

# Practice Problem

Example A random permutation has  
1 fixed point on the average.

Pf: Let  $X_i(\sigma) = \begin{cases} 1 & \text{if } \sigma_i = i \\ 0 & \text{o.w.} \end{cases}$ , where  $\sigma$  is a

random permutation on  $\{1, 2, \dots, n\}$ .

Let  $X$  = the # of loops in  $\sigma$ .

$$EX = \sum_{i=1}^n E[X_i] = \sum_{i=1}^n \frac{(n-1)!}{n!} = 1$$

QED

# Hamiltonian Paths

Thm the first use of the probabilistic method  
(Szele 1943)  $\exists$  a tournament on  $n$  vertices that has at least  $\frac{n!}{2^{n-1}}$  Hamiltonian dipaths.

pf:  $X \stackrel{\text{def}}{=} \text{the } *$  of Hamiltonian dipaths.

$X_\sigma \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } \sigma \text{ is a Hamiltonian dipath} \\ 0 & \text{o.w.} \end{cases}$

$$\begin{aligned} EX &= E \sum_{\sigma \in S_n} X_\sigma \\ &= \sum_{\sigma \in S_n} P_r \{ \sigma_1 \prec_2 \dots \prec_n \text{ is a Ham. dipath} \} \\ &= n! \cdot \frac{2^{\binom{n}{2} - (n-1)}}{2^{\binom{n}{2}}} = \frac{n!}{2^{n-1}} \end{aligned}$$

QED

## Two Quickies

Thm  $\exists$  a  $\overset{\text{edge}}{2}$ -coloring of  $K_n$  with at most  $\binom{n}{a} 2^{\binom{a}{2}}$  monochromatic  $K_a$ .

Pf:  $X_s \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } s \text{ is monochromatic} \\ 0 & \text{o.w.} \end{cases}$

$$\mathbb{E} \sum_{s \in \binom{n}{a}} X_s = \sum_{s \in \binom{n}{a}} P(X_s=1) = \binom{n}{a} 2 \cdot 2^{-\binom{a}{2}}.$$

QED

Thm  $\exists$  a  $\overset{\text{edge}}{2}$ -coloring of  $K_{m,n}$  with at most  $\binom{m}{a} \binom{n}{b} 2^{1-ab}$  monochromatic  $K_{a,b}$ .

Pf:  $X_{S,T} \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } [S,T] \text{ is monochromatic} \\ 0 & \text{o.w.} \end{cases}$

$$\mathbb{E} \sum_{S \in \binom{m}{a}, T \in \binom{n}{b}} X_{S,T} = \binom{m}{a} \binom{n}{b} 2 \cdot 2^{-ab}.$$

QED

# Splitting Graphs I.

Thm

Any graph with  $m$  edges contains a bipartite subgraph with at least  $\frac{m}{2}$  edges.

~~pf:~~ Let  $S$  be a random vertex subset.

$$\text{let } X_{uv} = \begin{cases} 1 & \text{if } |\{u,v\} \cap S| = 1 \\ 0 & \text{o.w.} \end{cases}$$

$$\begin{aligned} \mathbb{E} \sum_{uv \in E(G)} X_{uv} &= \sum_{uv \in E(G)} \mathbb{E} X_{uv} = \sum_{uv \in E(G)} 1 - \Pr\{\{u,v\} \subseteq S \text{ or } \{u,v\} \cap S = \emptyset\} \\ &= \frac{m}{2} \end{aligned}$$

QED

# Splitting Graphs II

Thm If  $G$  has  $\frac{2n}{2n+1}$  vertices and  $e$  edges then it contains a bipartite subgraph with at least  $\frac{en}{(2n-1)}$  edges.

$$\frac{e(n+1)}{(2n+1)}$$

Pf: Let  $\Omega = \begin{bmatrix} 2n \\ n \end{bmatrix}$  with  $P(\omega) = 1/|\Omega|$ .

