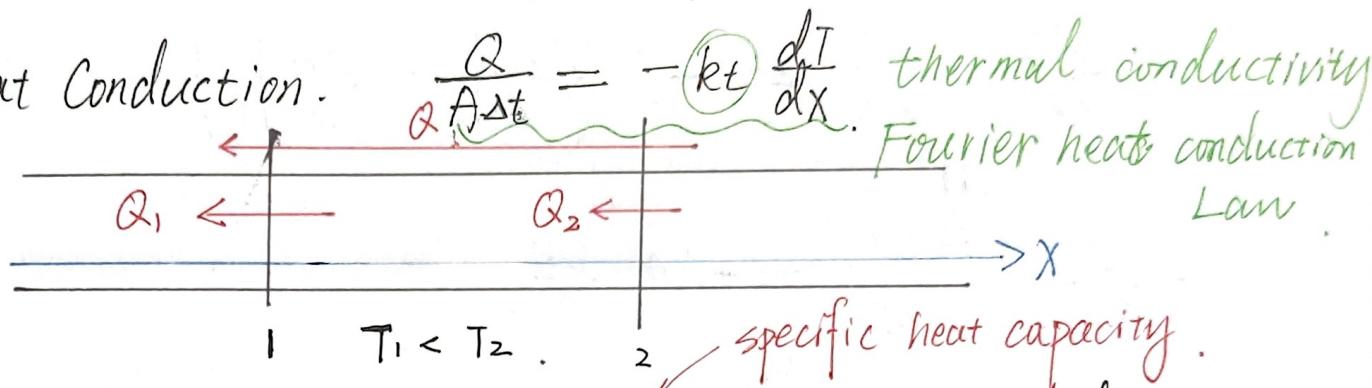


transport of Energy : Heat Conduction

transport of Momentum : Viscosity

transport of Particles : Diffusion

1. Heat Conduction.

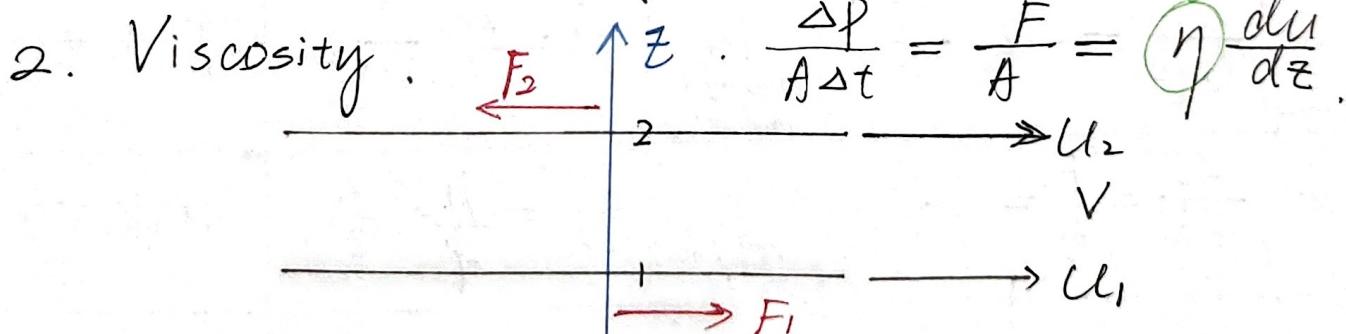


$$Q = Q_2 - Q_1 ; Q = -C_p m dT = -c_p A \Delta x dT.$$

$$\frac{\partial Q}{\partial t} = \frac{\partial Q_2}{\partial t} - \frac{\partial Q_1}{\partial t} = - \left(A k_t \frac{\partial T}{\partial x} \Big|_2 - A k_t \frac{\partial T}{\partial x} \Big|_1 \right)$$

$$-c_p A \Delta x \frac{\partial T}{\partial t} = -A k_t \left(\frac{\partial T}{\partial x} \Big|_2 - \frac{\partial T}{\partial x} \Big|_1 \right)$$

