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Turán's Theorem Formalization

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0.1 Concentrating support on a clique - Improve Operation

Definition 1 (A better distribution). Better Given a weight function W , define $w^* := \text{Better}(W)$ such that $\text{supp}(w^*) \subseteq \text{supp}(W)$ and

$$W.\text{fw} \leq w^*.\text{fw}.$$

In words: w^* is a distribution improving or preserving the edge weight and supported within that of W .

Definition 2 (Single transfer). Improve Given distinct vertices $v_j \neq v_i$, define $w' := \text{Improve}(W, v_j, v_i)$ by transferring a small amount of weight from v_j (loose) to v_i (gain):

$$w'(v_j) = W(v_j) - \varepsilon, \quad w'(v_i) = W(v_i) + \varepsilon,$$

for some small $\varepsilon > 0$, and $w'(v) = W(v)$ for $v \neq v_i, v_j$. In words: weight is shifted from v_j to v_i preserving total weight.

Lemma 3 (Sum splitting along the partition). *Improve_partition_sum_splitThe_sum_over_edges_E of the function \mathbf{vp} splits as*

$$\sum_{e \in E} \mathbf{vp}(w', e) = \sum_{\substack{e \in E \\ e \ni v_i}} \mathbf{vp}(w', e) + \sum_{\substack{e \in E \\ e \ni v_j}} \mathbf{vp}(w', e) + \sum_{\substack{e \in E \\ e \not\ni v_i, v_j}} \mathbf{vp}(w', e).$$

In words: the total sum decomposes into parts incident to v_i , incident to v_j , and the rest.

Lemma 4 (Gain-incidence increases). *Improve_gain_contribution_increaseThe_increment_inthe_sum_over_edges_inc is*

$$\sum_{e \ni v_i} \mathbf{vp}(w', e) - \sum_{e \ni v_i} \mathbf{vp}(W, e) = \varepsilon \sum_{v_k \in N(v_i)} W(v_k),$$

where $N(v_i)$ is the neighborhood of v_i . In words: the gain vertex's contribution increases by ε times the sum of weights of its neighbors.

Lemma 5 (Loose-incidence becomes zero). *Improve_loose_contribution_zeroThe_sum_over_edges_incident_to_v_j after the transfer satisfies*

$$\sum_{e \ni v_j} \mathbf{vp}(w', e) = 0.$$

In words: the loose vertex's incident edge contributions vanish after the transfer.

Lemma 6 (Unchanged complement). *Improve_unchanged_edge_sum_Fore_dges_e not incident to v_i or v_j ,*

$$\mathbf{vp}(w', e) = \mathbf{vp}(W, e).$$

In words: edges outside the gain and loose neighborhoods remain unchanged.

Lemma 7 (Transfer does not decrease fw). *Improve_{total_weight_nondecl} : Improve_{partition_sum_split}, lem : Improve_{fw}. In words: the total edge weight does not decrease after applying Improve.*

Lemma 8 (Improve strictly reduces support). *Improve_{support_strictly_reduced} : Improve_{If_the_neighborhood}. In words: the transfer strictly reduces the number of vertices with positive weight.*

Theorem 9 (Support of Better is a clique). *Better_{forms_clique} : Improve, lem : Improve_{total_weight_nondecl}. Better(W) forms a clique:*

$$\forall v_a, v_b \in \text{supp}(w^*), v_a \neq v_b \implies \{v_a, v_b\} \in E.$$

In words: every two distinct vertices with positive weight in w^ are adjacent.*

0.2 The Enhance Operation

Definition 10 (Enhance). Enhance Given distinct non-adjacent vertices v_j, v_i , define $w' := \text{Enhance}(W, v_j, v_i, \varepsilon)$ by transferring $\varepsilon > 0$ weight from v_j to v_i :

$$w'(v_j) = W(v_j) - \varepsilon, \quad w'(v_i) = W(v_i) + \varepsilon, \quad w'(v) = W(v) \text{ for } v \neq v_i, v_j,$$

with the condition $\{v_j, v_i\} \notin E$. In words: Enhance transfers weight between non-adjacent vertices to increase edge weight.

Lemma 11 (Sum over support). *sum_{over_support} The total vertex weights satisfies $\sum_{v \in \text{supp}(W)} W(v) = 1$. In words: the weights sum to 1 over the support.*

Lemma 12 (Supported edge partition). *supported_{edge_partition} The edgeset E partitions as*

$$E = E_{v_i} \cup E_{v_j} \cup E_{\text{rest}},$$

where

$$E_{v_i} = \{e \in E : v_i \in e\}, \quad E_{v_j} = \{e \in E : v_j \in e\}, \quad E_{\text{rest}} = E \setminus (E_{v_i} \cup E_{v_j}).$$

In words: edges are split into those incident to v_i , to v_j , and the rest.