Let  $X_{xy}^{(\omega)} = \begin{cases} 1 & \text{if } |\{x, y\} \cap \omega| = 1, \text{ where } xy \in E(G). \\ 0 & \text{o.w.} \end{cases}$

$$E \sum_{x,y} X_{xy} = e P(\{\omega : |\{x, y\} \cap \omega| = 1\}) = e \frac{\binom{2n}{n} - \binom{2n-2}{n} - \binom{2n-2}{n-2}}{\binom{2n}{n}} = \frac{en}{2n-1}.$$

where  $xy \in E(G)$

Note: If  $\Omega = \begin{bmatrix} 2n+1 \\ n \end{bmatrix}$  with  $P(\omega) = 1/|\Omega|$  then

$$P(\{\omega : |\{x, y\} \cap \omega| = 1\}) = \frac{n+1}{2n+1}.$$

QED

# Splitting Graphs III

Thm let  $V = V_1 + \dots + V_k$ ,  $|V_i| = n$ ,  $\forall i$ . Then  $\exists S \subseteq V$  for which

$$\left| * \{E \in \binom{S}{k} : |E \cap V_i| = 1, \forall i\} - * \{E \in \binom{S}{k} : |E \cap V_i| \neq 1 \text{ for some } i\} \right| \\ \geq C_k n^k \text{ for some constant } C_k > 0 \text{ independent of } n.$$

Pf: Let  $S$  be a random vertex subset s.t.  $P(v \in S) = p_i$ ,  $\forall v \in V_i$  (mutually independent). For  $E \in \binom{V}{k}$ ,  $h(E) \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } E \text{ is crossing w.r.t. } V_1, \dots, V_k, \\ -1 & \text{o.w.} \end{cases}$ ,  $X_E \stackrel{\text{def}}{=} \begin{cases} h(E) & \text{if } E \subseteq S \\ 0 & \text{o.w.} \end{cases}$

$$E \left| \sum_{E \in \binom{V}{k}} X_E \right| \geq \left| \sum_E EX_E \right| = \left| \sum_{E \in \binom{V}{k}} h(E) P(E \subseteq S) \right| \\ = \left| \sum_{a_1 + \dots + a_k = k} \left( \sum_{\substack{|E \cap V_i| = a_i \\ i=1, \dots, k; E \in \binom{V}{k}}} h(E) p_1^{a_1} \dots p_k^{a_k} \right) \right| = n^k |f(p_1, p_2, \dots, p_k)| \geq n^k C_k \quad (\exists \text{ such } p_1, \dots, p_k \in [0, 1])$$

P15 Lemma 2.2.4 P15

where  $f(x_1, \dots, x_k) = \sum_{a_1 + \dots + a_k = k} \frac{*}{n^k} x_1^{a_1} x_2^{a_2} \dots x_k^{a_k}$  is a homogeneous polynomial of degree  $k$  with coefficient  $(x_1 x_2 \dots x_k) = 1$  and

$$|\text{coefficient } (x_1^{a_1} x_2^{a_2} \dots x_k^{a_k})| \leq \frac{(a_1)(a_2) \dots (a_k)}{n^k} \leq 1, \quad x_1, \dots, x_k \in [0, 1].$$

QED

# Balancing Vectors I

**Thm**

Suppose  $U_1, \dots, U_n \in \mathbb{R}^n$  with  $|U_i| = 1$ . Then  $\exists \varepsilon_1, \dots, \varepsilon_n \in \{\pm 1\}$  s.t.

$$|\varepsilon_1 U_1 + \varepsilon_2 U_2 + \dots + \varepsilon_n U_n| \leq \sqrt{n}$$

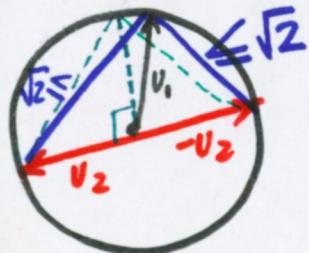
( $\geq$ ) column vector

**Pf:**  $X_1, \dots, X_n \stackrel{iid}{\sim} P(X_i = \pm 1) = \frac{1}{2}$ .  $Y \stackrel{\text{def}}{=} |\sum_{i=1}^n X_i U_i|$

$$\mathbb{E} Y^2 = \mathbb{E} \left( \left[ \sum_{i=1}^n X_i U_i \right]^t \left[ \sum_{i=1}^n X_i U_i \right] \right) = \sum_{1 \leq i, j \leq n} \mathbb{E}(X_i X_j) U_i^t U_j = \sum_{i=1}^n \mathbb{E}(X_i^2) |U_i|^2 = n.$$