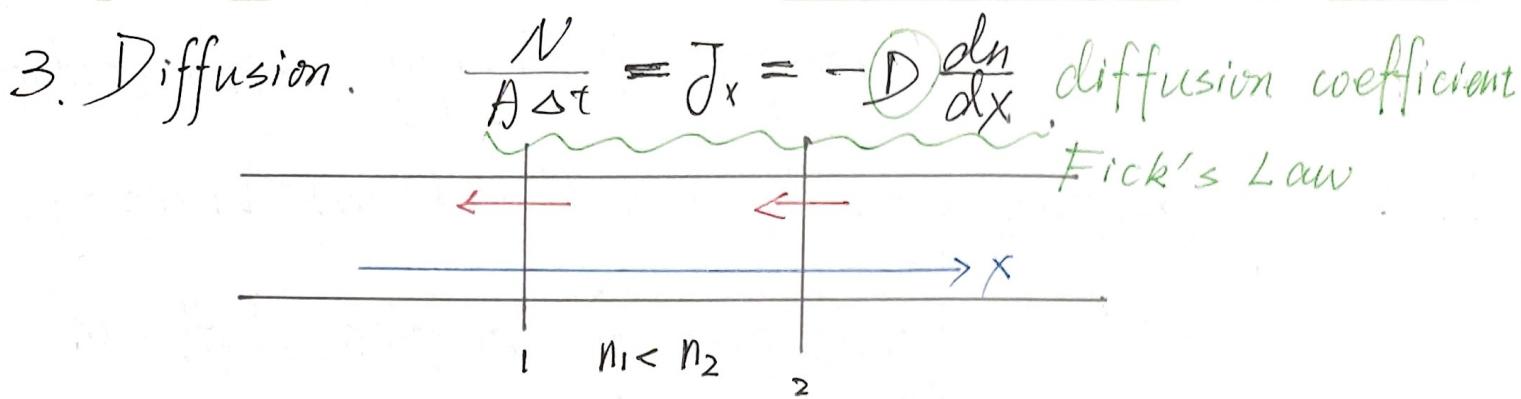
Heat equation . $\frac{\partial T}{\partial t} = \frac{k_t}{c_p} \frac{\partial^2 T}{\partial x^2} = K \frac{\partial^2 T}{\partial x^2}$ coefficient of viscosity



$$\frac{\Delta P}{A \Delta t} \Big|_1^2 = \frac{F_2}{A} - \frac{F_1}{A} = \eta \left(\frac{\partial u}{\partial z} \Big|_2 - \frac{\partial u}{\partial z} \Big|_1 \right)$$

$$\frac{m}{A} \frac{\partial u}{\partial t} = f A \frac{\Delta z}{A} \frac{\partial u}{\partial t} = f \Delta z \frac{\partial u}{\partial t}$$

$$\frac{\partial u}{\partial t} = \frac{\eta}{f} \frac{\partial^2 u}{\partial z^2}$$



$$\frac{N_2}{A\Delta t} - \frac{N_1}{A\Delta t} = -D \left(\frac{\partial n}{\partial x} \Big|_2 - \frac{\partial n}{\partial x} \Big|_1 \right)$$

$$-\frac{\partial}{\partial t} \frac{n A \Delta x}{A \cancel{\Delta t}} = -\Delta x \cdot \frac{\partial n}{\partial t}$$

$$\frac{\partial n}{\partial t} = D \frac{\partial^2 n}{\partial x^2} . \text{ Fick's second law}$$

4. Conclusion.

Transport of	Equation (I). for surfaces	Constant	Equation (II). for functions	Function
Energy <u>Heat conduction</u>	$\frac{Q}{A\Delta t} = -k_t \frac{\partial T}{\partial x}$	k_t Thermal Conductivity	$\frac{\partial T}{\partial t} = \frac{k_t}{c_p} \frac{\partial^2 T}{\partial x^2}$	$T(x, t)$
Momentum <u>Viscosity</u>	$\frac{\Delta P}{A\Delta t} = \frac{ F_x }{A} = \eta \frac{du_x}{dz}$	η coefficient of viscosity	$\frac{\partial u_x}{\partial t} = \eta \frac{\partial^2 u_x}{\partial z^2}$	$u_x(z, t)$
Particles Diffusion	$\frac{N}{A\Delta t} = J_x = -D \frac{dn}{dx}$	D Diffusion coefficient	$\frac{\partial n}{\partial t} = D \frac{\partial^2 n}{\partial x^2}$	$n(x, t)$

Who Determines k_t , η , D . ~ Analysis By Kinetic Theory

* For Ideal Gas.

