

Ricci curvature of Cartesian product graphs.

①

Recall $G = (V(G), E(G))$, $H = (V(H), E(H))$.

The Cartesian product $G \times H$ is defined as

$$V(G \times H) = V(G) \times V(H).$$

$(u_1, v_1) \sim_{G \times H} (u_2, v_2)$ iff either $u_1 \sim_G u_2, v_1 = v_2 \in V(H)$

or $u_1 = u_2 \in V(G), v_1 \sim_H v_2$

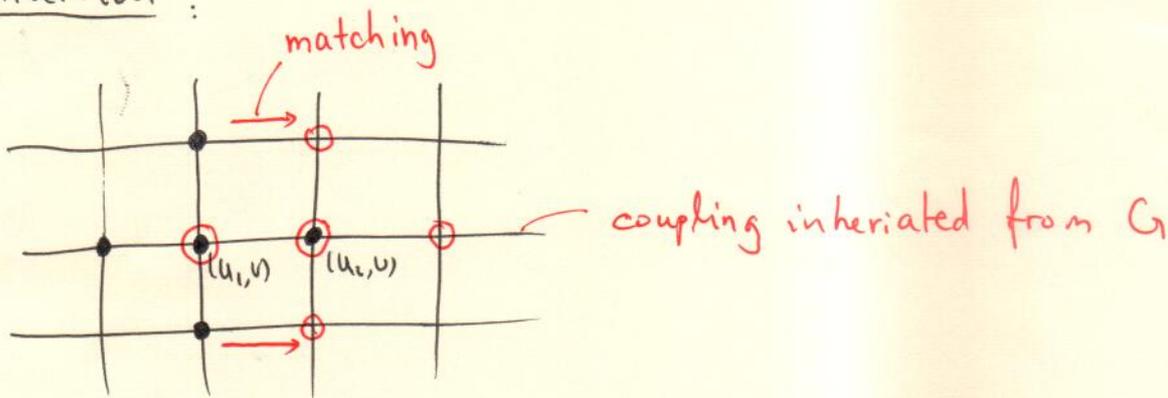
Assume G is d_G -regular, H is d_H -regular.

~~Thm~~ - Lemma 1. For $p \in [0, 1]$, we have

$$K_p^{G \times H}((u_1, v), (u_2, v)) \geq \frac{d_G}{d_G + d_H} K_p^G(u_1, u_2)$$

$u_1 \sim_G u_2$

Intuition:



• $m_{(u_1, v)}^\alpha$ ○ $m_{(u_2, v)}^\alpha$

(u_1, v)

(u_2, v)

$\{(x_1, v) : x_1 \sim_G u_1\}$

\xrightarrow{A}

$\{(x_2, v) : x_2 \sim_G u_2\}$

$\{(u_1, y_1) : y_1 \sim_H v\}$

$\xrightarrow{\text{matching}}$

$\{(u_2, y_2) : y_2 \sim_H v\}$

$$A: V(G) \times V(G) \rightarrow \mathbb{R}$$

is the coupling s.t.

$$W_1^G(m_{u_1}^\alpha, m_{u_2}^\alpha)$$

$$= \sum_{\{x_1, x_2 \in V(G)\}} A^{u_1, u_2}(x_1, x_2) d(x_1, x_2)$$

Coupling: $B: V(G \times H) \times V(G \times H) \rightarrow \mathbb{R}$

$$B((u_1, y_1), (u_2, y_1)) = \frac{d_H}{d_G + d_H} m_v^\alpha(y_1), \quad y_1 \neq v$$

$$B((x_1, v), (x_2, v)) = \frac{d_G}{d_G + d_H} A^{u_1, u_2}(x_1, x_2), \quad (x_1, x_2) \neq (u_1, u_2)$$

$$B((u_1, v), (u_2, v)) = \frac{d_G}{d_G + d_H} A^{u_1, u_2}(u_1, u_2) + \frac{d_H}{d_G + d_H} \cdot \alpha. \quad (2)$$

↑
上述两种情形均适用

One can check B is a coupling of $m_{(u_1, v)}^\alpha$ & $m_{(u_2, v)}^\alpha$ and calculate the corresponding transport cost.

