

REACTION-DIFFUSION EQUATIONS

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ABSTRACT. Notes on reaction-diffusion equations and some biological applications. If unspecified, maps are assumed to be as smooth as necessary.

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1. THE DIFFUSION EQUATION

Let Ω be a domain of \mathbb{R}^n , and consider a map $u : \Omega \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ where we denote a point of the domain by (x, t) (representing spatial and time coordinates respectively). Then the general form of a single reaction-diffusion equation is

$$u_t = D\Delta u + f \circ u,$$

where $f : \mathbb{R} \rightarrow \mathbb{R}$ is a smooth map that describes the ‘reaction’, while $D > 0$ is the diffusion coefficient for the Laplace operator on the spatial coordinates. A solution u is a state variable that typically describes the concentration or density of a substance or population. For more general situations, f may depend explicitly on t , x , and the gradient of u .

Remark 1.1. Throughout these notes a domain is an open connected set while a region is a set whose interior is a domain.

1.1. Derivation. Let $u : \mathbb{R}^3 \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ represent a density and $J : \mathbb{R}^3 \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}^3$ its diffusion flux (amount of substance per unit volume per unit time). For a compact region $\Omega \subset \mathbb{R}^3$ with piecewise smooth boundary Γ , *ceteris paribus* (assuming only conservation of mass) we have

$$\frac{d}{dt} \int_{\Omega} u(x, t) dV = - \int_{\Gamma} J(x, t) \cdot dS.$$

The divergence theorem implies that

$$\int_{\Gamma} J \cdot dS = \int_{\Omega} \operatorname{div} J dV,$$

and so we obtain

$$\int_{\Omega} (u_t + \operatorname{div} J) dV = 0.$$

As this holds for all volumes Ω , we have

$$u_t + \operatorname{div} J = 0, \tag{2}$$

called the *continuity equation*. Now Fick's first law asserts that

$$J = -D \nabla u,$$

which says that the only flux is diffusive flux. Combining this with (2) we arrive at Fick's second law, bearing an uncanny resemblance to the heat equation:

$$u_t = D \Delta u. \tag{3}$$

This is a simple case of the diffusion equation where D is constant.

1.2. Interlude on the Leibniz Integral Rule. In PDE we often have occasion to differentiate integrals, and there are many different variations of this scattered throughout the literature. Here we make no attempt at generality; we only present a proof of the quite simple case required in the derivation above.

Proposition 1.4. *Let $\Omega \subset \mathbb{R}^3$ be a compact region with measure zero boundary and $u : \mathbb{R}^3 \times \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ a smooth map. Then*

$$\frac{d}{dt} \int_{\Omega} u(x, t) dx = \int_{\Omega} u_t(x, t) dx.$$

Proof. First observe that since u is continuously differentiable, both integrals exist for all t . Now writing

$$\frac{d}{dt} \int_{\Omega} u(x, t) dx = \lim_{h \rightarrow 0} \int_{\Omega} \frac{u(x, t+h) - u(x, t)}{h} dx,$$

we want to show that for all $\epsilon > 0$ we can choose h such that

$$\left| \int_{\Omega} \left(\frac{u(x, t+h) - u(x, t)}{h} - u_t(x, t) \right) dx \right| < \epsilon.$$

Applying the mean value theorem and invoking the uniform continuity of u_t , we obtain the desired result. \square

1.3. The Fundamental Solution. Consider the problem

$$u_t = Du_{xx} \quad \text{for } x \in \mathbb{R}, \ t > 0; \quad (5)$$

$$u(x, 0) = \phi(x). \quad (6)$$

We first make some observations regarding the ‘invariance’ of solutions (ignoring the initial condition).

- (i) Any linear combination of solutions is also a solution.
- (ii) Any translation of a solution is also a solution.
- (iii) Any scaling $u(\sqrt{a}x, at)$ for $a > 0$ of a solution $u(x, t)$ is also a solution.
- (iv) Any derivative of a (sufficiently nice) solution is also a solution.
- (v) Any integral of a solution is also a solution in the following way. If $u(x, t)$ is a solution, then so is the convolution

$$v(x, t) = \int_{-\infty}^{\infty} u(x - y, t) f(y) dy$$

where f is sufficiently nice (so that the integral converges in a sufficiently nice manner).

