}, \boldsymbol{n}'} \right] \\ \times M_i(t - t' | \boldsymbol{n}', v) p(\boldsymbol{n}', t' | \boldsymbol{n}_0, v),$$
(5)

where the Laplace transforms of the memory functions are given in terms of the inter-reaction 177 time densities as

$$\tilde{M}_i(\lambda|\boldsymbol{n}, v) = \frac{\lambda \phi_i(\lambda|\boldsymbol{n}, v)}{1 - \sum_{i=1}^{m_r} \tilde{\phi}_i(\lambda|\boldsymbol{n}, v)}.$$
(6)

¹⁷⁸ Using Eq. (13) together with the well-mixed description, we obtain

$$\tilde{M}_i(\lambda|\boldsymbol{n}, v) = \tilde{K}(\lambda|v)a_i(\boldsymbol{n}), \tag{7}$$

with $a_i(\mathbf{n}) = \kappa_i h_i(\mathbf{n})$ and $a(\mathbf{n}) = \sum_{i=1}^{m_r} a_i(\mathbf{n})$, where the κ_i are microscopic rate constants and the $h_i(\mathbf{n})$ encode the dependency of the rates on the state. In the absence of segregation, $\psi_c(t) = \delta(t)$. Equation (3) with $\tilde{\psi}_c(\lambda) = 1$ leads to $\tilde{K}(\lambda|v) = 1$, and therefore $K(t|v) = \delta(t)$. This recovers the classical chemical master equation [25]. We thus see that K plays the role of a memory kernel describing reaction slowdown due to segregation. Reactions under master equation (5), are discussed in general terms in [28]. Note that for catalytic reactions, $\kappa_i = 0$ implies immobile copy numbers do not change, and it is sufficient to consider the new role of the mobile components.

As an example, consider the catalytic degradation reaction introduced in Section III. ¹⁸⁸ We have $h_1(\mathbf{n}) = n_M n_I$, the number of pairs of mobile–immobile particles, and $a(\mathbf{n}) =$ ¹⁹⁰ $a_1(\mathbf{n}) = \kappa n_M n_I$. The intrinsic waiting time density is given by $\phi_1^r(t|\mathbf{n}) = a(\mathbf{n}) \exp[a(\mathbf{n})t]$, ¹⁹¹ and the single memory function is $\tilde{M}_1(\lambda|\mathbf{n}, v) = \tilde{K}(\lambda|v)a(\mathbf{n})$. The stoichiometry vector is ¹⁹² $\mathbf{s}_1 = (-1, 0)$. Thus, the master equation for the mobile components becomes

$$\frac{\partial p(n_M, t)}{\partial t} = -\int_0^t dt' \, K(t - t') \left[a(n_M, n_{0,I}) p(n_M, t') - a(n_M + 1, n_{0,I}) p(n_M + 1, t') \right], \tag{8}$$

where we have omitted the dependency on the initial condition and v for notational brevity. 193 The ensemble-averaged probability of a given state at a given time is in general obtained 194 195 by averaging over velocities and initial conditions. Denote by $\gamma_M(\cdot|v)$ the initial distribution of mobile components at injection and by $\gamma_I(\cdot|v)$ the initial distribution of immobile reactant 196 numbers across reactive regions, given velocity v. The initial copy number distribution at 197 the first reactive region is thus $\gamma(\boldsymbol{n}|v) = \gamma_M(\boldsymbol{n}_M|v)\gamma_I(\boldsymbol{n}_I|v)$. For equal initial immobile 198 component copy numbers $n_{0,I}$ in each reactive patch, $\gamma_I(n|v) = \delta_{n,n_{0,I}}$. The probability 199 200 of finding the state N(t) = n at time t is $p(n,t) = \langle p[n,t|N_0(V),V] \rangle$, where for each $_{201} V = v$ the initial condition $N_0(v)$ is distributed according to $\gamma(\cdot|v)$. We note also that ²⁰² the propagator contains all necessary information to compute spatial quantities. Spatial ²⁰³ distributions may be obtained by multiplying the propagator by the probability that mobile $_{204}$ reactants are at position x at time t before averaging, which, assuming mobile species start 205 at x = 0 at t = 0, is given here by the Dirac delta $\delta(x - Vt)$. Similarly, concentrations fluxes ²⁰⁶ at a control plane at distance x as a function of time t are obtained by multiplying by the ²⁰⁷ FPT to distance x, given here by $\delta(t - x/V)$.

208 B. General reactions

We now study the general case where reactions may involve net production or consumption of immobile components, and where different reactive regions may initially comprise and different copy numbers of immobile reactants. In this case, as the mobile reactants reach each reactive region, they encounter the initial resident copy numbers. These are then depleted or produced according to reaction. Once the mobile components exit a given reactive region and arrive at the subsequent one, they again encounter resident copy numbers according to their initial distribution. This leads naturally to the concept of restart of immobile components. Developing a generalized master equation for these dynamics thus requires place account for restart.

218 1. Inter-reaction times under restart

In order to make use of the techniques developed in [28], we require that the dynamics be a Markov process in reaction step. For exponential reactive region lengths, whenever a reaction fires, the leftover reactive region length is identically distributed with the full region length, as discussed above. However, if the initial number of immobile reactants depends on the reactive region length, the number of immobile reactants then gives information about the region length, and the inter-reaction times are no longer independent of past history. We assume here that this is not the case.