1. k_t

$$\frac{Q}{A\Delta t} = -k_t \frac{dT}{dx}$$

$$Q = \left(\frac{1}{2}\right) \text{ probability } 50\% \cdot \left(\frac{1}{2d}\right) l \quad d = \text{dimension} = \left(\frac{1}{2}\right)^3$$

$$Q = \frac{1}{2} (U_{1 \rightarrow 2} - U_{2 \rightarrow 1}) = -\frac{1}{2} \left(\frac{fNk}{2} T_2 - \frac{fNk}{2} T_1 \right)$$

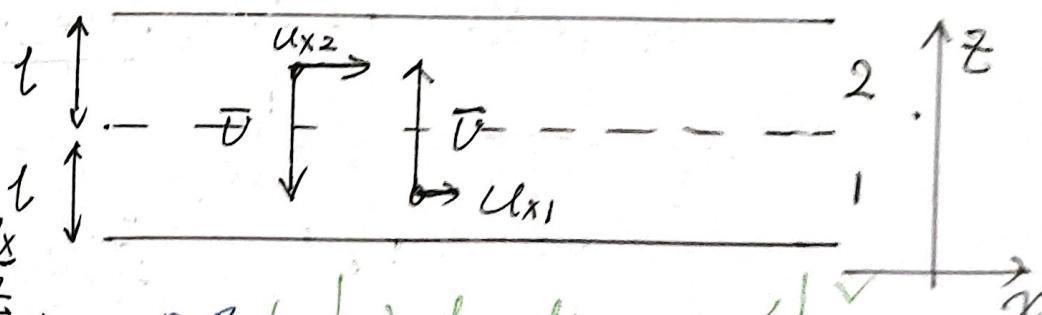
$$\frac{Q}{A\Delta t} = -\frac{\frac{fNk}{2} l}{4A\Delta t} \frac{T_2 - T_1}{l} = -k_t \frac{dT}{dx}$$

$$\therefore k_t = \frac{fNk l}{4A\Delta t} = \frac{fk N l}{4 A l} \frac{l}{\Delta t} = \frac{fk}{4} \frac{N}{V} \cancel{E} \cancel{V}$$

$$l = \frac{V}{4\pi r^2 N} \quad \bar{v} = \sqrt{\frac{fkT}{m}}$$

$$\therefore k_t = \frac{fk}{4} \cdot \frac{1}{4\pi r^2} \cdot \sqrt{\frac{fk}{m}} \cdot \sqrt{T} \Rightarrow k_t \propto \sqrt{T}$$

2. η .



$$\left| \frac{\Delta P}{A\Delta t} \right| = \frac{|F_x|}{A} = \eta \frac{du_x}{dz}$$

$$\left| \Delta P \right| = \left(\frac{1}{2} \right) \text{ probability } 50\% \left(\frac{1}{2d} \right) d = \text{dimension} \left(\frac{1}{2} \right)^3$$

$$\left| \frac{\Delta P}{A\Delta t} \right| = \frac{mNl}{2A\Delta t} \frac{u_{x2} - u_{x1}}{l} = \eta \frac{du_x}{dz}$$

$$\therefore \eta = \frac{mNl}{2A\Delta t} = \frac{mNl}{2Al} \cdot \frac{l}{\Delta t} = \frac{m}{2} \cancel{N} \cancel{V} \cancel{E} \cancel{V}$$

$$\eta = \frac{m}{2} \frac{1}{4\pi r^2} \sqrt{\frac{fk}{m}} \cdot \sqrt{T} \Rightarrow \eta \propto \sqrt{T}$$