**QED**

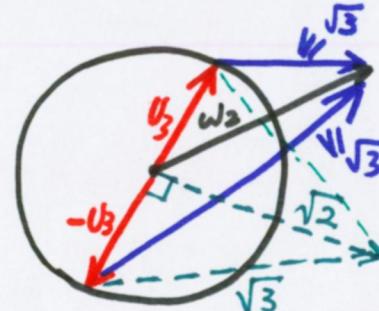
nonprobabilistic proof: **Idea** let  $w_i = \varepsilon_1 U_1 + \dots + \varepsilon_i U_i$ .  $i=1, 2, \dots, n$ .



$$\Rightarrow |U_1 + U_2| \leq \sqrt{2}$$

$$\text{let } w_2 = U_1 + U_2$$

$$\text{i.e. } \varepsilon_1 = 1, \varepsilon_2 = 1.$$



$$\Rightarrow |w_2 - U_3| \leq \sqrt{3}$$

$$\Rightarrow |U_1 + U_2 - U_3| \leq \sqrt{3}$$

There we conclude that  $\exists \varepsilon_1, \dots, \varepsilon_n \in \{\pm 1\}$  s.t.

$$|\varepsilon_1 U_1 + \dots + \varepsilon_i U_i| \leq \sqrt{i} \quad i=1, 2, \dots, n.$$

**QED**

# Balancing Vectors II

Thm

Suppose  $u_1, \dots, u_n \in \mathbb{R}^n$  with  $|u_i| \leq 1$ .  $p_1, \dots, p_n$  are fixed values in  $[0, 1]$ .

Then  $\exists \varepsilon_1, \dots, \varepsilon_n \in \{0, 1\}$  s.t.  $|\sum_{i=1}^n (\varepsilon_i - p_i) u_i| \leq \sqrt{n}/2$

Pf: let  $Y = |\sum_{i=1}^n (X_i - p_i) u_i|$ , where  $X_1, X_2, \dots, X_n$  are independent rvs  $X_i \sim B(1, p_i)$ .

$$\mathbb{E} Y^2 = \mathbb{E}(\star^* \star) = \sum_{i=1}^n |u_i|^2 \mathbb{E}(X_i - p_i)^2 = \sum_{i=1}^n p_i(1-p_i) |u_i|^2 \leq n/4.$$

**QED**

non probabilistic proof: We claim that  $\exists \varepsilon_1, \dots, \varepsilon_n \in \{0, 1\}$  s.t.  $|\sum_{i=1}^n (\varepsilon_i - p_i) u_i| \leq \frac{\sqrt{n}}{2}$

for  $s = 1, 2, \dots, n$ . Indeed, let  $w_s = \sum_{i=1}^s (\varepsilon_i - p_i) u_i$ . we have

$|w_s|^2 \leq |\varepsilon_1 - p_1|^2$ . If  $p_1 \leq \frac{1}{2}$  then choose  $\varepsilon_1 = 0$ . If  $p_1 \geq \frac{1}{2}$  then choose  $\varepsilon_1 = 1$ .