Consider a given chemical state $\mathbf{N}(t) = \mathbf{n}$, velocity V = v, and lengths of a consecutive reactive/non-reactive region pair $L_{r,c}$. The effect of restart may be treated as a special reaction, which we number i = 0. It fires after a time L_r/v and leads to restart of the immobile components after a time $(L_r + L_c)/v$. The next reaction to fire is the one with the minimum inter-reaction time, including restart. The inter-reaction time densities, given the chemical state, velocity, and region lengths, are defined such that $\phi_i(t|\mathbf{n}, v, L_r, L_c) dt$ is the probability of reaction *i* firing and the inter-reaction time being in [t, t + dt]. Under 233 restart, we write

$$\phi_i(t|\boldsymbol{n}, v, L_r, L_c) = b_i(\boldsymbol{n}, v, L_r)\phi_{|i}(t|\boldsymbol{n}, v, L_r, L_c),$$
(9)

where $b_i(\boldsymbol{n}, v, L_r)$ is the propensity of reaction *i*, i.e., the probability that it fires next, and $\phi_{|i}(\cdot|\boldsymbol{n}, v, L_r, L_c)$ is the inter-reaction time PDF of reaction *i* given that it will fire next. Restart occurs if the minimum reaction time is larger than the restart time L_r/v . This happens with probability $\int_{L_r/v}^{\infty} dt \, \phi^r(t|\boldsymbol{n}) = \exp[-a(\boldsymbol{n})L_r/v]$. Otherwise, with probability $1 - \exp[-a(\boldsymbol{n})L_r/v]$, the normal reaction with the minimum inter-reaction time fires. Thus,

$$b_i(\boldsymbol{n}, v, L_r) = \begin{cases} e^{-a(\boldsymbol{n})L_r/v} \\ \left[1 - e^{-a(\boldsymbol{n})L_r/v}\right] \frac{a_i(\boldsymbol{n})}{a(\boldsymbol{n})} \end{cases}$$
(10)

²³⁹ Here and in what follows, the first case refers to i = 0 (restart), and the second to $1 \le i \le m_r$ ²⁴⁰ (regular reactions). The inter-reaction time PDFs must also be conditioned on $t < L_r/v$ for ²⁴¹ the regular reactions, and the waiting time associated with restart is $(L_r + L_c)/v$, so that

$$\phi_{|i}(t|\boldsymbol{n}, v, L_r, L_c) = \begin{cases} \delta\left(t - \frac{L_r + L_c}{v}\right) \\ \frac{a(\boldsymbol{n}) \exp[-a(\boldsymbol{n})t]}{1 - \exp[-a(\boldsymbol{n})L_r/v]} H\left(\frac{L_r}{v} - t\right) \end{cases},$$
(11)

²⁴² where H is the Heaviside step function. Thus, according to Eq. (9),

$$\phi_i(t|\boldsymbol{n}, v, L_r, L_c) = \begin{cases} e^{-a(\boldsymbol{n})L_r/v}\delta\left(t - \frac{L_r + L_c}{v}\right) \\ a_i(\boldsymbol{n})e^{-a(\boldsymbol{n})t}H\left(\frac{L_r}{v} - t\right) \end{cases}$$
(12)

²⁴³ Defining $\phi_i(t|\boldsymbol{n}, v) = \langle \phi_i(t|\boldsymbol{n}, v, L_r, L_c) \rangle$, we find the Laplace transforms

$$\tilde{\phi}_{i}(\lambda|\boldsymbol{n}, v) = \begin{cases} \tilde{\psi}_{r} \left[\lambda + a(\boldsymbol{n})|v\right] \tilde{\psi}_{c}\left(\lambda|v\right) \\ a_{i}(\boldsymbol{n}) \frac{1 - \tilde{\psi}_{r}[\lambda + a(\boldsymbol{n})|v]}{\lambda + a(\boldsymbol{n})} \end{cases}$$
(13)

The inter-reaction times are thus fully determined by first-passage properties together with the rates $a_i(\mathbf{n})$.

246 2. Generalized chemical master equation under restart

We turn to the generalized master equation incorporating the effect of restart. The transition probabilities $\gamma_i(n|n')$ corresponding to the regular reactions remain given by Eq. (4), as the effect of these reactions does not change. The effect of the restart reaction, conditioned 250 on a given velocity v, is characterized by:

$$\gamma_0(\boldsymbol{n}|\boldsymbol{n}',v) = \gamma_I(\boldsymbol{n}_I|v)\delta_{\boldsymbol{n}_M,\boldsymbol{n}'_M},\tag{14}$$

meaning that mobile reactants remain unaffected, and immobile copy numbers are redrawn the initial distribution as discussed above. As shown in detail in Appendix D, the generalized master equation corresponding to the dynamics under restart retains the same the form as Eq. (5), with the memory functions again given in terms of the inter-reaction time densities according to Eq. (6). However, sums over reactions in both these equations now the extend to i = 0, the reaction describing restart, and the inter-reaction time densities are given by Eq. (13). Direct computation shows that the modified inter-reaction times lead to the same memory functions for the regular reactions, as given by Eq. (7). Restart is associated with the memory function

$$M_0(\boldsymbol{n}|\boldsymbol{n}', v) = \tilde{K}(\lambda|v)\tilde{\psi}_c(\lambda)/\mu(v).$$
(15)

These results follow from the fact that normal reactions proceed at rate $a_i(\mathbf{n})$ in reactive regions, whereas restart occurs at rate $1/\mu(v)$ and is associated with an additional delay corresponding to traversing a non-reactive region. Realizations of these dynamics may be simulated with recourse to a generalized Gillespie algorithm under restart, which we outline the Appendix E.

²⁶⁵ Consider as an example the full degradation of Section III, with equal initial copy numbers ²⁶⁶ of immobile components $n_{0,I}$ in each reactive region. The stoichiometry vector is $\mathbf{s}_1 =$ ²⁶⁷ (-1, -1), and $\gamma_0 = \delta_{n_I, n_{0,I}} \delta_{n_M, n'_M}$. Similarly to Eq. (8) for the catalytic degradation example, ²⁶⁸ we find the master equation

$$\frac{\partial p(n_M, n_I, t)}{\partial t} = -\int_0^t dt' K(t - t') \left[a(n_M, n_I) p(n_M, n_I, t') - a(n_M + 1, n_I + 1) p(n_M + 1, n_I + 1, t') \right]$$

$$- \int_0^t dt' K_c(t - t') \left[a(n_M, n_I) p(n_M, n_I, t') - a(n_M, n_{0,I}) p(n_M, n_{0,I}, t') \right],$$
(16)

where the memory kernel associated with restart is given by the convolution $K_c(t) = \int_0^t dt' K(t-t')\rho_c(t')$, and we have again omitted dependencies on the initial condition and $v_{271} v_{c}$.