To give an example of why these hold, we show (ii). Let $T : \mathbb{R} \rightarrow \mathbb{R}$ be a translation defined by $x \mapsto x - a$. If u solves $u_t = Du_{xx}$, consider $\bar{u}(x, t) = u(T(x), t)$. We then have

$$\bar{u}_t(x, t) = u_t(T(x), t)$$

and

$$\bar{u}_x(x, t) = u_x(T(x), t)T'(x) = u_x(T(x), t), \text{ so that } \bar{u}_{xx}(x, t) = u_{xx}(T(x), t).$$

Let us suppose for a moment that our initial condition is given by

$$Q(x, 0) = \begin{cases} 1 & \text{for } x > 0 \\ 0 & \text{for } x < 0. \end{cases}$$

To determine $Q(x, t)$, we begin with invariance under scaling (as this leaves the initial condition unchanged). In particular we want $Q(x, t)$ to remain unchanged under scaling. Of course, if the dependency of Q on x and t is only on the ratio x/\sqrt{t} we have such an invariance, and hence we restrict Q to be of the following form (with an additional factor for convenience later):

$$Q(x, t) = g(p), \text{ where } p = \frac{x}{\sqrt{4Dt}}.$$

Now this allows us to turn (5) into an ordinary differential equation that we can solve. We have

$$\begin{aligned} Q_t &= \frac{dg}{dp} \frac{\partial p}{\partial t} = -\frac{x}{2t\sqrt{4Dt}} g'(p) \\ Q_x &= \frac{dg}{dp} \frac{\partial p}{\partial x} = \frac{1}{\sqrt{4Dt}} g'(p) \\ Q_{xx} &= \frac{dQ_x}{dp} \frac{\partial p}{\partial x} = \frac{1}{4Dt} g''(p), \end{aligned}$$

and hence

$$0 = Q_t - DQ_{xx} = -\frac{x}{2t\sqrt{4Dt}} g'(p) - \frac{1}{4t} g''(p) = \frac{1}{t} \left(-\frac{p}{2} g'(p) - \frac{1}{4} g''(p) \right).$$

We thus have the ODE

$$g'' + 2pg' = 0,$$

and a first integration yields

$$g' = c_1 e^{-p^2}.$$

Integrating once more we arrive at the solution

$$g(p) = c_1 \int e^{-p^2} dp + c_2.$$

To impose conditions on c_1 and c_2 we let

$$Q(x, t) = c_1 \int_0^{x/\sqrt{4Dt}} e^{-p^2} dp + c_2,$$

which is valid for all $t > 0$. Then the initial conditions must be satisfied in the limit $t \searrow 0$; for $x > 0$ we have

$$1 = c_1 \int_0^\infty e^{-p^2} dp + c_2 = c_1 \frac{\sqrt{\pi}}{2} + c_2;$$

for $x < 0$ we have

$$0 = c_1 \int_0^{-\infty} e^{-p^2} dp + c_2 = -c_1 \frac{\sqrt{\pi}}{2} + c_2.$$

This implies that $c_1 = \frac{1}{\sqrt{\pi}}$ and $c_2 = \frac{1}{2}$, so that

$$Q(x, t) = \frac{1}{2} + \frac{1}{\sqrt{\pi}} \int_0^{x/\sqrt{4Dt}} e^{-p^2} dp.$$

Now let $S = Q_x$, which also solves (5). For an arbitrary smooth map ϕ where $\phi(x) = 0$ when $|x|$ is sufficiently large, let

$$u(x, t) = \int_{\mathbb{R}} S(x - y, t) \phi(y) dy.$$

We know that u solves (5), and want to show it satisfies (6) as well. To this end, we observe that integrating by parts gives

$$\begin{aligned} u(x, t) &= \int_{\mathbb{R}} \frac{\partial Q}{\partial x}(x - y, t) \phi(y) dy \\ &= - \int_{\mathbb{R}} \frac{\partial Q}{\partial y}(x - y, t) \phi(y) dy \\ &= -\phi(y) Q(x - y, t) \Big|_{-\infty}^{\infty} + \int_{\mathbb{R}} \phi'(y) Q(x - y, t) dy. \end{aligned}$$

Then

$$u(x, 0) = \int_{\mathbb{R}} Q(x - y, 0) \phi'(y) dy = \int_{-\infty}^x \phi'(y) dy = \phi(x).$$

Computing the spatial derivative of Q we find

$$S = \frac{1}{\sqrt{4\pi Dt}} e^{-x^2/4Dt},$$

and so

$$u(x, t) = \frac{1}{\sqrt{4\pi Dt}} \int_{\mathbb{R}} e^{-(x-y)^2/4Dt} \phi(y) dy. \quad (7)$$

This is called the *fundamental solution*, and while it is rare to be able to solve the integral explicitly, for some ϕ it may be written using the well-known error function

$$\operatorname{erf} x = \frac{2}{\sqrt{\pi}} \int_0^x e^{-p^2} dp$$

that satisfies

$$\operatorname{erf}(0) = 0 \quad \text{and} \quad \lim_{x \rightarrow \infty} \operatorname{erf}(x) = 1.$$

1.4. The Inhomogeneous Equation. Here we solve

$$\begin{aligned} u_t - Du_{xx} &= f(x, t) \quad \text{for } x \in \mathbb{R}, t > 0 \\ u(x, 0) &= \phi(x) \end{aligned} \tag{8}$$

where ϕ describes the initial distribution and f is a ‘source’ or ‘sink’ term. But let us first consider the analogous problem for ordinary differential equations.