Alternatively, we formulate the proof a bit "abstractly".

Proof: First, we observe W_1 satisfies a convexity property. For any $\lambda \in [0, 1]$, ~~μ_1, ν be two prob~~
 $\mu_i, \nu_i, i=1, 2$ be probability measures. Then

$$W_1(\lambda \mu_1 + (1-\lambda) \mu_2, \lambda \nu_1 + (1-\lambda) \nu_2) \leq \lambda W_1(\mu_1, \nu_1) + (1-\lambda) W_1(\mu_2, \nu_2). \quad (*)$$

(This has been shown last time when ^{we} prove the concavity of $K_\alpha(x, y)$. Indeed, let $\xi_i, i=1, 2$ be the optimal coupling of μ_i & ν_i, μ_2 & ν_2 respectively. Then $\lambda \xi_1 + (1-\lambda) \xi_2$ is a coupling of $\lambda \mu_1 + (1-\lambda) \mu_2$ & $\lambda \nu_1 + (1-\lambda) \nu_2$. The convexity $(*)$ follows from a direct calculation.)

Observe that

$$m_{(u_1, v)}^\alpha \stackrel{(\ast)}{=} \frac{d_G}{d_G + d_H} m_{u_1}^\alpha \otimes \delta_v + \frac{d_H}{d_G + d_H} \delta_{u_1} \otimes m_v^\alpha$$

where the tensorization product is defined as

$$(m_{u_1}^\alpha \otimes \delta_v)(x, y) := m_{u_1}^\alpha(x) \cdot \delta_v(y).$$

Therefore, we have

(3)

$$W_1(m_{(u_1, v)}^\alpha, m_{(u_2, v)}^\alpha)$$

$$= W_1(\lambda m_{u_1}^\alpha \otimes \delta_v + (1-\lambda) \delta_{u_1} \otimes m_v^\alpha, \lambda m_{u_2}^\alpha \otimes \delta_v + (1-\lambda) \delta_{u_2} \otimes m_v^\alpha)$$

$$\text{with } \lambda = \frac{d_G}{d_G + d_H}$$

By convexity, we derive

$$W_1(m_{(u_1, v)}^\alpha, m_{(u_2, v)}^\alpha)$$

$$\leq \lambda \underbrace{W_1(m_{u_1}^\alpha \otimes \delta_v, m_{u_2}^\alpha \otimes \delta_v)}_{= W_1^G(m_{u_1}^\alpha, m_{u_2}^\alpha)} + (1-\lambda) \underbrace{W_1(\delta_{u_1} \otimes m_v^\alpha, \delta_{u_2} \otimes m_v^\alpha)}$$

$$\leq d(u_1, u_2) = 1$$

Since there is a perfect matching

$$(u_1, y) \rightarrow (u_2, y)$$

for any $y \in \text{supp}(m_v^\alpha)$

$$\leq \lambda W_1^G(m_{u_1}^\alpha, m_{u_2}^\alpha) + (1-\lambda)$$

Consequently, we obtain

$$K_\alpha^{G \times H}((u_1, v), (u_2, v)) = 1 - W_1(m_{(u_1, v)}^\alpha, m_{(u_2, v)}^\alpha)$$

$$\geq 1 - \lambda W_1^G(m_{u_1}^\alpha, m_{u_2}^\alpha) - (1-\lambda)$$

$$= \lambda (1 - W_1^G(m_{u_1}^\alpha, m_{u_2}^\alpha)) = \lambda K_\alpha^G(u_1, u_2). \quad \square$$

In particular, the positivity of K_P is preserved under taking Cartesian product.