272 C. Restart and catalytic reactions

²⁷³ Consider equal initial immobile copy numbers in all reactive regions as in Section IV A. ²⁷⁴ Since for catalytic reactions $s_i^I = 0$, immobile copy numbers do not change due to either ²⁷⁵ restart or regular reactions, the i = 0 term in the master equation is null, and we recover ²⁷⁶ catalytic dynamics, Eq. (5).

The catalytic description also plays a role as the limiting behavior for slow reaction. For 278 small Damköhler number, $Da \ll 1$, the dynamics are transport-dominated at the scale of a 279 single region, meaning that many reactive regions must be visited before appreciable change 280 due to reaction can occur. For fixed initial immobile copy numbers, the reset mechanism 281 ensures there is no appreciable change in immobile copy numbers. The catalytic description 282 is then valid for arbitrary reactions, and the subordination formulation of [28] holds. Note 283 also that under these conditions, for arbitrary initial copy numbers, the dynamics are inde-284 pendent of the specific reactive region length distribution as long as it has a finite mean, see 285 Appendix C.

286 D. Large-scale kinetics

Next, we obtain the large-scale description corresponding to the mesoscopic master equations developed in the previous sections. In the large particle number limit, we work in terms of concentrations $\mathbf{C}(t) = \mathbf{N}(t)/\overline{n}$ as introduced in Section III. Correspondingly, we define $\mathcal{L}_{290} a_i^C(\mathbf{c}) = \kappa_i^C h_i^C(\mathbf{c}), a^C(\mathbf{c}) = \sum_{i=1}^{m_r} a_i^C(\mathbf{c}), \text{ with } \kappa_i^C = \overline{n} \sum_{j=1}^{m_s} r_{ij} - 1 \kappa_i / \prod_{j=1}^{m_s} r_{ij}!$ the macroscopic rate constants and $h_i^C(\mathbf{c}) = \prod_{j=1}^{m_s} c_j^{r_{ij}}$. Note that these quantities are fully defined in terms of Section III, we have $\kappa^C = \overline{n}\kappa$ and $h_1^C(\mathbf{n}) = c_M c_I$, from which $a^C(\mathbf{c}) = a_1^C(\mathbf{c}) = \kappa^C c_A c_B$, the usual rate in the well-mixed rate laws for concentration. The average initial numbers of immobile components for velocity v are denoted by $\overline{\mathbf{c}}_{0,I}(v)$ and the average concentrations for initial condition \mathbf{c}_0 and velocity v by $\overline{\mathbf{c}}(t|\mathbf{c}_0, v) = \langle \mathbf{C}(t)|\mathbf{c}_0, v \rangle$.

297 1. Dynamical equations

In Appendix F, we show that the ensemble-averaged concentrations obey the integrodifferential dynamical equations

$$\frac{d\overline{\mathbf{c}}_{M}(t|\mathbf{c}_{0},v)}{dt} = \int_{0}^{t} dt' K(t-t'|v) \sum_{i=1}^{m_{r}} \mathbf{s}_{i}^{M} \langle a_{i}^{C}[\mathbf{C}(t')]|\mathbf{c}_{0},v\rangle,$$

$$\frac{d\overline{\mathbf{c}}_{I}(t|\mathbf{c}_{0},v)}{dt} = \sum_{i=1}^{m_{r}} \mathbf{s}_{i}^{I} \langle a_{i}^{C}[\mathbf{C}(t)]|\mathbf{c}_{0},v\rangle$$

$$- \mu(v)^{-1} [\overline{\mathbf{c}}_{I}(t|\mathbf{c}_{0},v) - \overline{\mathbf{c}}_{0,I}(v)].$$
(17)

³⁰⁰ Segregation induces memory effects in the form of a convolution with a memory kernel for the ³⁰¹ mobile components. For the immobile components, restart leads to a mean-reverting forcing ³⁰² term. For *mild segregation*, the memory is short-term, and the late-time rate equations are ³⁰³ time-local, whereas memory is long-range under *strong segregation* and the late-time rate ³⁰⁴ equations involve non-local fractional derivatives [35].

In the classical rate laws, fluctuations vanish in the large-particle-number limit and 305 $\langle a_i^C[\mathbf{C}(t)]|\mathbf{c}_0,v\rangle = a_i^C[\mathbf{\bar{c}}(t|\mathbf{c}_0,v)]$ [27]. Here, this is not the case. Under strong segrega-306 tion, realizations of the dynamics break ergodicity weakly due to large fluctuations in the 307 inter-reaction times, leading to persistent fluctuations about the average concentrations [28]. 308 Under *mild segregation*, short-term memory effects coupled with the discontinuous changes 309 in immobile concentrations caused by restart also prevent the fluctuations from vanishing. 310 ³¹¹ While equations for higher moments of the concentration can be found by appropriate averaging of the generalized master equation (5), these depend on still higher moments, and 312 closed rate equations for the average components do not exist in general. In other words, fluc-313 tuations play an important role in the mean behavior, analogous to Ovchinnikov–Zeldovich 314 segregation in a bimolecular annihilation reaction among diffusing components [13, 36]. 315 A common approach is to employ moment closure approximations [37, 38], a technique 316 that must be adapted to specific reactions and which we do not explore here. We employ 317 stochastic algorithms, outlined in Appendix E, to numerically solve for the exact average 318 concentration. We note here that the mean-reverting term in the large-scale description 319 320 of immobile concentration depends only on average concentrations. Thus, the catalytic ³²¹ description (for catalytic reactions or the low-Damköhler limit of general reactions, see Sec-₃₂₂ tions IVA and IVC) holds at the large scale even if initial immobile copy numbers vary 323 across reactive regions.