3. D.

$$\frac{N}{A\Delta t} = \bar{J}_x = -D \frac{dn}{dx}$$

$$N = \frac{1}{2}(n_1 A l - n_2 A l) \quad \text{probability } 50\% \quad \left(\frac{1}{2}d\right)$$

$d = \text{dimension} = \sqrt{\frac{l}{3}}$

$$\frac{N}{A\Delta t} = -\frac{1}{2} \frac{A l l}{A\Delta t} \frac{n_2 - n_1}{l} = -D \frac{dn}{dx}$$

$$\therefore D = \frac{1}{2} l \frac{d}{\Delta t} = \frac{1}{2} (l \bar{v}) \quad , \quad l = \frac{V}{4\pi r^2 N} \quad \bar{v} = \sqrt{\frac{fkT}{m}}$$

$$D = \frac{1}{2} \frac{V}{4\pi r^2 N} \sqrt{\frac{fk}{m}} \cdot \bar{v} \quad \text{However. } PV = NkT$$

$$\therefore \frac{N}{V} = \frac{kT}{P} \quad \therefore D = \frac{1}{2} \cdot \frac{1}{4\pi r^2} \sqrt{\frac{fk}{m}} \cdot k \cdot \frac{(T)^{\frac{3}{2}}}{P} \propto \frac{(T)^{\frac{3}{2}}}{P}$$

4. Conclusions.

(1) Expressions of $k_t \cdot \eta \cdot D$.

Macroscopic: $\frac{Q}{A\Delta t} = -k_t \frac{dT}{dx} \quad , \quad k_t = \frac{fk(N)}{4} \cdot (\bar{v}) \cdot (l)$

$$\frac{F_x}{A} = \frac{\Delta P}{A\Delta t} = \eta \frac{du_x}{dz} \quad , \quad \eta = \frac{m}{2} \left(\frac{N}{V} \right) (\bar{v}) (l)$$

$$\bar{J}_x = \frac{N}{A\Delta t} = -D \frac{dn}{dx} \quad , \quad D = \left[\frac{1}{2} \frac{l^2}{\Delta t} \right] = \frac{1}{2} (l \bar{v})$$

Microscopic

(2) Gas: $k_t \propto \bar{v} T$, $\eta \propto \bar{v} T$, $D \propto (T)^{\frac{3}{2}}$

$$T \uparrow \Rightarrow k_t \uparrow \quad T \uparrow \Rightarrow \eta \uparrow \quad T \uparrow \Rightarrow D \uparrow$$

Liquid & Solid:

$$T \uparrow \Rightarrow \eta \downarrow$$

more likely to transfer
 $T \uparrow \Rightarrow \uparrow$
 $T \uparrow \Rightarrow \text{interactions} \uparrow$

Caused by interactions between molecules.

关于扩散系数.

定义一: $\vec{J} = -D \nabla \cdot n(x, y, z)$; $\left\{ \begin{array}{l} J_x = -D \frac{\partial n}{\partial x} \\ J_y = -D \frac{\partial n}{\partial y} \\ J_z = -D \frac{\partial n}{\partial z} \end{array} \right.$

定义二: $\langle \vec{r}^2 \rangle = 2Dt \times d = 6Dt$ ($d=3$)

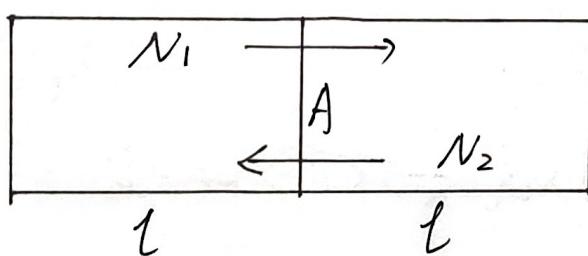
证明定义一与定义二等价的思路:

$$\text{定义一} \iff D = \frac{1}{2d} \frac{l^2}{\Delta t} \Leftrightarrow l^2 = 2D\Delta t \times d$$

$$\text{定义二} \iff \langle \vec{r}^2 \rangle = 2Dt \times d$$

(一) 推导 D 的微观表达式. $l^2 = 2D\Delta t \times d$.

