A APPENDIX

A.1 Proof of Theorem 3.1

We begin by introducing some notation from prior work [1]. First, recall that a degree sequence for a relation R and variable V is a vector $f_{R,V}$ such that $f_{R,V,i}$ is the frequency of the ith most frequent value, and $F_{R,V,i} = \sum_{j=1,i} f_{R,V,j}$ is the cumulative degree sequence (CDS). We represent relations R as tensors M_R with one dimension for each variable present in the relation. The value at a particular entry $M_{i_1,\dots,i_{|V_R|}}$ is equal to the frequency of that tuple, $(i_1,\dots,i_{|V_R|})$, in the relation R. Note that entries can be zero if that tuple does not appear in R. Additionally, the discrete derivative and integral of a tensor M on a variable V is defined as,

Definition A.1.

$$(\Delta_V M)_{v_i} = M_{v_i} - M_{v_i - 1} \tag{1}$$

$$(\Sigma_V M)_{v_i} = \sum_{j=1, v_i} M_j \tag{2}$$

A tensor M is *consistent* with a set of degree sequences f_R if the following is true,

$$(\sum_{V'\neq V} M)_i \leq f_{R,V,i} \qquad \forall \ i \in \mathbb{D}_V, V \in V_R$$

Briefly, this means that if we contract the tensor down to a single dimension, then the resulting vector is less than the DF $F_{R,V}$ at all points. The set \mathcal{M}_{f_R} is the set of tensors consistent with f_R .

Lastly, we define the value tensor, E^{F_R} , and use it to explicitly define the worst-case tensor, C^{F_R} .

Definition A.2. The *value* tensor, $E^{F_R} \in \mathbb{R}^{[n]}_{\perp}$, is defined by the following linear optimization problem:

$$\forall m \in [n]:$$
 $E_{m}^{F_{R}} \stackrel{\text{def}}{=} \text{Maximize: } \sum_{s \le m} M_{s}$ (3)

Where: $M \in \mathcal{M}_{f_R}$

The *worst-case* tensor, $C^{F_R} \in \mathbb{R}^{[n]}$, is defined as:

$$C^{f_R} \stackrel{\text{def}}{=} \Delta_{V_1} \cdots \Delta_{V_d} E^{F_R} \tag{4}$$

Or, equivalently,

$$\Sigma_{V_1} \cdots \Sigma_{V_d} C^{f_R} = E^{F_R} \tag{5}$$

Note that this worst-case tensor is equivalent to the worst-case instance, W(R) of a relation R, as depicted in Figure 2.

Getting back to Theorem 3.1 of this work, we start by proving that it holds for star queries before expanding to all Berge-acyclic queries. Consider the following query,

$$Q_{STAR} = R(V_1, \dots, V_d) S_1(V_1) \dots S_d(V_d)$$

If M is the count tensor of the relation R and $a^{(V_i)}$ is the count tensor of S_i , in this case a simple non-increasing vector, then we can express the query size as follows,

$$|Q_{STAR}(D)| = \mathbf{M} \cdot \mathbf{a}^{(V_1)} \cdots \mathbf{a}^{(V_d)}$$

Given this notation, we consider part of Theorem 3.2 from [1].

Theorem A.3. [Thm. 3.2 from [1]] Let f_R be the set of degree sequences as above, and let V, C defined by (3)-(5). Then:

(1) We can define the value tensor as follows,

$$\forall \boldsymbol{m} \in [\boldsymbol{n}]: \qquad \qquad \boldsymbol{E}_{\boldsymbol{m}}^{F_{\boldsymbol{R}}} = \min\left(F_{\boldsymbol{R}, V_1}(m_1), \dots, F_{\boldsymbol{R}, V_d}(m_d)\right) \tag{6}$$

- (2) For any non-increasing vectors $\mathbf{a}^{(V_p)} \in \mathbb{R}_+^{[n_p]}$, p = 2, d, the vector $\mathbf{C}^{f_R} \cdot \mathbf{a}^{(V_2)} \cdots \mathbf{a}^{(V_d)}$ is in $\mathbb{R}_+^{[n_1]}$ and non-increasing.
- (3) For all count tensors M_R , and all non-increasing vectors $\mathbf{a}^{(X_1)} \in \mathbb{R}_+^{[n_1]}, \dots, \mathbf{a}^{(X_d)} \in \mathbb{R}_+^{[n_d]}$:

$$M_R \cdot a^{(V_1)} \cdots a^{(V_d)} \le C^{f_R} \cdot a^{(V_1)} \cdots a^{(V_d)}$$

$$\tag{7}$$

Directly implying,

$$|Q_{STAR}(D)| \le |Q_{STAR}(W(s(D)))| \tag{8}$$

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Let \hat{F}_R be an upper bound of F_R , i.e. $\hat{F}_{R,V}(i) \ge F_{R,V}(i) \ \forall \ V \in V_R$, i, and define $\hat{f}_{R,V} = \Delta_V \hat{F}_{R,V}$ and \hat{f}_R as the set of these degree sequences which, as specified in Theorem 3.1, must be non-increasing. Further, note that item 1 and item 1 relies only on the properties of the worst-case instance's inherent structure, so it immediately applies to C^{F_R} . Based on the above, we can prove the following lemma,