Thus  $\exists \varepsilon_1 \in \{0, 1\}$  s.t.  $|w_1|^2 \leq 1/4$ . By induction on  $s$ , let  $X \sim B(1, p_i)$

and  $f(X) = |w_{i-1} + (X - p_i) u_i|^2$ .

$$\mathbb{E} f(X) = \mathbb{E} \{ |w_{i-1}|^2 + 2(X - p_i) w_{i-1}^T u_i + |u_i|^2 (X - p_i)^2 \}$$

$$= |w_{i-1}|^2 + |u_i|^2 \text{Var } X = |w_{i-1}|^2 + |u_i|^2 p_i(1-p_i) \leq \frac{i-1}{4} + \frac{1}{4} = \frac{i}{4}.$$

**QED**

# Unbalancing Lights

Thm

Suppose  $A \in M_{n \times n}[\pm 1]$ . Then  $\exists x, y \in M_{n \times 1}[\pm 1]$  s.t.

$$x^t A y = (\sqrt{\frac{2}{\pi}} + o(1)) n^{\frac{3}{2}}$$

Pf: To show  $\exists y_1, \dots, y_n \in \{\pm 1\}$  s.t.  $\sum_{i=1}^n |\sum_{j=1}^n a_{ij} y_j| \geq (\sqrt{\frac{2}{\pi}} + o(1)) n^{\frac{3}{2}}$ .

Let  $Y_1, \dots, Y_n$  iid  $P(Y_i = 1) = P(Y_i = -1) = \frac{1}{2}$ . Let  $S_n^i = \sum_{j=1}^n a_{ij} Y_j$ . (note:  $a_{ij} Y_1, \dots, a_{in} Y_n \sim \text{iid}$ )

Then CLT  $\Rightarrow \frac{S_n^i - \mathbb{E} S_n^i}{\sqrt{\text{Var } S_n^i}} \xrightarrow{\text{converges in dis.}} N(0, 1) \Rightarrow \frac{S_n^i - 0}{\sqrt{n}} \rightarrow N(0, 1)$  ( $\because \mathbb{E} a_{ij} Y_j = 0$ ,  $\mathbb{E} Y_j^2 = 1$ )

Therefore  $\mathbb{E} |S_n^i| \sim \int_{\mathbb{R}} \sqrt{n} |t| \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$

$$= \frac{2\sqrt{n}}{\sqrt{2\pi}} \int_0^\infty t e^{-\frac{t^2}{2}} dt = \sqrt{\frac{2n}{\pi}} \quad \text{as } n \rightarrow \infty$$

$$\mathbb{E} \sum_{i=1}^n |\sum_{j=1}^n a_{ij} Y_j| = \sum_{i=1}^n \mathbb{E} |S_n^i| = \sum_{i=1}^n (\sqrt{\frac{2}{\pi}} + o(1)) n^{\frac{1}{2}} = (\sqrt{\frac{2}{\pi}} + o(1)) n^{\frac{3}{2}}$$

So we can choose  $x_i$  with the same sign as  $\sum_{j=1}^n a_{ij} y_j$  to get

$$\sum_{i=1}^n \sum_{j=1}^n x_i a_{ij} y_j = \sum_{i=1}^n x_i (\sum_{j=1}^n a_{ij} y_j) = \sum_{i=1}^n |\sum_{j=1}^n a_{ij} y_j| \geq (\sqrt{\frac{2}{\pi}} + o(1)) n^{\frac{3}{2}} \quad \text{QED}$$

$\alpha(G)$

Thm  $\alpha(G) \geq \sum_v \frac{1}{d_v + 1}$

pf (Method 1) let  $\pi$  be a random permutation of  $V = \{1, 2, \dots, n\}$

Let  $I = \{v \in V : \pi(v) > \pi(x) \text{ for any } x \in N(v)\}$

$$I_v = \begin{cases} 1 & \text{if } v \in I \\ 0 & \text{o.w.} \end{cases}$$

$$\begin{aligned} E|I| &= \sum_{v \in V} E I_v \\ &= \sum_{v \in V} \Pr \{ \pi(v) > \pi(x) \text{ for any } x \in N(v) \} \\ &= \sum_{v \in V} \frac{\binom{n}{d_v+1} (n-d_v-1)! d_v!}{n!} = \sum_v \frac{1}{d_v + 1} \end{aligned}$$

**QED** (Method 1)

**pf**

(Method 2) (Wei 1981, Griggs 1983)

Idea: 由  $G$  中不斷拿走現存 vertex  $\phi$  degree 最小的 vertex  $x$  及其鄰居 i.e. delete  $N[x]$ .