借鉴方法. { 研究对象选分界面两侧厚度为 l 的物体 (参考 k_t)
从一维推广到 d 维. (参考理想气体 P). }



$$\begin{aligned} N_1 &= n_1 \cdot A l \\ N_2 &= n_2 \cdot A l \\ J_x &= \frac{1}{2} \left(\frac{N_1 - N_2}{A \Delta t} \right) \end{aligned}$$

(1) 一维.

$$J_x = \frac{1}{2} \left(\frac{N_1 - N_2}{A \Delta t} \right) = -\frac{1}{2} \frac{A l (n_2 - n_1)}{A \Delta t} = -\frac{1}{2} \frac{l^2}{\Delta t} \frac{dn}{dx}$$

(2) 一维推广到 d 维.

① 将 l 代换为 l_x (l 在 x 轴方向上分量).

$$l^2 = l_x^2 + l_y^2 + l_z^2; l_x^2 = \frac{1}{d} l^2$$

② 将概率/占比 $\frac{1}{2}$ 代换为 $\frac{1}{2d}$. (d 个维度, $2d$ 个方向)

$$J_x = -\frac{1}{2d} \frac{l^2}{\Delta t} \frac{dn}{dx} = -\frac{1}{2} \frac{l^2}{d \Delta t} \frac{dn}{dx}$$

$$D = \frac{1}{2} \frac{l^2}{d \Delta t} \Leftrightarrow l^2 = 2D \Delta t \times d$$

(二) 由 $\ell^2 = 2D\Delta t \times d$, 据 $\langle \vec{r}^2 \rangle = 2Dt \times d$,

取 $t = \lambda \cdot \Delta t$ 若 $\langle \vec{r}^2 \rangle = \lambda \ell^2$ 则可以推出
所以, 即证 $\langle \vec{r}^2 \rangle = \lambda \ell^2$.

参考 QMB 笔记, One Dimensional Random Walk.

对于一维有 $\langle \vec{r}_x^2 \rangle = \lambda_x \ell^2 = \lambda \ell_x^2$. 从一维到 d 维 / 三维

$$\vec{r}^2 = \vec{r}_x^2 + \vec{r}_y^2 + \vec{r}_z^2$$

另法:

$$\langle \vec{r}^2 \rangle = \langle \vec{r}_x^2 \rangle + \langle \vec{r}_y^2 \rangle + \langle \vec{r}_z^2 \rangle = \lambda \ell^2 \quad \langle \vec{r}_x^2 \rangle = \left(\sum_i \sum_j \vec{r}_{xi} \cdot \vec{r}_{xj} \right)$$

$$\left\{ \begin{array}{l} (1) \langle \vec{r}_x^2 \rangle = \lambda_x \ell^2 = \left(\frac{\lambda}{3}\right) \ell^2 = \frac{1}{3} \lambda \ell^2 = \sum_i \sum_j \langle \vec{r}_{xi} \cdot \vec{r}_{xj} \rangle \\ (2) \langle \vec{r}_x^2 \rangle = \lambda \ell_x^2 = \lambda \left(\frac{\ell^2}{3}\right) = \frac{1}{3} \lambda \ell^2 \end{array} \right. \text{其中 } \langle \vec{r}_{xi} \cdot \vec{r}_{xj} \rangle = \begin{cases} \ell^2 & (i=j), (i \neq j \text{ 独立}) \\ \langle \vec{r}_x^2 \rangle & \end{cases}$$

$$\text{从而 } \langle \vec{r}^2 \rangle = \lambda \ell^2.$$

$$\text{进而, } \ell^2 = 2D\Delta t \times d.$$

$$\lambda \ell^2 = 2D(\lambda \Delta t) \times d \Leftrightarrow \langle \vec{r}^2 \rangle = 2Dt \times d.$$

(三) D 的双重含义.