Lemma 2: $K_\alpha^{G \times H}((u_1, v), (u_2, v)) \leq \frac{d_G + \alpha d_H}{d_G + d_H} K_{\alpha'}^G(u_1, u_2)$

for $\alpha \in [0, 1]$, $u_1 \sim_G u_2$, $\alpha' = \alpha \cdot \frac{d_G + d_H}{d_G + \alpha d_H}$

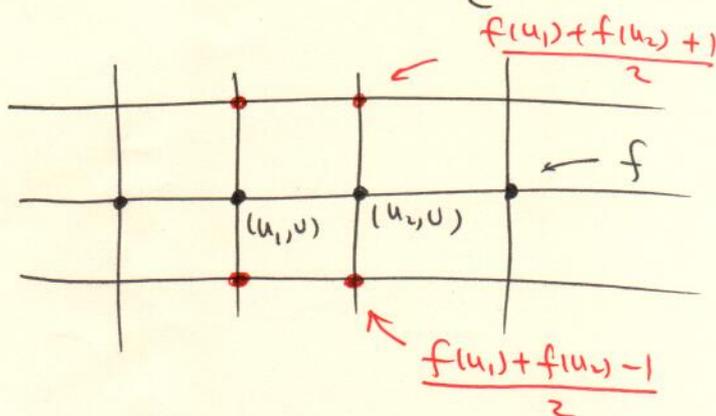
Proof: We do it by duality formula. (4)

Let $f: V(G) \rightarrow \mathbb{R}$ be a 1-Lipschitz function

s.t.
$$W_1^G(m_{u_1}^{\alpha'}, m_{u_2}^{\alpha'}) = \sum_{x \in V(G)} f(x) (m_{u_1}^{\alpha'}(x) - m_{u_2}^{\alpha'}(x))$$

Construct $F: V(G \times H) \rightarrow \mathbb{R}$ such that.

$$F(x, y) = \begin{cases} f(x), & \text{if } y = v \\ \frac{f(u_1) + f(u_2) + 1}{2}, & \text{if } x = u_1, y \sim v, y \neq v \\ \frac{f(u_1) + f(u_2) - 1}{2}, & \text{if } x = u_2, y \sim v, y \neq v \end{cases}$$



$$W_r(m_{(u_1, v)}^{\alpha}, m_{(u_2, v)}^{\alpha})$$

$$\geq \sum_{(x, y)} F(x, y) (m_{(u_1, v)}^{\alpha}(x, y) - m_{(u_2, v)}^{\alpha}(x, y))$$

$$= \lambda \sum_{(x, y)} F(x, y) (m_{u_1}^{\alpha} \otimes \delta_v(x, y) - m_{u_2}^{\alpha} \otimes \delta_v(x, y))$$

$$+ (1-\lambda) \sum_{(x, y)} F(x, y) (m_{u_1}^{\alpha} \otimes \delta_{u_1} \otimes m_v^{\alpha}(x, y) - \delta_{u_2} \otimes m_v^{\alpha}(x, y))$$

$$= \lambda \sum_x f(x) (m_{u_1}^{\alpha}(x) - m_{u_2}^{\alpha}(x))$$

$$+ (1-\lambda) \frac{f(u_1) + f(u_2) + 1}{2} \sum_{y \neq v} m_v^{\alpha}(y) - (1-\lambda) \frac{f(u_1) + f(u_2) - 1}{2} \sum_{y \neq v} m_v^{\alpha}(y)$$

$$\otimes + (1-\lambda) f(u_1) \alpha - f(u_2) \alpha \cdot (1-\lambda)$$

$$\begin{aligned}
&= (1-\alpha)(1-\lambda) + \lambda \sum_{x \neq u_1} f(x) m_{u_1}^\alpha(x) - \lambda \sum_{x \neq u_2} f(x) m_{u_2}^\alpha(x) \quad (\oplus) \\
&\quad + \lambda f(u_1) \cdot \alpha - \lambda f(u_2) \alpha \\
&\quad + (1-\lambda) f(u_1) \alpha - (1-\lambda) f(u_2) \alpha \\
&= (1-\alpha)(1-\lambda) + \alpha \left(\sum_{x \neq u_1} f(x) m_{u_1}^\alpha(x) + f(u_1) \right) - \alpha \left(\sum_{x \neq u_2} f(x) m_{u_2}^\alpha(x) + f(u_2) \right)
\end{aligned}$$