For constant $A \in \mathbb{R}^{n \times n}$, $f \in C((a, b), \mathbb{R}^n)$, and $0 \in (a, b)$ with $-\infty \leq a < b \leq \infty$, the initial-value problem

$$u' - Au = f, \quad u(0) = u_0$$

is solved (using variation of parameters) by

$$u(t) = e^{At}u_0 + \int_0^t e^{A(t-s)}f(s)ds.$$

Indeed, with $f \equiv 0$, the solution reads $u(t) = S(t)u_0$ where $S(t) = e^{At}$. Then when $f \not\equiv 0$, we multiply the equation by integrating factor $S(-t)$:

$$S(-t)u'(t) - S(-t)Au(t) = \frac{d}{dt}[S(-t)u(t)] = S(-t)f(t).$$

Integrating this from 0 to t we obtain

$$S(-t)u(t) = u_0 + \int_0^t S(-s)f(s)ds$$

and then multiplying by $S(t)$ yields

$$u(t) = S(t)u_0 + \int_0^t S(t-s)f(s)ds.$$

In analogy to this, we want to show that

$$u(x, t) = \int_{\mathbb{R}} S(x-y, t)\phi(y)dy + \int_0^t \int_{\mathbb{R}} S(x-y, t-s)f(y, s)dyds \tag{9}$$

is a solution to (8). Note that it suffices to check the second term. We have

$$\begin{aligned} u_t &= \frac{\partial}{\partial t} \int_0^t \int_{\mathbb{R}} S(x-y, t-s)f(y, s)dyds \\ &= \int_0^t \int_{\mathbb{R}} S_t(x-y, t-s)f(y, s)dyds + \lim_{s \rightarrow t} \int_{\mathbb{R}} S(x-y, t-s)f(y, s)dy \\ &= \int_0^t \int_{\mathbb{R}} DS_{xx}(x-y, t-s)f(y, s)dyds + \lim_{\epsilon \rightarrow 0} \int_{\mathbb{R}} S(x-y, \epsilon)f(y, t)dy \\ &= D \frac{\partial^2}{\partial x^2} \int_0^t \int_{\mathbb{R}} S(x-y, t-s)f(y, s)dyds + f(x, t) \\ &= Du_{xx} + f \end{aligned}$$

Observing that

$$\lim_{t \rightarrow 0} u(x, t) = \phi(x),$$

we have the desired solution. This is often referred to as *Duhamel's principle*.

1.5. Uniqueness on Cylindrical Domains. Now let us consider the Dirichlet problem for the diffusion equation with inhomogeneity f :

$$\begin{aligned} u_t - Du_{xx} &= f(x, t) \quad \text{for } 0 < x < l \text{ and } t > 0; \\ u(x, 0) &= \phi(x); \\ u(0, t) &= g(t); \\ u(l, t) &= h(t). \end{aligned} \tag{10}$$

Proposition 1.11. *The Dirichlet problem as given above has at most one solution.*

Proof. Assume u_1 and u_2 both satisfy (10), and let $w = u_1 - u_2$. Then w satisfies

$$\begin{aligned} w_t - Dw_{xx} &= 0 \quad \text{for } 0 < x < l \text{ and } t > 0; \\ w(x, 0) &= 0; \text{ and} \\ w(0, t) &= 0, \quad w(l, t) = 0. \end{aligned}$$

Now consider

$$\left(\frac{1}{2} w^2 \right)_t - (Dw_x w)_x + Dw_x^2;$$

observe that this is zero and integrate with respect to x :

$$0 = \int_0^l \left(\frac{1}{2} w^2 \right)_t dx + D \int_0^l w_x^2 dx.$$

Note that the middle term vanishes due to the boundary conditions, and by the Leibniz integral rule

$$\frac{d}{dt} \int_0^l \frac{1}{2} w^2 dx = -D \int_0^l w_x^2 dx \leq 0.$$

This implies that the integral $\int_0^l w^2 dx$ is monotone decreasing in t , so that

$$0 \leq \int_0^l w^2 dx \leq \int_0^l w^2(x, 0) dx = 0.$$

Hence $w = 0$, concluding the proof. The method of proof employed here is a common one in proving uniqueness of solutions, called the *energy method*. \square