1. 由定义一; $\vec{J} = -D \nabla \cdot n(x, y, z) = -D \left(\frac{\partial n}{\partial x}, \frac{\partial n}{\partial y}, \frac{\partial n}{\partial z} \right)$.

Flux \vec{J} 沿 $n(x, y, z)$ 负梯度方向. (n 减少最快方向).

$|\vec{J}| = \frac{\Delta n}{\Delta t}$, D 为其中比例系数. n 浓度, n 宏观一群粒子.

2. 由定义二; $\langle \vec{r}^2 \rangle = 2D\Delta t \cdot d$. \vec{r} 位移 ~ 微观单个分子位移平方的期望 正比于时间. D 为比例系数的一部分.

Random Walks 中的维数

$$\langle X^2 \rangle - \langle X \rangle^2 = D(X) = \sigma^2 \cdot \text{方差}$$

由于 $\langle \vec{X} \rangle = \vec{0}$ 所以 $\sigma^2 = \langle X^2 \rangle$

当 \vec{X} 为一维随机变量: $\langle \vec{X}^2 \rangle = 2Dt = \sigma^2$

为二维: $\langle \vec{X}^2 \rangle = 4Dt = \sigma^2$

为三维: $\langle \vec{X}^2 \rangle = 6Dt = \sigma^2$

$$\vec{r} = (x, y, z) = (r, \theta, \varphi). \quad \text{三维}$$

$$\vec{r} = (x, y) = (r, \theta) \quad \text{投影到二维}$$

$\langle \vec{r}^2 \rangle$ 是 二维随机变量的均方: 所以 $\langle \vec{r}^2 \rangle = 4Dt$

对 \vec{r} 取模 $| \vec{r} | = r$ 是 一维随机变量, 方差为 $\sigma^2(r) = 2Dt$

正态分布 $P(r) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{r^2}{2\sigma^2}}$ 当 $\sigma^2 = 2Dt$
 $(\mu=0)$

$$P(r) = \frac{1}{\sqrt{4\pi Dt}} \cdot e^{-\frac{r^2}{4Dt}}$$

t 可以取 Δt . 对应 r 取 $\Delta |\vec{r}| \neq |\Delta \vec{r}|$
一维随机变量

解析. $\langle |\vec{r}|^2 \rangle$ = 一维随机变量 $|\vec{r}|^2$ 的期望

= $\sigma^2(r) =$ 一维随机变量 r 的方差

$\Delta |\vec{r}|$ 是 \vec{r} 向径向的投影. (d 维 \rightarrow 1 维) 的变化量 \sim 维数

从而其方差 $\sigma^2(\Delta |\vec{r}|) = 2D\Delta t$

* $\sigma^2(\Delta |\vec{r}|^2) = 2D\Delta t$; $\sigma^2(\vec{r}) = 2Dt \times d$.

* 一维变量直方图: 曲线拟合 (平面)

二维变量直方图: 曲面拟合 (空间)

* \vec{r} 为 d 维向量 \sim d 维 Random Walk.

$$\sigma^2(\vec{r}) = \langle \vec{r}^2 \rangle = (2Dt) \times d$$

\vec{r} 投影到径向 (r 轴) \rightarrow d 维 \rightarrow 1 维 Random Walk

$|\vec{r}|$. 再取其在 Δt 内的变化量 $|\Delta \vec{r}|$

$$\underline{\sigma^2(|\Delta \vec{r}|)} = 2D \Delta t$$

求均方 $\langle \vec{r}^2 \rangle$; 求 $\langle \vec{r}^2 \rangle - t$ 斜率.

$$D = \frac{k_B T}{6\pi \eta r} \quad \text{热学角度} \sim D \sim \text{统计角度}$$

$$D = \mu k_B T \quad \text{扩散系数}$$

$$\left. \begin{array}{l} \text{均方: } \langle \vec{r}^2 \rangle = (2Dt) \times d \\ \text{方差: } \sigma^2(\Delta \vec{r}) = 2D\Delta t \end{array} \right\}$$

求 $|\Delta \vec{r}|$ 的方差