324 2. Asymptotics

We now examine the asymptotic behavior of the large-scale kinetics. For mild segregais tion, under which the inter-reactive-region lengths have a mean, the Laplace transform of the corresponding FPT is approximated by $\tilde{\psi}_c(\lambda|v) \approx 1 - \alpha \mu(v)\lambda$ for $\lambda \ll 1/[\alpha \mu(v)]$, coris responding to large times compared to the mean time to traverse a non-reactive region. To leading order in $\lambda \ll 1/[(1 + \alpha)\mu(v)]$ (corresponding to large times compared to the mean time to traverse a reactive and a non-reactive region),

$$\tilde{M}_i(\lambda | \boldsymbol{n}, v) = \frac{1}{1 + \alpha} \begin{cases} 1/\mu(v) \\ a_i(\boldsymbol{n}) \end{cases}$$
(18)

³³¹ This leads to time-local equations for the mobile components at late times,

$$\frac{d\overline{\boldsymbol{c}}_M(t|\boldsymbol{c}_0, v)}{dt} = (1+\alpha)\mu(v)^2 \sum_{i=1}^{m_r} \boldsymbol{s}_i^M \langle \boldsymbol{a}_i^C[\boldsymbol{C}(t)] | \boldsymbol{c}_0, v \rangle.$$
(19)

For strong segregation, under which the inter-reactive region lengths do not have a mean, we have instead the small- λ expansion $\tilde{\psi}_c(\lambda|v) \approx 1 - [\alpha \mu(v)\lambda]^{\beta}$, $0 < \beta < 1$. To leading order in $\lambda \ll 1/[(1+\alpha)\mu(v)]$,

$$\tilde{M}_{i}(\lambda|\boldsymbol{n}, v) = \mu(v)\lambda \left[\alpha\mu(v)\lambda\right]^{-\beta} \begin{cases} 1/\mu(v) \\ a_{i}(\boldsymbol{n}) \end{cases},$$
(20)

335 yielding, at late times, the time-nonlocal equations

$$\frac{d^{\beta} \overline{\boldsymbol{c}}_{M}(t | \boldsymbol{c}_{0}, v)}{dt^{\beta}} = [\alpha \mu(v)]^{\beta} \mu(v) \sum_{i=1}^{m_{r}} \boldsymbol{s}_{i}^{M} \langle a_{i}^{C} [\boldsymbol{C}(t)] | \boldsymbol{c}_{0}, v \rangle, \qquad (21)$$

³³⁶ where d^{β}/dt^{β} denotes the Riemann–Liouville fractional derivative of order β [35].

337 V. LARGE-SCALE DYNAMICS OF REACTION UNDER SEGREGATION

In order to illustrate the main features of the theoretical developments in the context $_{339}$ of a particular reaction, let us return to the large-scale dynamics of mobile concentration $_{340}$ for the degradation reaction $\mathcal{S}_M + \mathcal{S}_I \rightarrow_{\kappa} \varnothing$ introduced in Section III. As the Damköhler



Figure 3. Temporal evolution of concentration for $S_M + S_I \rightarrow_{\kappa} \emptyset$ under *mild segregation* for different values of Damköhler number Da. Symbols represent stochastic simulations based on the Gillespie algorithm (10⁷ initial mobile particles and $5 \cdot 10^6$ immobile particles per reactive region, averaged over 10⁵ realizations) and solid lines employ the well-mixed rate equations in reactive regions (10⁶ realizations). Non-reactive region lengths are exponentially-distributed. Fixed parameters are $\alpha = 2, v = 4$, and $\kappa^C = 2$, and Da is set by varying ℓ_r . Dash–dotted lines (overlapping the solid lines for low Da) are numerical solutions of Eq. (17) using $\langle C_M C_I \rangle = \bar{c}_M \bar{c}_I$. Left: Average mobile concentration. The dashed line is the analytical solution in the limit of small Damköhler. Right: Concentration fluctuations.

³⁴¹ number $\text{Da}(v) = \kappa^C \mu(v) \overline{c}_{0,I}(v) \to 0$, the changes in immobile concentration due to reaction ³⁴² in each region become arbitrarily small, so that $\langle C_I C_M \rangle \approx \overline{c}_{0,I}(v) \langle C_M \rangle$ at all times. Thus, ³⁴³ for small Da, the late-time rate equation for the mobile component under *mild segregation* ³⁴⁴ is, according to Eq. (19),

$$\frac{d\overline{c}_M(t|\boldsymbol{c}_0, v)}{dt} = -\frac{\mathrm{Da}(v)}{(1+\alpha)\mu(v)}\overline{c}_M(t|\boldsymbol{c}_0, v).$$
(22)

There is no appreciable reaction before the late-time equation is valid, so that the initial 346 condition $c_{0,M}$ may be employed, and

$$\overline{c}_M(t|\boldsymbol{c}_0, v) = c_{0,M} \exp\left[-\frac{\mathrm{Da}(v)t}{(1+\alpha)\mu(v)}\right].$$
(23)

³⁴⁷ Under strong segregation, the Laplace transform of the late-time equation (21) is, for ³⁴⁸ $Da(v) \ll 1$,

$$\frac{[\alpha\mu(v)\lambda]^{\beta}}{\mathrm{Da}(v)} \left[\tilde{\bar{c}}_{M}(t|\boldsymbol{c}_{0},v) - \lambda^{-1}c_{0,M}\right] = -\tilde{\bar{c}}_{M}(\lambda|\boldsymbol{c}_{0},v).$$
(24)



Figure 4. Temporal evolution of concentration for $S_M + S_I \rightarrow_{\kappa} \emptyset$ under strong segregation for different values of Damköhler number Da. Symbols represent stochastic simulations based on the Gillespie algorithm (10⁷ initial mobile particles and $5 \cdot 10^6$ immobile particles per reactive region) and solid lines employ the well-mixed rate equations in reactive regions. All results are averaged over 10⁵ realizations. Non-reactive region lengths are Lévy-stable-distributed with exponent $\beta =$ 0.7. Fixed parameters are $\alpha = 2$, v = 4, and $\kappa^C = 2$, and Da is set by varying ℓ_r . Left: Average mobile concentration. The dashed line is the late-time analytical solution in the limit of small Damköhler. Right: Concentration fluctuations.