Lemma 13 (Enhance gain sum). *Enhance_{gain_sum} Under ??, the change in the sum over edges incident to v_i satisfies*

$$\sum_{e \in E_{v_i}} \text{vp}(w', e) - \sum_{e \in E_{v_i}} \text{vp}(W, e) = \varepsilon \sum_{v_k \in N(v_i)} W(v_k).$$

In words: the gain vertex's edge contribution increases by ε times the sum of its neighbors' weights.

Lemma 14 (Enhance loose sum). *Enhance_{loose_sum} Under ??, the sum over edges incident to v_j satisfies*

$$\sum_{e \in E_{v_j}} \text{vp}(w', e) = 0.$$

In words: the loose vertex's incident edge contributions become zero after Enhance.

Definition 15 (Bijection inside the clique). *Define a bijection $\phi : \{e \in E_{v_j} \setminus \{s(v_j, v_i)\}\} \rightarrow \{e \in E_{v_i} \setminus \{s(v_j, v_i)\}\}$ mapping edges incident to v_j (except $s(v_j, v_i)$) to edges incident to v_i (except $s(v_j, v_i)$). In words: this bijection pairs edges incident to v_j with edges incident to v_i within the clique.*

Lemma 16 (Bijection preserves). *For any edge e incident to v_j (excluding $s(v_j, v_i)$), the "other" vertex weight satisfies*

$$W(\text{other}(e, v_j)) = W(\text{other}(\phi(e), v_i)).$$

In words: the bijection preserves weights at the other endpoints of edges.

Lemma 17 (Loose/gain equality). *Enhance_{sum} loose gain equal def : the bij lemma : the bij sum The total weight tr*
 $\sum_{e \in E_{v_j}} \mathbf{vp}(w', e) + \sum_{e \in E_{v_i}} \mathbf{vp}(w', e) \geq \sum_{e \in E_{v_j}} \mathbf{vp}(W, e) + \sum_{e \in E_{v_i}} \mathbf{vp}(W, e)$. In words: the combined edge contributions of loose and gain vertices do not decrease after Enhance.

Lemma 18 (Complement unchanged). *Enhance_{sum} complement unchanged For edge $e \in E_{\text{rest}}$,*

$$\mathbf{vp}(w', e) = \mathbf{vp}(W, e).$$

In words: edges not incident to v_i or v_j remain unaffected by Enhance.

Lemma 19 (Edge contribution increase). *Enhance_{edge} gain loose increase Then edge contributions satisfies* $\sum_{e \in E_{v_i} \cup E_{v_j}} \mathbf{vp}(W, e)$. In words: the total contribution from gain and loose vertices does not decrease.

Lemma 20 (Support edges unchanged). *Enhance_{support} edge sum For any vertex $v \notin \{v_i, v_j\}$, the edge contributions satisfy*

$$\sum_{e \ni v} \mathbf{vp}(w', e) = \sum_{e \ni v} \mathbf{vp}(W, e).$$

In words: vertices outside gain and loose retain their edge contributions after Enhance.

Theorem 21 (Enhance increases edge weight). *Enhance_{total weight} tric inclem : supported edge partition, lemma*
 $w'.\text{fw} > W.\text{fw}$. In words: the Enhance operation strictly improves the total edge weight.

0.3 Equalizing the weights on the clique - Enhanced

Definition 22 (Carefully chosen ε). *Define the $_ \varepsilon := \max \left\{ \varepsilon > 0 \mid w'(v_j) = W(v_j) - \varepsilon \geq 0, \quad w'(v_i) = W(v_i) + \varepsilon \right\}$*
In words: the $_ \varepsilon$ is the maximal ε transferring weight from the argmax vertex v_j to the argmin vertex v_i without violating support constraints.

Definition 23 (Maximising the number of uniform vertices). $\text{max}_u \text{uniform}_s \text{supportDefin} m := \max\{k \mid \exists w \text{ with } \text{supp}(w) \subseteq \text{supp}(W), w.\text{fw} \geq W.\text{fw}, \text{ and } w(v) = \frac{1}{k} \text{ for at least } k \text{ vertices}\}$. In words: m is the maximal number of vertices with uniform weight $1/k$ achievable without decreasing edge weight.

Lemma 24 (Best uniform distribution exists). $\text{exists}_{best_u} \text{uniformDef} : \text{max}_u \text{uniform}_s \text{support} \text{There exists } w_M \text{ with } \text{supp}(w_M) \subseteq \text{supp}(W), w_M.\text{fw} \geq W.\text{fw}, \text{ and with at least } m \text{ vertices having weight } 1/m. \text{ In words: a maximiser } w_M \text{ achieving the maximal uniform vertex count exists.}$

Definition 25 (UniformBetter). $\text{UniformBetter lem:exists}_{best_u} \text{uniformDefin} w_M := \text{UniformBetter}(W)$ as such a maximiser with maximal uniform support.