The measure

$$P_{u_1}(x) = \begin{cases} \lambda m_{u_1}^\alpha(x) = \lambda \frac{1-\alpha}{d_G}, & x \sim u_1 \\ \alpha & x = u_1 \end{cases}$$

satisfies $\sum_x P_{u_1}(x) = (1-\alpha)\lambda + \alpha$

Similarly, for $P_{u_2}(x) := \begin{cases} \lambda m_{u_2}^\alpha(x) = \lambda \frac{1-\alpha}{d_G}, & x \sim u_2 \\ \alpha & x = u_2 \end{cases}$

has $\sum_x P_{u_2}(x) = (1-\alpha)\lambda + \alpha$.

Scaling them to be probability measures:

$$\frac{1}{(1-\alpha)\lambda + \alpha} P_{u_1}(\cdot) \quad \text{and} \quad \frac{1}{(1-\alpha)\lambda + \alpha} P_{u_2}(\cdot)$$

$\Rightarrow W_1 (m_{(u_1, u_1)}^\alpha, m_{(u_2, u_2)}^\alpha)$
 $\oplus = (1-\alpha)(1-\lambda) +$

Observe that

$$\frac{1}{(1-\alpha)\lambda + \alpha} P_{u_1}(x) = \begin{cases} \frac{(1-\alpha)\lambda}{(1-\alpha)\lambda + \alpha} \frac{1}{d_G} & x \sim u_1 \\ \frac{\alpha}{(1-\alpha)\lambda + \alpha} & x = u_1 \end{cases}$$

$$= m_{u_1}^{\alpha'}(x)$$

with $\alpha' = \frac{\alpha}{(1-\alpha)\lambda + \alpha}$.

Similarly,

(6)

$$\frac{1}{(1-\alpha)\lambda + \alpha} p_{u_2}(x) = m_{u_2}^{\alpha'}(x)$$

$$\Rightarrow W_1(m_{(u_1, v)}^{\alpha}, m_{(u_2, v)}^{\alpha})$$

$$\Rightarrow (1-\alpha)(1-\lambda) + W_1^G(m_{u_1}^{\alpha'}(x), m_{u_2}^{\alpha'}(x)) \cdot ((1-\alpha)\lambda + \alpha)$$

$$\Rightarrow k_{\alpha}^{G \times H}((u_1, v), (u_2, v))$$

$$= 1 - W_1(m_{(u_1, v)}^{\alpha}, m_{(u_2, v)}^{\alpha})$$

$$\leq 1 - (1-\alpha)(1-\lambda) - ((1-\alpha)\lambda + \alpha) W_1^G(m_{u_1}^{\alpha'}(x), m_{u_2}^{\alpha'}(x))$$

$$= ((1-\alpha)\lambda + \alpha) (1 - W_1^G(m_{u_1}^{\alpha'}(x), m_{u_2}^{\alpha'}(x)))$$

$$= ((1-\alpha)\lambda + \alpha) k_{\alpha'}^G(u_1, u_2). \quad \square$$

$$\frac{k_{\alpha}^{G \times H}((u_1, v), (u_2, v))}{1 - \alpha} = \frac{((1-\alpha)\lambda + \alpha)}{1 - \alpha} k_{\alpha'}^G(u_1, u_2)$$

$$= \lambda \cdot \frac{((1-\alpha)\lambda + \alpha)}{(1-\alpha)\lambda} k_{\alpha'}^G(u_1, u_2)$$

$$= \lambda \frac{1}{1-\alpha'} k_{\alpha'}^G(u_1, u_2)$$