

Entropy

- thermodynamic entropy 1 mole of ideal gas

$$dU = p dv - \theta dS$$

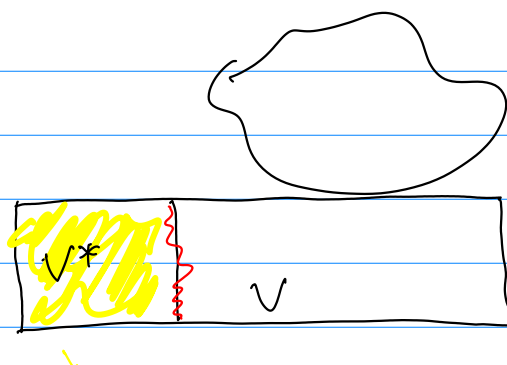
$$pV = R\theta$$

expand isothermally, quasi-statically no change in U

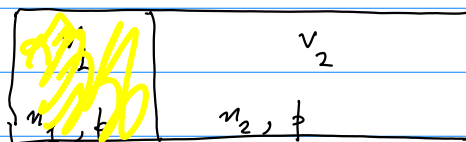
$$0 = \int_{V^*}^V dU = \int_{V^*}^V p dV - \theta \int_0^{\Delta S} dS$$

$$= \int_{V^*}^V R\theta \frac{dV}{V} - \theta \int_0^{\Delta S} dS$$

$$\Delta S = R \log \frac{V}{V^*} = -R \log \frac{V^*}{V}$$

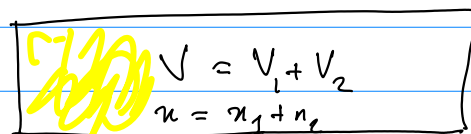


2. Entropy of Mixing



$$U_1 = pV_1 n_1 \quad U_2 = pV_2 n_2$$

$$pV_1 = n_1 R\theta \quad pV_2 = n_2 R\theta$$



p_1 = partial pressure of Gas 1

$$p_{\text{new}} = p_1 + p_2$$

p_2 = partial pressure of Gas 2

$$p_{\text{new}} = p_1 + p_2 = \frac{n_1 R\theta}{V} + \frac{n_2 R\theta}{V}$$

$$= \frac{R\theta}{V} (n_1 + n_2)$$

$$= \frac{R\theta}{V} \left(\frac{pV_1}{R\theta} + \frac{pV_2}{R\theta} \right)$$

$$= p$$

$$p_{\text{new}} = p$$

Dalton's Law
(extra thermodynamical hypothesis)

$$\Delta S_1 = -n_1 R \log \frac{V_1}{V} \quad \Delta S_2 = -n_2 R \log \frac{V_2}{V}$$

$$\Delta S = -R \left(n_1 \log \frac{n_1}{n_1+n_2} + n_2 \log \frac{n_2}{n_1+n_2} \right) \quad \text{entropy of mixing}$$

proportional to configurational entropy with probabilities

$$p_1 = \frac{n_1}{n}, \quad p_2 = \frac{n_2}{n} \quad n = n_1 + n_2$$

Configurational entropy

(change the sign)

negative of Shannon entropy

$$\Phi(p) = \sum_{k=1, \dots, r} p_k \log p_k \quad \sum p_k = 1, \quad p_k \geq 0 \quad 0 \cdot \log 0 = 0$$

$$X = \{x_1, \dots, x_r\} \quad \text{Prob}(w_i = x_k) = p_k, \quad k = 1, \dots, r$$

$$\log \frac{1}{r} \leq \Phi(p) \leq 0$$

minimum of Φ : $p_k = \frac{1}{r}$ uniform distribution $\Phi = \log \frac{1}{r}$
"most random"

maximum of Φ : $p_k = 1$ for one k , $p_n = 0$ otherwise $\Phi = 0$
least random
 k occurs and n others

Suppose p^* , p dependent

successive distributions of a Markov chain

$$P = (p_{ij}^*)$$

$$p = p^* P$$

$$p(x_i, x_j) = p_i^* p_{ij}^*$$

$$\Phi(p(x_i, x_j)) = \sum p(x_i, x_j) \log p(x_i, x_j)$$

$$= \sum p_i^* p_{ij}^* \log p_i^* p_{ij}^*$$

$$= \sum p_i^* p_{ij}^* \log p_i^* + \sum p_i^* p_{ij}^* \log p_{ij}^*$$

$$= \Phi(p^*) + \sum p_i^* \Phi_i(P)$$

entropy of i^{th} row of P

$$H(x_i) = \sum p(x_i, y_j) \log p(x_i, y_j)$$

$$p_i \log p_j = \sum_i p_i^* p_{ij} \log \left(\sum_i p_i^* p_{ij} \right) \stackrel{\text{Jensen}}{\geq} \sum p_i^* p_{ij} \log p_{ij}$$

sum n_j

Jensen

$$H(p) \geq \sum p_i^* H_i(P)$$

$$H(p(x, y)) \geq H(p^*) + H(p)$$

conditioning reduces uncertainty !