$$\begin{aligned} G &\longrightarrow \sum_{v \in G} \frac{1}{d_G(v)+1} \quad \star \\ G' = G - N[x] &\longrightarrow \sum_{v \in G'} \frac{1}{d_{G'}(v)+1} \quad D \\ &\vdots \end{aligned}$$

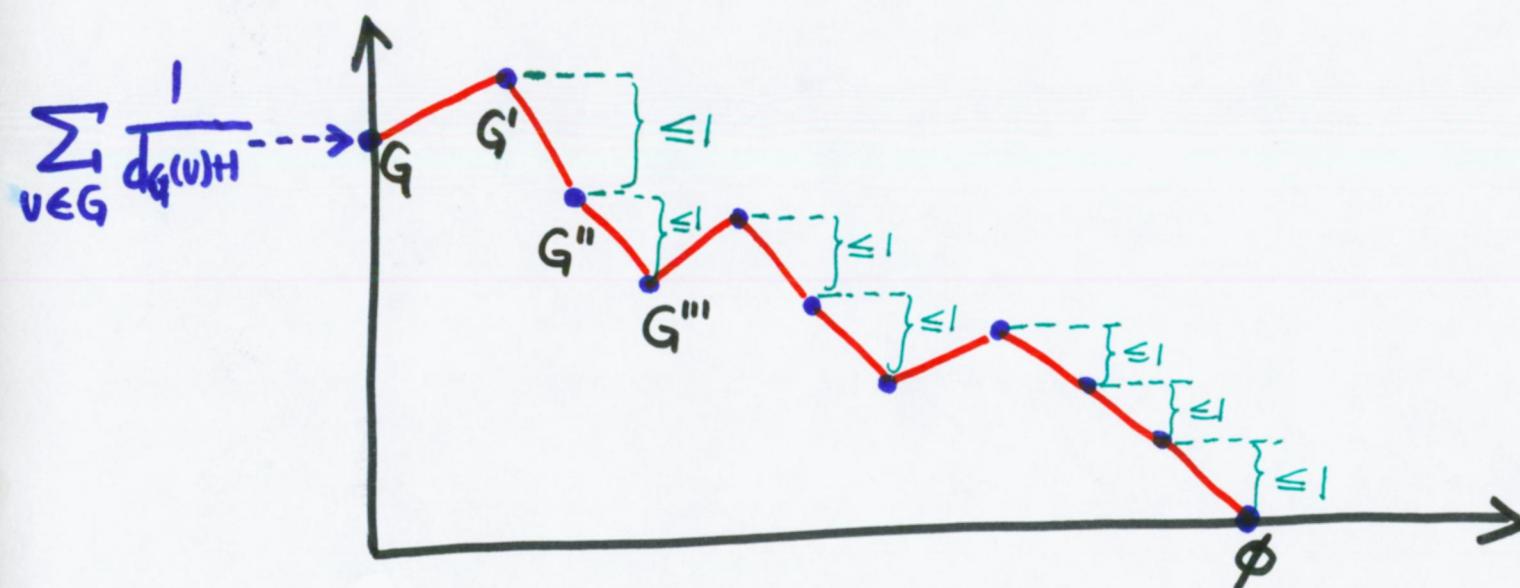
Claim: If  $\star > D$  then  $\star - D \leq 1$

**Pf**: If  $d_G(x) = \delta(G)$ , let  $N_G(x) = \{v_1, v_2, \dots, v_\ell\}$ , then

$$\begin{aligned} \star - D &\leq \frac{1}{d_G(v_1)+1} + \dots + \frac{1}{d_G(v_\ell)+1} + \frac{1}{d_G(x)+1} \\ &\leq \frac{d_G(x)+1}{d_G(x)+1} = 1 \end{aligned}$$

# NF (continued)

由下面圖示可知，去掉的步驟數 (i.e. 去掉步驟做的  
次數) 至少  $\geq \sum_{v \in G} \frac{1}{d_G(v)+1}$



Remark: see J. R. Griggs JCT ser.B 34, 22-39 (1983)

$$\alpha \geq \frac{\gamma}{\bar{d} + 1}$$

Thm  $\alpha(G) \geq \frac{v_G}{\bar{d} + 1}$ , where  $\bar{d} = \frac{2e_G}{v_G}$  the average degree of G