1.6. A (Weak) Maximum Principle.

Proposition 1.12. *Let $u(x, t)$ satisfy the diffusion equation in the rectangle $\mathcal{R} = [0, l] \times [0, T] = \{0 \leq x \leq l, 0 \leq t \leq T\}$. Then the maximum value of u on \mathcal{R} is assumed either on the initial line $t = 0$ or the boundary lines $x = 0, x = l$.*

Proof. Let M denote the maximum of u on the three boundary lines $t = 0, x = 0$, and $x = l$, and for $\epsilon > 0$, define

$$v(x, t) = u(x, t) + \epsilon x^2.$$

It is clear that

$$v(x, t) \leq M + \epsilon l^2$$

on the three boundary lines, and in addition we have the inequality

$$v_t - Dv_{xx} = u_t - Du_{xx} - 2\epsilon D = -2\epsilon D < 0. \quad (13)$$

If we assume that v achieves its maximum in the interior we know that $v_t = 0$ and $v_{xx} \leq 0$, a contradiction to (13). Now, again for the sake of contradiction, assume the maximum is attained at (x_0, T) where $0 < x_0 < l$. We still have that $v_{xx} \leq 0$, and

$$v_t(x_0, t_0) = \lim_{\delta \searrow 0} \frac{v(x_0, t_0) - v(x_0, t_0 - \delta)}{\delta} \geq 0,$$

another contradiction to (13). Hence

$$v(x, t) \leq M + \epsilon l^2$$

on \mathcal{R} , so that

$$u(x, t) \leq M + \epsilon(l^2 - x^2).$$

This concludes the proof. \square

Corollary 1.14. *The analogous result holds for the minimum of u .*

Proof. Apply the maximum principle to $-u(x, t)$. \square

1.7. Relation to Brownian Motion. Consider a 1-dimensional symmetric random walk with step length Δx , and suppose r steps are made per unit of time so that $\Delta t = \frac{1}{r}$ is the time of a single step.

Let $\xi_k \in \{-\Delta x, \Delta x\}$ be the step in the time interval $[(k-1)\Delta t, k\Delta t]$ for $k \in [n]$, and suppose $x_0 = 0$ is the initial point. Then after time $t = n\Delta t$ the position is given by the random variable

$$x_n = \sum_{k=1}^n \xi_k.$$

We define $u(x, t)$ to be the probability that the particle is at position x at time t , i.e.,

$$u(x, t) = P(x_n = x)$$

where $x = m\Delta x$ and $t = n\Delta t$ for $m \in \mathbb{Z}$, $n \in \mathbb{N}$. To find this probability, first let n_l and n_r denote the number of left steps and right steps respectively. Then we have $n_l + n_r = n$ and $n_r - n_l = m$, which together give

$$n_r = \frac{n + m}{2}.$$

The probability being given by the binomial distribution is then

$$P(x_n = m\Delta x) = \binom{n}{n_r} \left(\frac{1}{2}\right)^{n_r} \left(\frac{1}{2}\right)^{n-n_r} = \frac{1}{2^n} \binom{n}{\frac{n+m}{2}}.$$

The expectation is

$$\begin{aligned} E(x_n) &= E\left(\sum_{k=1}^n \xi_k\right) = \sum_{k=1}^n E(\xi_k) \\ &= \sum_{k=1}^n \left((- \Delta x)P(\xi_k = -\Delta x) + \Delta x P(\xi_k = \Delta x)\right) = 0, \end{aligned}$$

while the variance is

$$\begin{aligned} V(x_n) &= V\left(\sum_{k=1}^n \xi_k\right) = \sum_{k=1}^n V(\xi_k) \\ &= \sum_{k=1}^n \left((-\Delta x)^2 P(\xi_k = -\Delta x) + (\Delta x)^2 P(\xi_k = \Delta x) \right) = (\Delta x)^2 n. \end{aligned}$$

Rewriting as $V(x_n) = (\Delta x)^2 r t$ we note that $\sigma(x_n) = \sqrt{V(x_n)} \sim \sqrt{t}$, which has the physical interpretation that while the particle's expected position is its starting point due to symmetry, for growing t this probability decreases (slower than linearly).