Lemma A.4. For all non-increasing vectors $\mathbf{a}^{(V_1)} \in \mathbb{R}^{[n_1]}_{\perp}, \dots, \mathbf{a}^{(V_d)} \in \mathbb{R}^{[n_d]}_{\perp}$:

$$C^{f_R} \cdot \boldsymbol{a}^{(V_1)} \cdots \boldsymbol{a}^{(V_d)} \le C^{\hat{F}_R} \cdot \boldsymbol{a}^{(V_1)} \cdots \boldsymbol{a}^{(V_d)}$$

$$\tag{9}$$

Directly implying,

$$|Q_{STAR}(W(s(D)))| \le |Q_{STAR}(W(\Delta \hat{S}))| \tag{10}$$

PROOF. Following the original proof of item 2, we begin by simplifying the problem using 1-0 vectors. In particular, let $b^{(m)} \in \mathbb{R}^n$ be the vector with m 1's followed by n-m 0's. Because the a^{V_i} are non-increasing integral vectors, they can be represented as a sum of 1-0 vectors, so it suffices to consider the case where each of them is a 1-0 vector. In this case, the problem description becomes,

$$C^{f_R} \cdot \boldsymbol{b}^{(m_1)} \cdots \boldsymbol{b}^{(m_d)} < C^{\hat{F}_R} \cdot \boldsymbol{b}^{(m_1)} \cdots \boldsymbol{b}^{(m_d)}$$

Multiplying against $b^{(m)}$ is the same as summing over the first m indices, so this can be alternatively expressed as,

$$\Sigma_{m_1} \dots \Sigma_{m_d} C^{\hat{f}_R} \leq \Sigma_{m_1} \dots \Sigma_{m_d} C^{\hat{f}_R}$$

Considering the definition of the value tensor $E_{m}^{f_{R}}$, we can rephrase this as follows where $m = (m_{1}, \dots, m_{d})$,

$$E_{\boldsymbol{m}}^{f_R} \leq E_{\boldsymbol{m}}^{\hat{f}_R}$$

Lastly, we insert the alternative definition of E^{f_R} provided in item 1 and the fact that each \hat{F}_{R,V_i} is an upper bound of F_{R,V_i} to prove the

$$\min(F_{R,V_1}(m_1), \dots, F_{R,V_d}(m_d)) \leq \min(\hat{F}_{R,V_1}(m_1), \dots, \hat{F}_{R,V_d}(m_d))$$

To prove that this can be extended to general queries, we rely on more theory from [1].

Theorem A.5. Implied by Thm. 4.2 of [1] If the following is true for a set of database instances, \mathcal{D} ,

- (1) For any non-increasing vectors $\mathbf{a}^{(V_p)} \in \mathbb{R}_+^{[n_p]}$, p = 2, d, the vector $\mathbf{C}^{f_R} \cdot \mathbf{a}^{(V_2)} \cdots \mathbf{a}^{(V_d)}$ is in $\mathbb{R}_+^{[n_1]}$ and non-increasing.
- (2) For all relations $R \in \mathcal{D}$ with count tensor M_R , and all non-increasing vectors $\boldsymbol{a}^{(X_1)} \in \mathbb{R}_+^{[n_1]}, \dots, \boldsymbol{a}^{(X_d)} \in \mathbb{R}_+^{[n_d]}$:

$$M_R$$
, and an incitation into each g received $a = 24 + \dots + 4 + \dots +$

Then, for any Berge-acyclic query, Q,

$$|Q(D)| \le |Q(W(s))| \quad \forall D \in \mathcal{D}$$
 (12)

The first immediately holds for $C^{\hat{F}_R}$ from Theorem A.3 while the latter holds for the set of all database instances, D, such that $D \models S$ due to Lemma A.4.

REFERENCES

[1] Kyle Deeds, Dan Suciu, Magda Balazinska, and Walter Cai. 2022. Degree Sequence Bound For Join Cardinality Estimation. CoRR abs/2201.04166 (2022). arXiv:2201.04166 https://arxiv.org/abs/2201.04166