Pf: claim If  $x_1, \dots, x_k > 0$  then  $\frac{1}{x_1} + \dots + \frac{1}{x_k} \geq \frac{k^2}{x_1 + x_2 + \dots + x_k}$

Pf:  $x, y > 0 \Rightarrow \frac{y}{x} + \frac{x}{y} \geq 2$

so  $\sum_{1 \leq i < j \leq k} \left( \frac{x_i}{x_j} + \frac{x_j}{x_i} \right) \geq \binom{k}{2} 2 = k(k-1)$

$$\Rightarrow \frac{x_2 + x_3 + \dots + x_k}{x_1} + \frac{x_1 + x_3 + \dots + x_k}{x_2} + \dots + \frac{x_1 + x_2 + \dots + x_{k-1}}{x_k} \geq k(k-1)$$

$\Rightarrow$  Done! QED of Claim

$$\alpha(G) \geq \sum_v \frac{1}{d_v + 1} \geq \frac{v_G^2}{\sum_v (d_v + 1)} = \frac{v_G^2}{v_G + 2e_G}$$

QED

# Turán's Lower Bound

Thm

$$\alpha(G) \geq \frac{|V(G)|}{1 + \bar{d}}$$

average degree of  $G$ .

pf:

let  $n = |V(G)|$  and  $f(x) = \frac{1}{1+x}$ .  $\downarrow$  degree of  $v$

$$\begin{aligned}\alpha(G) &\geq n \sum_v \frac{1}{n} f(d(v)) \\ &\stackrel{\text{Jensen's Inequality}}{\geq} n f\left(\sum_v \frac{1}{n} d(v)\right) \\ &= n \frac{1}{1 + \bar{d}}\end{aligned}$$

**QED.**

# Convex Functions and Jensen's Inequality

A real-valued function  $f$  is *convex* on an interval  $I$  if and only if

$$f(ta + (1-t)b) \leq tf(a) + (1-t)f(b) \quad (1)$$

for all  $a, b \in I$  and  $0 \leq t \leq 1$ . This just says that a function is convex if the graph of the function lies below its secants. See pages 2 through 5 of Bjorn Poonen's paper, distributed at his talk on inequalities, for a discussion of convex functions and inequalities for convex functions. A number of common functions that are convex are also listed. Among those listed are  $-\ln x$  on  $(0, \infty)$ ,  $-\sin x$  on  $[0, \pi]$ ,  $-\cos x$  on  $[-\pi/2, \pi/2]$  and  $\tan x$  on  $[0, \pi/2]$ . To avoid the negative signs a complementary concept is defined. A real-valued function  $f$  is *concave* on an interval  $I$  if and only if

$$f(ta + (1-t)b) \geq tf(a) + (1-t)f(b) \quad (2)$$

for all  $a, b \in I$  and  $0 \leq t \leq 1$ . Therefore  $f$  is convex iff  $-f$  is concave. If you are familiar with derivatives then the following theorem about twice differentiable functions provides a way of telling if such a function is convex.

**Theorem 3** If  $f''(x) \geq 0$  for all  $x \in I$ , then  $f$  is convex on  $I$ .

Inequality (1) can be generalized to a convex function  $f$  with three variables  $x_1, x_2, x_3$  with weights  $t_1, t_2, t_3$ , respectively, such that  $t_1 + t_2 + t_3 = 1$ . Note that  $t_2 + t_3 = 1 - t_1$ . In this manner the three variable case can be transformed into the two variable case as follows.