³⁴⁹ Noticing that, for small λ , the initial-condition term on the left-hand side dominates, and ³⁵⁰ inverting the Laplace transform,

$$\bar{c}_M(t|\boldsymbol{c}_0, v) \approx \frac{c_{0,M}}{\mathrm{Da}(v)\Gamma(1-\beta)} \left[\frac{t}{\alpha\mu(v)}\right]^{-\beta}.$$
(25)

Simulation results for mild segregation are shown in Fig. 3 and for strong segregation 351 $_{352}$ in Fig. 4. We consider for concreteness exponential ρ_c for mild and Lévy-stable ρ_c for ³⁵³ strong segregation, as before. Note that, unlike in Fig. 2, time is nondimensionalized so as to highlight the collapse of the low-Damköhler behavior onto the Da-independent curve 354 valid for both the full degradation reaction and the catalytic degradation reaction \mathcal{S}_M + 355 $_{356} S_I \rightarrow_{\kappa} S_I$. Under this nondimensionalization, higher Da leads to slower decay due to the ³⁵⁷ effect of depletion of the immobile component. We compare a full stochastic algorithm employing the Gillespie method in reactive regions to a more efficient algorithm, valid for 358 large particle numbers, which makes use of the well-mixed rate equations as in Section III 359 (see Appendix \mathbf{E}). The results are in very good agreement. 360

For mild segregation, simulations suggest that $\langle C_I C_M \rangle \approx \overline{c}_{0,I}(v) \langle C_M \rangle$ holds at late times, as expected under finite-mean inter-reaction times (see the right panel of Fig. 3). Thus, the late-time concentration for each velocity v decays in general exponentially as argued above, but the leading coefficient differs because the appropriate initial condition for the late-time equation depends on the dynamics before it becomes valid. However, for large Da, the exponential asymptotic regime is not observed, as the concentration reaches very small values before it occurs. We show also the solution obtained by numerically integrating the closed rate equations obtained by setting $\langle C_M C_I \rangle = \bar{c}_M \bar{c}_I$ in Eq. (17) (see Appendix F). Its breakdown for Da $\gtrsim 1$ is due to the role of concentration fluctuations. Even though the fluctuations vanish at late times, they have an irreversible impact on the total reaction.

³⁷¹ Under strong segregation, the leading coefficient varies with Da for two reasons. First, ³⁷² the initial condition to be used with the asymptotic equation differs as above. Second, ³⁷³ the weak ergodicity breaking displayed by the inter-reaction times impedes $\langle C_I(t)C_M(t)\rangle \rightarrow$ ³⁷⁴ $\bar{c}_{0,I}\langle C_M(t)\rangle$ for late times, because there is a sufficiently high probability that $C_M(t)$ remains ³⁷⁵ large due to long non-reactive regions. Nonetheless, simulations suggest that $\langle C_I(t)C_M(t)\rangle \propto$ ³⁷⁶ $\bar{c}_I(t)\bar{c}_M(t) \approx c_{0,I}\bar{c}_M(t)$ at late times for all values of Da, see the right panel of Fig. 4, so that ³⁷⁷ the power-law behavior $\propto t^{-\beta}$ remains unaffected.

378 VI. CONCLUSIONS

The link between first-passage and inter-reaction times connects the kinetics of contact processes to spatial heterogeneity and segregation. We have quantified this link for advective transport under spatial segregation of immobile components and formulated the reaction dynamics in terms of a generalized master equation. The evolution of total mass may be obtained from this description, and mass fluxes and spatial reactant distributions may also be easily computed.

In contrast to the classical picture for well-mixed reactions, the resulting large-scale kinetics cannot be fully quantified in terms of the dynamical equations obtained by averaging over the chemical master equation. This is due to the presence of concentration fluctuations fluctuations on the order of the average values, which result from a combination of the restart mechanism and memory effects, both caused by segregation. For this reason, closed-form rate laws valid for all times do not exist in general. In the case of strong segregation, characterized by infinite-mean inter-reactive-region distances, memory is long-range and induces weak generating across trajectories, a typical feature of anomalous transport [39–44]. Future work will focus on extending this approach to more complex transport processes, including effects such as variable velocity within each trajectory [45, 46] and local mixing (e.g., diffusion). Moment closure approximations for the rate equations will also be the subject of further study.

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400 Appendix A: Well-mixed degradation kinetics

Consider the degradation reaction $S_M + S_I \rightarrow_{\kappa} \emptyset$ introduced in the main text. In a 402 well-mixed reactor, the kinetic rate laws are given by Eq. (1). For equal initial conditions, 403 $\mathbf{c}_0 = (c_{0,M}, c_{0,I})$ with $c_{0,M} = c_{0,I} = c_0$, the solution is

$$c_{M,I}^{wm}(t|\boldsymbol{c}_0) = \frac{c_0}{1 + \kappa^C c_0 t}.$$
 (A1)

404 For $c_{0,M} \neq c_{0,I}$, setting $c_{\max} = \max\{c_M^{wm}, c_I^{wm}\}$ and $c_{\min} = \min\{c_M^{wm}, c_I^{wm}\}$,

$$c_{\min}(t|\mathbf{c}_{0}) = c_{0,\min}g(t,\mathbf{c}_{0})e^{-\kappa^{C}t(c_{0,\max}-c_{0,\min})},$$

$$c_{\max}(t|\mathbf{c}_{0}) = c_{0,\max}g(t,\mathbf{c}_{0}),$$
(A2)

405 where

$$g(t, \boldsymbol{c}_0) = \frac{c_{0,\max} - c_{0,\min}}{c_{0,\max} - c_{0,\min} e^{-\kappa^C t(c_{0,\max} - c_{0,\min})}}.$$
(A3)

The simulations of the evolution of concentration due to this reaction in the presence 407 of segregation proceed as follows. A reactive region length ℓ is generated according to the 408 PDF ρ_r . The solution of the well-mixed rate laws is then applied to obtain the evolution 409 of concentrations for a time interval ℓ/v . Then, a non-reactive region length ℓ' is generated 410 according to ρ_c , and no reaction occurs for a time interval ℓ'/v . This procedure is iterated up 411 to a desired time, with the initial condition for the well-mixed rate laws in each reactive patch 412 being set according to the current mobile concentration and the initial resident immobile 413 concentration. If one considers instead the catalytic degradation reaction $S_M + S_I \rightarrow_{\kappa} S_I$ in a well-mixed reactor, the immobile species concentration does not change and remains equal to $c_{0,I}$. The rate law for the mobile component is the same as before, and we obtain

$$c_M^{wm}(t|\boldsymbol{c}_0) = e^{-\kappa^C c_{0,I} t}.$$
(A4)

⁴¹⁷ The simulations under segregation proceed in the same manner as above, using this solution ⁴¹⁸ for the mobile concentration.

