

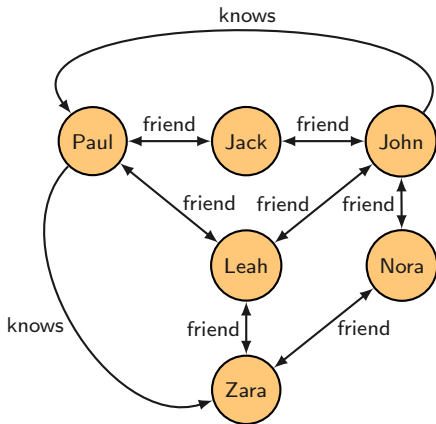
# A Polynomial-Time Approximation Algorithm for Counting Words Accepted by an NFA

Marcelo Arenas

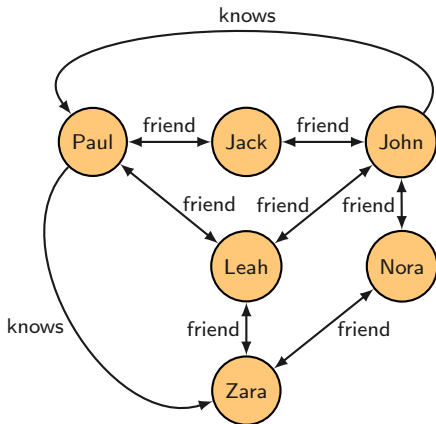
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Joint work with Luis Alberto Croquevielle, Rajesh Jayaram and Cristian Riveros

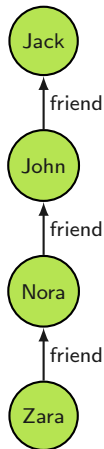
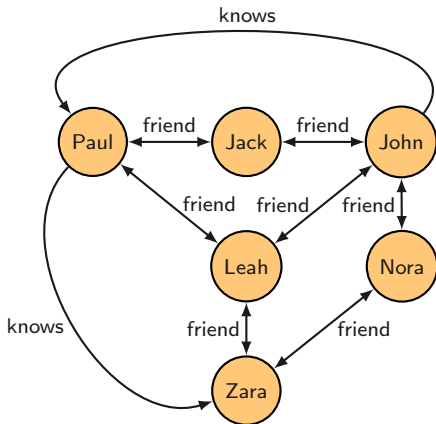
# A graph database $G$



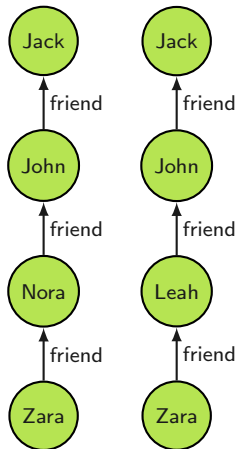
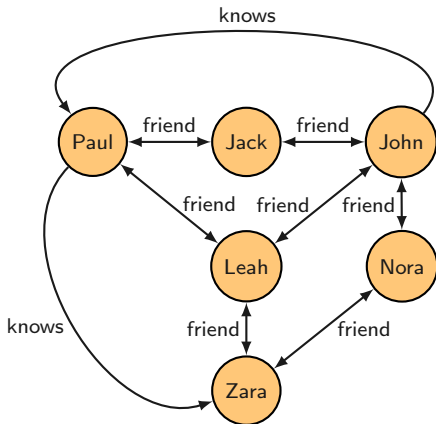
A query over  $G$ :  $(\text{friend} + \text{knows})^*$



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A query over  $G$ : (friend + knows)\*



# The length $n$ of paths as a parameter

Two fundamental problems:

- ▶ **COUNT( $G, r, n$ )**: count the number of paths  $p$  in  $G$  such that  $p$  conforms to regular expression  $r$  and the length of  $p$  is  $n$ 
  - ▶  $n$  is given in unary as  $0^n$
- ▶ **GEN( $G, r, n$ )**: generate uniformly at random a path  $p$  in  $G$  such that  $p$  conforms to  $r$  and the length of  $p$  is  $n$

# COUNT is #P-complete

Only approximate solutions are possible

- ▶ Best known approximations work in quasi-polynomial time

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Our goal is to construct an FPRAS  $\mathcal{B}$  for COUNT

- ▶ For every  $G, r, n$  and error  $\varepsilon \in (0, 1)$ :

$$\Pr\left(\left|\frac{\text{COUNT}(G, r, n) - \mathcal{B}(G, r, n, \varepsilon)}{\text{COUNT}(G, r, n)}\right| \leq \varepsilon\right) \geq \frac{3}{4}$$

- ▶  $\mathcal{B}$  works in time  $\text{poly}(\|G\|, \|r\|, n, \frac{1}{\varepsilon})$



# COUNT can be reduced to the following problem

- Input : An NFA  $A$ , a length  $n$  given in unary and  $\varepsilon \in (0, 1)$   
Output : Number of words  $w$  such that  $w \in \mathcal{L}(A)$  and  $|w| = n$

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$$A = (Q, \{0, 1\}, \Delta, I, F)$$

- ▶  $Q$  is a finite set of states
- ▶  $\Delta \subseteq Q \times \{0, 1\} \times Q$  is the transition relation
- ▶  $I \subseteq Q$  is a set of initial states
- ▶  $F \subseteq Q$  is a set of final (accepting) states

# The problem to solve

Assuming  $\mathcal{L}_n(A) = \mathcal{L}(A) \cap \{0, 1\}^n$

The task is to compute a number  $N$  that is a  $(1 \pm \varepsilon)$ -approximation of  $|\mathcal{L}_n(A)|$ :

$$(1 - \varepsilon)|\mathcal{L}_n(A)| \leq N \leq (1 + \varepsilon)|\mathcal{L}_n(A)|$$