$$\begin{aligned} f(t_1x_1 + t_2x_2 + t_3x_3) &= f\left(t_1x_1 + (1-t_1)\frac{t_2x_2 + t_3x_3}{t_2 + t_3}\right) \\ &\leq t_1f(x_1) + (1-t_1)f\left(\frac{t_2x_2 + t_3x_3}{t_2 + t_3}\right) \\ &= t_1f(x_1) + (1-t_1)f\left(\frac{t_2}{t_2 + t_3}x_2 + \frac{t_3}{t_2 + t_3}x_3\right) \\ &\leq t_1f(x_1) + (t_2 + t_3)\left(\frac{t_2}{t_2 + t_3}f(x_2) + \frac{t_3}{t_2 + t_3}f(x_3)\right) \\ &= t_1f(x_1) + t_2f(x_2) + t_3f(x_3). \end{aligned}$$

This process can be continued to produce an  $n$  variable version which is due to J.L.W.V. Jensen. It can be easily proved by mathematical induction using the above technique. Write your own proof and compare with the one given here. It will give you some good practice manipulating sigma notation.

**Theorem 4** (Jensen's Inequality 1906) Let  $f$  be a convex function on the interval  $I$ . If  $x_1, x_2, \dots, x_n \in I$  and  $t_1, t_2, \dots, t_n$  are nonnegative real numbers such that  $t_1 + t_2 + \dots + t_n = 1$ , then

$$f\left(\sum_{i=1}^n t_i x_i\right) \leq \sum_{i=1}^n t_i f(x_i).$$

**Proof by induction:** The case for  $n = 2$  is true by the definition of convex. Assume the relation holds for  $n$ , then we have

$$\begin{aligned} f\left(\sum_{i=1}^{n+1} t_i x_i\right) &= f\left(\sum_{i=1}^n t_i x_i + t_{n+1} x_{n+1}\right) = f\left(t_{n+1} x_{n+1} + (1-t_{n+1}) \frac{1}{1-t_{n+1}} \sum_{i=1}^n t_i x_i\right) \\ &\leq t_{n+1} f(x_{n+1}) + (1-t_{n+1}) f\left(\frac{1}{1-t_{n+1}} \sum_{i=1}^n t_i x_i\right) \\ &= t_{n+1} f(x_{n+1}) + (1-t_{n+1}) f\left(\sum_{i=1}^n \frac{t_i}{1-t_{n+1}} x_i\right) \\ &\leq t_{n+1} f(x_{n+1}) + (1-t_{n+1}) \sum_{i=1}^n \frac{t_i}{1-t_{n+1}} f(x_i) \\ &= \sum_{i=1}^n t_i f(x_i) + t_{n+1} f(x_{n+1}) \\ &= \sum_{i=1}^{n+1} t_i f(x_i). \end{aligned}$$

Thus showing that the assumption implies that the relation holds for  $n + 1$  and by the principle of Mathematical Induction holds for all natural numbers.

An easy consequence of Jensen's theorem is the following proof of the arithmetic mean-geometric mean inequality. (Problem 13 from Bjorn's paper)

**Theorem 5 (AM-GM Inequality)** If  $x_1, x_2, \dots, x_n \geq 0$  then

$$\frac{x_1 + x_2 + \cdots + x_n}{n} \geq \sqrt[n]{x_1 x_2 \cdots x_n}.$$

Proof. Since  $-\ln x$  is convex then  $\ln x$  is concave. By Jensen's theorem we have

$$\begin{aligned}\ln\left(\frac{x_1 + x_2 + \cdots + x_n}{n}\right) &\geq \frac{\ln x_1 + \ln x_2 + \cdots + \ln x_n}{n} \\ &= \frac{1}{n} \ln(x_1 x_2 \cdots x_n) \\ &= \ln[(x_1 x_2 \cdots x_n)^{\frac{1}{n}}]\end{aligned}$$

Since  $\ln x$  is monotonic increasing ( $f'(x) = \frac{1}{x} > 0$ ) for  $x > 0$  we have

$$\frac{x_1 + x_2 + \cdots + x_n}{n} \geq \sqrt[n]{x_1 x_2 \cdots x_n}.$$

The proof of Jensen's Inequality does not address the specification of the cases of equality. It can be shown that strict inequality exists unless all of the  $x_i$  are equal or  $f$  is linear on an interval containing all of the  $x_i$ .