To obtain Brownian motion, we let $\Delta x \rightarrow 0$ and $r \rightarrow \infty$ such that

$$\lim_{\Delta x \rightarrow 0, r \rightarrow \infty} (\Delta x)^2 r = 2D > 0.$$

Now our discrete formulation of the problem gives the following equation:

$$u(x, t + \Delta t) = \frac{1}{2}u(x - \Delta x, t) + \frac{1}{2}u(x + \Delta x, t).$$

If we consider the Taylor expansions

$$u(x, t + \Delta t) = u(x, t) + u_t \Delta t + \frac{1}{2}u_{tt} \Delta t^2 + \dots$$

and

$$u(x \pm \Delta x, t) = u(x, t) \pm u_x \Delta x + \frac{1}{2}u_{xx} \Delta x^2 \pm \dots,$$

we obtain

$$u_t \Delta t + \frac{1}{2}u_{tt} \Delta t^2 + \dots = \frac{1}{2}u_{xx} \Delta x^2 + \frac{1}{4}u_{xxx} \Delta x^4 + \dots.$$

Then dividing through by Δt and passing to the limit we get

$$u_t = Du_{xx},$$

which is precisely the diffusion equation! For our random walk we have initial condition $u(x, 0) = \delta(x)$, and so our fundamental solution is

$$u(x, t) = \frac{1}{\sqrt{4\pi Dt}} e^{-x^2/4\pi Dt}.$$

Remark 1.15. Of course, the assumption that

$$\frac{(\Delta x)^2}{2\Delta t}$$

tends to a finite limit implies that the velocity $(\Delta x)/(\Delta t)$ approaches infinity, which is not physically realizable. One can also observe this from the derived Gaussian (normal) solution (with mean 0 and variance $2Dt$). In particular, for all $x \in \mathbb{R}$ and $t > 0$ we have $u(x, t) > 0$, meaning there is a positive probability that the particle will be located arbitrarily far away from the initial point immediately. Hence the model is really only valid for large t .

1.8. Diffusion Variance.

Proposition 1.16. *Let u solve*

$$u_t = Du_{xx}$$

and suppose that

$$C = \int_{\mathbb{R}} u(x, t) dx$$

is independent of t and

$$\lim_{x \rightarrow \pm\infty} xu(x, t) = 0 \quad \text{and} \quad \lim_{x \rightarrow \pm\infty} x^2 u_x(x, t) = 0.$$

Intuitively, this means that we have a constant population and the density is ‘small at infinity’. Let

$$\sigma^2(t) = \frac{1}{C} \int_{\mathbb{R}} x^2 u(x, t) dx.$$

Then

$$\sigma^2(t) = 2Dt + \sigma^2(0)$$

for all t .

Proof. We apply integration by parts:

$$\begin{aligned} \frac{C}{D} \frac{d\sigma^2}{dt} &= \frac{1}{D} \frac{d}{dt} \int_{\mathbb{R}} x^2 u dx \\ &= \frac{1}{D} \int_{\mathbb{R}} x^2 u_t dx \\ &= \int_{\mathbb{R}} x^2 u_{xx} dx \\ &= \left[x^2 u_x \right]_{-\infty}^{\infty} - \int_{\mathbb{R}} 2xu_x dx \\ &= -[2xu]_{-\infty}^{\infty} + \int_{\mathbb{R}} 2u dx \\ &= 2 \int_{\mathbb{R}} u dx = 2C. \end{aligned}$$

Hence

$$\frac{d\sigma^2(t)}{dt} = 2D$$

and integration gives

$$\sigma^2(t) = 2Dt + \sigma^2(0).$$

In the special case that the particles are initially concentrated around $x = 0$ (so that $u(x, 0) = 0$ for all $|x| > \epsilon$), we obtain

$$\int_{\mathbb{R}} x^2 u(x, 0) dx = \int_{-\epsilon}^{\epsilon} x^2 u(x, 0) dx \leq \epsilon^2 \int_{-\epsilon}^{\epsilon} u(x, 0) dx = \epsilon^2 C.$$

Thus

$$\sigma^2(0) = \epsilon^2 \approx 0.$$

□

2. EXISTENCE AND UNIQUENESS

2.1. **Maximum and Comparison Principles.** Consider the following problem:

$$u_t = D\Delta u + f(x, t, u) \text{ on } \Omega \times (0, T)$$

$$u = 0 \text{ on } \partial\Omega \times (0, T)$$

$$u = u_0 \text{ on } \Omega \times \{0\}$$

where Ω is a bounded domain.

REFERENCES

- [1] Christina Kuttler. http://www-m6.ma.tum.de/~kuttler/script_reaktdiff.pdf