419 Appendix B: Intrinsic inter-reaction times

This Appendix provides a brief review of the intrinsic inter-reaction times, which char-421 acterize reactions in the absence of segregation. These correspond to the classical stochas-422 tic theory of well-mixed reaction [24]. Consider a given chemical state \boldsymbol{n} . Each reac-423 tion $i = 1, \ldots, m_r$ considered in isolation has an exponential inter-reaction time with rate 424 $a_i(\boldsymbol{n}) = \kappa_i h_i(\boldsymbol{n})$, where κ_i is a (microscopic) rate constant and $h_i(\boldsymbol{n})$ encodes the dependency 425 on the chemical state. For mass-action reactions,

$$h_i(\boldsymbol{n}) = \prod_{j=1}^{m_s} \frac{n_j!}{r_{ij}!(n_j - r_{ij})!}.$$
(B1)

⁴²⁶ It follows from the exponential character of reaction times, and the fact that the next reaction ⁴²⁷ to fire is the one with the minimum waiting time, that the inter-reaction time density is

$$\phi_i^r(t|\boldsymbol{n}) = \frac{a_i(\boldsymbol{n})}{a(\boldsymbol{n})} \phi^r(t|\boldsymbol{n}), \tag{B2}$$

⁴²⁸ where $a(\boldsymbol{n}) = \sum_{i=1}^{m_r} a_i(\boldsymbol{n})$, and

$$\phi^r(t|\boldsymbol{n}) = a(\boldsymbol{n})e^{-a(\boldsymbol{n})t},\tag{B3}$$

⁴²⁹ with $\phi^r(t|\mathbf{n}) dt$ the probability that the inter-reaction time is in [t, t + dt]. The Laplace ⁴³⁰ transform of the inter-reaction time density is thus

$$\tilde{\phi}_i^r(\lambda|\boldsymbol{n}) = \frac{a_i(\boldsymbol{n})}{\lambda + a(\boldsymbol{n})}.$$
(B4)

431 Appendix C: Number of traversed reactive regions

Here, we determine the distribution $\nu_r(\cdot|t_r, v)$ of the number $\eta_r(t_r, v)$ of fully traversed reactive regions between reactions, given time t_r spent in reactive regions and velocity v. ⁴³⁴ The length traversed in time t_r is vt_r . The number $\eta_r(t_r, v)$ of reactive regions traversed in ⁴³⁵ this time is such that their total length is smaller than vt_r , but the total length of $\eta_r(t_r, v) + 1$ ⁴³⁶ regions is larger than vt_r . Thus,

$$\nu_r(k|t_r, v) = \left\langle H\left(vt_r - \sum_{k'=1}^k L_{r,k'}\right) \times H\left(\sum_{k'=1}^{k+1} L_{r,k'} - vt_r\right) \right\rangle,$$
(C1)

⁴³⁷ where *H* is the Heaviside step function and the $L_{r,k'}$ are i.i.d. according to ρ_r . Conditioning ⁴³⁸ on the total length of the first *k* regions, we obtain

$$\nu_r(k|t_r, v) = \int_0^{vt_r} d\ell \left\langle \delta \left(\ell - \sum_{k'=1}^k L_{r,k} \right) \right\rangle$$
$$\times \int_{vt_r-\ell}^\infty d\ell' \, \rho_r(\ell'). \tag{C2}$$

439 Taking the Laplace transform with respect to t_r ,

$$\tilde{\nu}_r(k|\lambda, v) = \frac{1 - \tilde{\psi}_r(\lambda|v)}{\lambda} \tilde{\psi}_r(\lambda|v)^k.$$
(C3)

⁴⁴⁰ Thus, using $\rho_r(\ell) = e^{-\ell/\ell_r}/\ell_r$ and $\psi_r(t|v) = v\rho_r(vt)$, we have $\tilde{\nu}_r(k|\lambda, v) = \mu(v)[1+\mu(v)\lambda]^{-k}$, ⁴⁴¹ with $\mu(v) = \ell_r/v$. Inverting the Laplace transform,

$$\nu_r(k|t_r, v) = \frac{[t_r/\mu(v)]^k}{k!} e^{-t_r/\mu(v)},$$
(C4)

442 so that $\eta_r(t_r, v)$ is Poisson-distributed with mean $t_r/\mu(v)$.

Note that, as long as reactive region lengths have a finite mean ℓ_r , we have, for small $\lambda_{444} \lambda \ll 1/\mu(v)$ (corresponding to the large $t_r \gg \mu(v)$ limit), $\tilde{\psi}_r(\lambda|v) \approx 1-\mu(v)\lambda \approx 1/[1+\mu(v)\lambda]$, the Laplace transform of the exponential density. Thus, the distribution of the number of traversed reactive patches is always approximately Poisson for large t_r . Since typical reaction times are large when the Damköhler number is low, as explained in the main text, the specific distribution of reactive lengths does not play a role in that case.

⁴⁴⁹ Appendix D: Generalized master equation under restart

⁴⁵⁰ Consider the process $\mathcal{K}(t)$, which describes the number of reactions as a function of ⁴⁵¹ time, and write $N_{\mathcal{K}(t)} = N(t)$. We have $\mathcal{K}(t) = \sup\{k \mid T_k < t\}$, where T_k is the time ⁴⁵² of the kth reaction. The propagator can then be written as $p(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \langle \delta_{\boldsymbol{n}, \boldsymbol{N}_{\mathcal{K}(t)}} | \boldsymbol{n}_0, v \rangle$. ⁴⁵³ Conditioning on K(t) = k and $T_k = t'$,

$$p(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \int_0^\infty dt' \sum_{k \ge 0} \langle \delta_{\boldsymbol{n}, \boldsymbol{N}_k} \delta(T_k - t') \\ \times H(t - t') H[\tau_k - (t - t')] | \boldsymbol{n}_0, v \rangle,$$
(D1)

454 with the inter-reaction times $\tau_k = T_{k+1} - T_k$ independent of T_k . Thus,

$$p(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \int_0^t dt' \sum_{k \ge 0} R_k(\boldsymbol{n}, t' | \boldsymbol{n}_0, v)$$
$$\times \sum_{i=0}^{m_r} \int_{t-t'}^\infty dt'' \phi_i(t'' | \boldsymbol{n}, v),$$
(D2)

