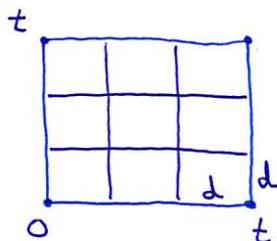


| E5.5.3 | Let N_A be the number of defects found by inspector A and N_B the number found by inspector B. We have

$$\begin{aligned} \Pr(N_A \geq 1 \text{ AND } N_B \geq 1) &= \Pr(N_A \geq 1) \Pr(N_B \geq 1) \\ &= [1 - \Pr(N_A = 0)][1 - \Pr(N_B = 0)] \\ &= [1 - (\frac{1}{2})^0 e^{-\frac{1}{2}}]^2 \\ &= [1 - e^{-\frac{1}{2}}]^2. \end{aligned}$$

| P5.5.3 |



$$\begin{aligned} \text{box side length} &= d = \frac{t}{n} \\ \text{number of boxes} &= n^2 \end{aligned}$$

Let R denote the number of reactions. Then R takes values in $\{0, \dots, n^2\}$.

The distribution of R is binomial:

$$\Pr(R = k) = \binom{n^2}{k} \left[\sum_{i=0}^{\infty} \frac{(\lambda d^2)^i}{i!} e^{-\lambda d^2} \right]^k \left[\sum_{j=0}^{\infty} \frac{(\lambda d^2)^j}{j!} e^{-\lambda d^2} \right]^{n^2-k}.$$

Let's try to apply the law of rare events.

$$\textcircled{1} \quad (\# \text{ successes expected}) = (\# \text{ trials})(\text{Prob. of success})$$

$$\text{(A) \# trials} = n^2 = \frac{t^2}{d^2}$$

$$\begin{aligned} \text{(B) Prob. of success} &= e^{-\lambda d^2} \sum_{i=0}^{\infty} \frac{(\lambda d^2)^i}{i!} \\ &= \left[1 - \lambda d^2 + \frac{\lambda^2 d^4}{2} - \dots \right] \left[\frac{\lambda^2 d^4}{2} + \dots \right]. \end{aligned}$$

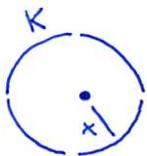
To leading order in d , Prob. of success $\approx \frac{\lambda^2 d^4}{2}$.

$$\textcircled{2} \quad \text{So } (\# \text{ trials})(\text{Prob. of success}) \approx \left(\frac{t^2}{d^2} \right) \left(\frac{\lambda^2 d^4}{2} \right) = \frac{\lambda^2 [dt]^2}{2}$$

and this converges to $\frac{(\lambda \mu)^2}{2}$.

By the law of rare events, the distribution of R is Poisson with mean $(\frac{\lambda \mu}{2})^2$ in the $t \rightarrow \infty; d \rightarrow 0; t d \rightarrow \mu > 0$ limit.

| P5.5.5 | Let $F_D(x)$ denote the CDF for the distance between a particle and its nearest neighbor.



$$\begin{aligned} \text{We have } F_D(x) &= \Pr(D \leq x) \\ &= 1 - \Pr(D > x) \\ &= 1 - \Pr(\#\text{ particles in } K = 0) \\ &= 1 - e^{-\nu \pi x^2}. \end{aligned}$$

To compute the mean distance, we use the tail probability integral

$$\begin{aligned} &\int_0^\infty 1 - F_D(x) dx \\ &= \int_0^\infty e^{-\nu \pi x^2} dx \quad \text{substitution: } \frac{z^2}{2} = \nu \pi x^2 \\ &= \frac{1}{\sqrt{2\nu\pi}} \int_0^\infty e^{-z^2/2} dz \quad \text{or } z = (\sqrt{2\nu\pi})x \\ &= \frac{1}{\sqrt{2\nu}} \left[\frac{1}{\sqrt{2\pi}} \int_0^\infty e^{-z^2/2} dz \right] = \frac{1}{2\sqrt{\nu}}. \end{aligned}$$

| P5.5.6 | Arguing as in Problem 5.5.5,

$$\begin{aligned} F_R(x) &= \Pr(R \leq x) \\ &= 1 - e^{-\lambda [\frac{4}{3}\pi x^3]} \end{aligned}$$

The PDF $f_R(x)$ is given by $f_R(x) = 4\lambda\pi x^2 e^{-\lambda [\frac{4}{3}\pi x^3]}$.

| E8.1.3 | (a) Let $\phi(x) = \frac{1}{\sqrt{2t\pi}} e^{-x^2/2t}$. The chain rule implies $\phi'(x) = -x\phi(x)$.

$$(b) \frac{\partial^2 p}{\partial x^2} = \frac{1}{t^{3/2}} \phi\left(\frac{y-x}{\sqrt{t}}\right) \left[\frac{(y-x)^2}{t} - 1 \right]$$

$$\frac{\partial p}{\partial t} = \left(\frac{1}{2t^{3/2}} \right) \phi\left(\frac{y-x}{\sqrt{t}}\right) \left[\frac{(y-x)^2}{t} - 1 \right]$$

$$\text{Thus } \frac{\partial p}{\partial t} = \frac{1}{2} \frac{\partial^2 p}{\partial x^2}.$$

| E8.1.4(a) | A simple observation: $-B(t)$ is a standard Brownian motion as well.

$$\text{Thus } E[B(u)B(u+v)B(u+v+w)]$$

$$= E[-B(u)(-B(u+v))(-B(u+v+w))]$$

$$= -E[B(u)B(u+v)B(u+v+w)],$$

$$\text{forcing } E[B(u)B(u+v)B(u+v+w)] = 0.$$

| E8.1.5(a) | Computing the covariance for s, t , we have

$$\begin{aligned} \text{Cov}[U(s)U(t)] &= E[U(s)U(t)] - E[U(s)]E[U(t)] \\ &= e^{-(s+t)} E[B(e^{2s})B(e^{2t})] \\ &= e^{-(s+t)} \min\{e^{2s}, e^{2t}\} = e^{-|s-t|}. \end{aligned}$$