Definition 26 (Enhanced). $\text{Enhanced def:Enhance, def:the}_{eps} \text{Defin} w^+ := \text{Enhanced}(W) := \text{Enhance}(W, v_j, v_i, \text{the_}\varepsilon)$, where $v_j = \arg \max_v W(v)$, $v_i = \arg \min_v W(v)$. In words: Enhanced transfers maximal weight from the heaviest to the lightest vertex.

Lemma 27. $\text{Enhanced}_u \text{naffectedDef} : \text{Enhance, def} : \text{EnhancedFor any vertex } v \text{ with } W(v) = \frac{1}{|\text{supp}(W)|},$

$$w^+(v) = W(v).$$

In words: vertices already at uniform weight remain unchanged under Enhanced.

Lemma 28. $\text{Enhanced}_e \text{fect}_{argmax} \text{Def} : \text{Enhance, def} : \text{EnhancedThe weight at the } argmax \text{ vertex } v_j \text{ after Enhanced satisfies}$

$$w^+(v_j) = \frac{1}{|\text{supp}(W)|}.$$

In words: Enhanced reduces the $argmax$ vertex's weight to the uniform level.

Lemma 29. $\text{Enhanced}_i \text{nc}_u \text{uniform}_c \text{ountDef} : \text{Enhanced, lem} : \text{Enhanced}_e \text{fect}_{argmax}, \text{lem} : \text{Enhanced}_u \text{naffected increases after Enhanced:}$

$$|\{v : w^+(v) = 1/|\text{supp}(W)|\}| > |\{v : W(v) = 1/|\text{supp}(W)|\}|.$$

In words: Enhanced increases the count of uniform weight vertices.

Lemma 30. def:UniformBetter The support of W forms a clique if and only if the support of $w_M = \text{UniformBetter}(W)$ forms a clique:

$$\text{supp}(W) \text{ clique} \iff \text{supp}(w_M) \text{ clique}.$$

In words: the clique property is preserved by UniformBetter.

Lemma 31 (Uniform weights on the support). $\text{UniformBetter}_{constant_s} \text{upportDef} : \text{UniformBetter, def} : \text{Enhanced}_u \text{naffected}$

$$w_M(v) = \frac{1}{|\text{supp}(w_M)|}.$$

In words: the weights of all support vertices in UniformBetter are uniform.

Lemma 32 (Edge values under UniformBetter). *UniformBetter_{edges} value lemma : UniformBetter_{constant} supports*
 $\{v_a, v_b\}$ with $v_a, v_b \in \text{supp}(w_M)$,

$$\text{vp}(w_M, e) = \left(\frac{1}{|\text{supp}(w_M)|} \right)^2.$$

In words: every supported edge has value equal to the square of the uniform vertex weight.

Lemma 33 (Edge count in a clique). *clique_{size} lemma : UniformBetter_{actsIf} |supp(w_M)| = k , then*

$$|\{e \in E : e \subseteq \text{supp}(w_M)\}| = \frac{k(k-1)}{2}.$$

In words: the supported edges form a complete graph on k vertices.

Lemma 34 (A light computation). *computation For $k > 0$,*

$$\frac{k(k-1)}{2} \cdot \left(\frac{1}{k} \right)^2 = \frac{1}{2} \left(1 - \frac{1}{k} \right).$$

In words: the total edge weight for a clique with uniform weights simplifies to $\frac{1}{2}(1 - \frac{1}{k})$.

Lemma 35 (Monotonicity of the bound). *bound bound_{real} The function $f(k) := \frac{1}{2}(1 - \frac{1}{k})$ is nondecreasing for $k \geq 1$. In words: the bound increases as the clique size increases.*

Theorem 36 (Final bound inside a clique). *final_{bound} lemma : Better_{forms} clique, lemma : Better_{nondecr}, lemma : Better_{bound}*
is supported on a clique of size $k \leq p-1$, then

$$\text{W.fw} \leq \frac{1}{2} \left(1 - \frac{1}{p-1} \right).$$

In words: the total edge weight is bounded by the Turán bound for cliques of size less than p .

Definition 37 (Uniform weights over all vertices). *UnivFun Define*

$$\text{UnivFun}(G)(v) := \frac{1}{|V|} \quad \forall v \in V.$$

In words: the uniform vertex weight function assigns equal weight $1/|V|$ to every vertex.

Lemma 38 (Total weight under UnivFun). *UnivFun_{weight} def : UnivFun The total edge weights satisfies (UnivFun_{weight} lemma)*
 $|E| \cdot \left(\frac{1}{|V|} \right)^2$. *In words: the total edge weight under uniform vertex weights equals the number of edges times the square of the uniform weight.*

Theorem 39 (Turán's Theorem). *turans def:UnivFun, lem:UnivFun_weight, lem : finale_bound, lem : computationLet $p \geq 2$ and let G be a p -clique-free graph. Then*

$$|E| \leq \frac{1}{2} \left(1 - \frac{1}{p-1} \right) |V|^2.$$

In words: the number of edges in a p -clique-free graph is bounded by Turán's theorem.