⁴⁵⁵ where $R_k(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \langle \delta_{\boldsymbol{n}, \boldsymbol{N}_k} \delta(T_k - t) | \boldsymbol{n}_0, v \rangle$ is the probability per time of arriving at state ⁴⁵⁶ \boldsymbol{n} at time t given k reaction steps. The dynamics are Markov in reaction step number k, ⁴⁵⁷ and $R_k(\boldsymbol{n}, t | \boldsymbol{n}_0, v)$ solves the Chapman-Kolmogorov equation

$$R_{k+1}(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \int_0^t dt' \sum_{\boldsymbol{n}' \ge 0} \sum_{i=0}^{m_r} \gamma_i(\boldsymbol{n} | \boldsymbol{n}', v)$$
$$\times \phi_i(t - t' | \boldsymbol{n}', v) R_k(\boldsymbol{n}', t'), \tag{D3}$$

⁴⁵⁸ with $R_0(\boldsymbol{n}, t | \boldsymbol{n}_0, v) = \delta_{\boldsymbol{n}, \boldsymbol{n}_0} \delta(t), \gamma_i(\boldsymbol{n} | \boldsymbol{n}', v) = \gamma_i(\boldsymbol{n} | \boldsymbol{n}')$ given by Eq. (4), and $\gamma_0(\boldsymbol{n} | \boldsymbol{n}', v)$ given ⁴⁵⁹ by Eq. (14). Laplace transforming Eqs. (D2) and (D3) summed over k,

$$\tilde{R}(\boldsymbol{n},\lambda|\boldsymbol{n}_{0},v) = \delta_{\boldsymbol{n},\boldsymbol{n}_{0}} + \sum_{\boldsymbol{n}'\geq 0} \sum_{i=0}^{m_{r}} \gamma_{i}(\boldsymbol{n}|\boldsymbol{n}',v) \tilde{\phi}_{i}(\lambda|\boldsymbol{n}',v) \tilde{R}(\boldsymbol{n}',\lambda|\boldsymbol{n}_{0},v),$$

$$\tilde{p}(\boldsymbol{n},\lambda|\boldsymbol{n}_{0},v) = \tilde{R}(\boldsymbol{n},\lambda|\boldsymbol{n}_{0},v) \frac{1-\sum_{i=0}^{m_{r}} \tilde{\phi}_{i}(\lambda|\boldsymbol{n},v)}{\lambda},$$
(D4)

⁴⁶⁰ where $R(\boldsymbol{n},t|\boldsymbol{n}_0,v) = \sum_{k\geq 0} R_k(\boldsymbol{n},t|\boldsymbol{n}_0,v)$. Eliminating \tilde{R} , we find

$$\lambda \tilde{p}(\boldsymbol{n}, \lambda | \boldsymbol{n}_{0}, v) - \delta_{\boldsymbol{n}, \boldsymbol{n}_{0}} = \sum_{\boldsymbol{n}' \ge 0} \sum_{i=0}^{m_{r}} \left[\gamma_{i}(\boldsymbol{n} | \boldsymbol{n}', v) - \delta_{\boldsymbol{n}, \boldsymbol{n}'} \right] \\ \times \tilde{M}_{i}(\lambda | \boldsymbol{n}', v) \tilde{p}(\boldsymbol{n}', \lambda | \boldsymbol{n}_{0}, v),$$
(D5)

⁴⁶¹ where $\tilde{M}_i(\lambda | \boldsymbol{n}, v)$ is given according to Eq. (6), for all $0 \leq i \leq m_r$ and with the sum extending ⁴⁶² to i = 0. Laplace inversion leads to the same form as the generalized master equation (5), ⁴⁶³ with the sum extending to i = 0.

464 Appendix E: Stochastic simulation algorithms

We describe a generalized Gillespie algorithm that takes into account restart as described in the main text. Velocity v is to be sampled from the PDF $\xi(\cdot)$. In order to simulate dot dynamics up to time t_m (or distance ℓ_m), the following algorithm should be repeated for a prescribed number of realizations:

469 1. Set time t = 0 (or distance $\ell = 0$). Generate \boldsymbol{n} according to the initial state distribu-470 tion.

471 2. Generate ℓ_1 according to ρ_r and Δt_r according to $\phi^r(\cdot|\mathbf{n})$, see Eq. (B3).

3. Find the next reaction *i*: If $\Delta t_r > \ell_1/v$ set i = 0. Else generate $0 < r \leq a(\mathbf{n})$ from the uniform distribution and set *i* such that $a_i(\mathbf{n}) < r \leq a_{i+1}(\mathbf{n})$, see Eq. (B3).

474 4. If i = 0 generate ℓ_2 according to ρ_c and set $\Delta t = (\ell_1 + \ell_2)/v$. Else set $\Delta t = \Delta t_r$. 475 Increment t by Δt (ℓ by $v\Delta t$).

5. If $t < t_m$ ($\ell < \ell_m$) update \boldsymbol{n} according to $\gamma_i(\cdot | \boldsymbol{n}, v)$, see Eq. (4), and go to 2. Else set $t = t_m$ ($\ell = \ell_m$) and end.

⁴⁷⁸ For large particle numbers, this procedure may be replaced by a more efficient method to ⁴⁷⁹ find the concentrations:

- 480 1. Set time t = 0 (or distance $\ell = 0$). Generate c according to the initial concentration 481 distribution.
- ⁴⁸² 2. Generate ℓ_1 according to ρ_r and Δt_r according to $\phi^r(\cdot|\boldsymbol{n})$, see Eq. (B3). Set $\Delta t = \min\{\Delta t_r, t_m t\}$.
- 3. Update c according to the well-mixed rate equations over the interval $[t, t + \Delta t]$. Increment t by Δt (ℓ by $v\Delta t$). If $t = t_m$ ($\ell = \ell_m$) end.
- 486 4. Generate ℓ_2 according to ρ_c . Increment t by $\Delta t = \min\{(\ell_1 + \ell_2)/v, t t_m\}$ (ℓ by $v\Delta t$).

487 5. If $t < t_m$ ($\ell < \ell_m$) generate c_I according to the initial concentration distribution and 488 go to 2. Else end.

⁴⁸⁹ Note that this algorithm reduces to the one outlined in Appendix A for the reactions con-⁴⁹⁰ sidered therein.

491 Appendix F: Large-scale kinetics

Following [28], we define $p^{C}(\boldsymbol{c},t|\boldsymbol{c}_{0},v) = \overline{n}p(\overline{n}\boldsymbol{c},t|\overline{n}\boldsymbol{c}_{0},v), \ M_{i}^{C}(t|\boldsymbol{c},v) = M_{i}(t|\overline{n}\boldsymbol{c},v)/\overline{n},$ $i = 1, \ldots, m_{r}, \text{ and } M_{0}^{C}(t|\boldsymbol{c},v) = M_{0}(t|\overline{n}\boldsymbol{c},v).$ We multiply Eq. (D5) by, and sum over, $\boldsymbol{n},$ $i = 1, \ldots, m_{r}, = \overline{n} \int d\boldsymbol{c}$ in the limit $\overline{n} \to \infty$. Integrating by parts, we obtain

$$\lambda \tilde{\overline{\boldsymbol{c}}}(\lambda | \boldsymbol{c}_{0}, v) - \boldsymbol{c}_{0} = \sum_{i=1}^{m_{r}} \boldsymbol{s}_{i} \mathcal{L}_{t \to \lambda} \langle \tilde{M}_{i}^{C}[\lambda | \boldsymbol{C}(t)] | \boldsymbol{c}_{0}, v \rangle$$
$$- \mathcal{L}_{t \to \lambda} \langle [\boldsymbol{C}(t) - (\boldsymbol{C}_{M}(t), \overline{\boldsymbol{c}}_{0,I}(v))] \tilde{M}_{0}^{C}[\lambda | \boldsymbol{C}(t)] | \boldsymbol{c}_{0}, v \rangle, \tag{F1}$$

⁴⁹⁵ where $\mathcal{L}_{t\to\lambda}f(t) = \tilde{f}(\lambda)$. Substituting Eq. (6) for the memory functions, we have,

$$\lambda \tilde{\overline{c}}_{M}(\lambda | c_{0}, v) - c_{0,M} = \sum_{i=1}^{m_{r}} s_{i}^{M} \tilde{K}(\lambda | v) \mathcal{L}_{t \to \lambda} \langle a_{i}^{C} [C(t)] | c_{0}, v \rangle,$$

$$\left[\lambda \tilde{\overline{c}}_{I}(\lambda | c_{0}, v) - \overline{c}_{0,I}(v)\right] \left[1 + \frac{\tilde{K}(\lambda | v) \tilde{\psi}_{c}(\lambda | v)}{\mu(v) \lambda}\right] = +\sum_{i=1}^{m_{r}} s_{i}^{I} \tilde{K}(\lambda | v) \mathcal{L}_{t \to \lambda} \langle a_{i}^{C} [C(t)] | c_{0}, v \rangle.$$
(F2)

⁴⁹⁶ Using Eq. (3) for the memory kernel, we find that $1 - \tilde{K}(\lambda|v)[1 - \tilde{\psi}_c(\lambda|v)]/[\mu(v)\lambda] = \tilde{K}(\lambda|v)$, ⁴⁹⁷ so that the second equation may be rearranged and divided by $\tilde{K}(\lambda|v)$ to give

$$\lambda \tilde{\overline{c}}_{I}(\lambda | \boldsymbol{c}_{0}, v) - \overline{\boldsymbol{c}}_{0,I}(v) = -\frac{\lambda \tilde{\overline{c}}_{I}(\lambda | \boldsymbol{c}_{0}, v) - \overline{\boldsymbol{c}}_{0,I}(v)}{\mu(v)\lambda} + \sum_{i=1}^{m_{r}} \boldsymbol{s}_{i}^{I} \mathcal{L}_{t \to \lambda} \langle \boldsymbol{a}_{i}^{C}[\boldsymbol{C}(t)] | \boldsymbol{c}_{0}, v \rangle.$$
(F3)

⁴⁹⁸ Inverse-Laplace-transforming yields Eq. (17).

In order to integrate Eqs. (17) for the degradation reaction $S_M + S_I \rightarrow \emptyset$, under the assumption $\langle C_I C_M \rangle = \bar{c}_I \bar{c}_M$ and mild segregation, note that the memory kernel (3) reads $\tilde{K}(\lambda|v) = 1 - \alpha/[1 + \alpha + \alpha\mu(v)\lambda]$, so that, inverting the Laplace transform, K(t|v) = $_{502} \delta(t) - \exp\{-(1+\alpha)t/[\alpha\mu(v)]\}/\mu(v)$. Equations (17) then read

$$\frac{d\overline{c}_{M}(t|\boldsymbol{c}_{0},v)}{dt} = -\kappa^{C}\overline{c}_{M}(t|\boldsymbol{c}_{0},v)\overline{c}_{I}(t|\boldsymbol{c}_{0},v) + \frac{\kappa^{C}}{\mu(v)}\int_{0}^{t} dt' e^{-\frac{1+\alpha}{\alpha\mu(v)}(t-t')}\overline{c}_{M}(t'|\boldsymbol{c}_{0},v)\overline{c}_{I}(t'|\boldsymbol{c}_{0},v), \quad (F4)$$

$$\frac{d\overline{c}_{I}(t|\boldsymbol{c}_{0},v)}{dt} = -\kappa^{C}\overline{c}_{M}(t|\boldsymbol{c}_{0},v)\overline{c}_{I}(t|\boldsymbol{c}_{0},v) - \frac{\overline{c}_{I}(t|\boldsymbol{c}_{0},v)}{\mu(v)}.$$

⁵⁰³ We solve these equations numerically using a combination of the forward Euler method for ⁵⁰⁴ the derivative with the trapezoidal rule for the convolution integral. We use a discretization ⁵⁰⁵ time step $\Delta t = 10^{-2} \min\{\mu, 1/(\kappa^C c_{0,M}), 1/(\kappa^C c_{0,I})\}.$

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