Besides, number  $N$  has to be computed in time  $\text{poly}(m, n, \frac{1}{\varepsilon})$  with  $m = |Q|$

# First component: unroll automaton $A$

Construct  $A_{unroll}$  from  $A$ :

- ▶ for each state  $q \in Q$ , include copies  $q^0, q^1, \dots, q^n$  in  $A_{unroll}$
- ▶ for each transition  $(p, a, q) \in \Delta$  and  $i \in \{0, 1, \dots, n-1\}$ , include transition  $(p^i, a, q^{i+1})$  in  $A_{unroll}$

Besides, eliminate from  $A_{unroll}$  unnecessary states: each state  $q^i$  is reachable from an initial state  $p^0$  ( $p \in I$ )

## Second component: a sketch to be used in the estimation

Define  $\mathcal{L}(q^i)$  as the set of strings  $w$  such that there is a path from an initial state  $p^0$  to  $q^i$  labeled with  $w$

▶ Notice that  $|w| = i$

Besides, define for every  $X \subseteq Q$ :

$$\mathcal{L}(X^i) = \bigcup_{q \in X} \mathcal{L}(q^i)$$

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Then the task is to compute an estimation of  $|\mathcal{L}(F^n)|$

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$$\text{Let } \kappa = \left\lceil \frac{nm}{\varepsilon} \right\rceil$$

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Let  $\kappa = \lceil \frac{nm}{\varepsilon} \rceil$

We maintain for each state  $q^i$ :

- ▶  $N(q^i)$ : a  $(1 \pm \kappa^{-2})^i$ -approximation of  $|\mathcal{L}(q^i)|$
- ▶  $S(q^i)$ : a multiset of uniform samples from  $\mathcal{L}(q^i)$  of size  $2\kappa^7$

Data structure to be inductively computed:

$$\text{sketch}[i] = \{N(q^j), S(q^j) \mid 0 \leq j \leq i \text{ and } q \in Q\}$$



# The algorithm template

1. Construct  $A_{unroll}$  from  $A$
2. For each state  $q \in I$ , set  $N(q^0) = |\mathcal{L}(q^0)| = 1$  and  $S(q^0) = \mathcal{L}(q^0) = \{\lambda\}$
3. For each  $i = 1, \dots, n$  and state  $q \in Q$ :
  - (a) Compute  $N(q^i)$  given  $\text{sketch}[i - 1]$
  - (b) Sample polynomially many uniform elements from  $\mathcal{L}(q^i)$  using  $N(q^i)$  and  $\text{sketch}[i - 1]$ , and let  $S(q^i)$  be the multiset of uniform samples obtained
4. Return an estimation of  $|\mathcal{L}(F^n)|$  given  $\text{sketch}[n]$

# Computing an estimation $N(F^n)$ of $|\mathcal{L}(F^n)|$

We use notation  $N(X^i)$  for an estimation  $|\mathcal{L}(X^i)|$

- ▶ Such an estimation is not only needed in the last step of the algorithm, but also in the inductive construction of  $\text{sketch}[i]$

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But the following holds, given a linear order  $<$  on  $Q$ :

$$|\mathcal{L}(X^i)| = \sum_{p \in X} |\mathcal{L}(p^i) \setminus \bigcup_{q \in X: q < p} \mathcal{L}(q^i)|$$

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So we will use the following approximation:

$$N(X^i) = \sum_{p \in X} N(p^i) \frac{|S(p^i) \setminus \bigcup_{q \in X: q < p} \mathcal{L}(q^i)|}{|S(p^i)|}$$

# Computing an estimation $N(X^i)$ of $|\mathcal{L}(X^i)|$

$N(X^i)$  can be computed in polynomial time in the size of sketch[ $i$ ]

- ▶  $S(p^i) \setminus \bigcup_{q \in X: q < p} \mathcal{L}(q^i)$  is constructed by checking for each  $w \in S(p^i)$  whether  $w$  is not in  $\mathcal{L}(q^i)$  for every  $q \in X$  with  $q < p$

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What guarantees that  $N(X^i)$  is a good estimation of  $|\mathcal{L}(X^i)|$ ?



# The main property to maintain

$\mathcal{E}(i)$  holds if for every  $p \in Q$  and  $X \subseteq Q$ :

$$\left| \frac{|\mathcal{L}(p^i) \setminus \bigcup_{q \in X} \mathcal{L}(q^i)|}{|\mathcal{L}(p^i)|} - \frac{|S(p^i) \setminus \bigcup_{q \in X} \mathcal{L}(q^i)|}{|S(p^i)|} \right| < \frac{1}{\kappa^3}$$

# The use of the main property

## Proposition

*If  $\mathcal{E}(i)$  holds and  $N(p^i)$  is a  $(1 \pm \kappa^{-2})^i$ -approximation of  $|\mathcal{L}(p^i)|$  for every  $p \in Q$ , then  $N(X^i)$  is a  $(1 \pm \kappa^{-2})^{i+1}$ -approximation of  $|\mathcal{L}(X^i)|$  for every  $X \subseteq Q$*

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# The use of the main property

For each state  $p \in Q$  and  $b = 0, 1$ , define:

$$R_b(p^1) = \{q^0 \mid (q^0, b, p^1) \text{ is a transition in } A_{unroll}\}$$

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Then  $\mathcal{L}(p^1) = \mathcal{L}(R_0(p^1)) \cdot \{0\} \uplus \mathcal{L}(R_1(p^1)) \cdot \{1\}$

▶ So that  $|\mathcal{L}(p^1)| = |\mathcal{L}(R_0(p^1))| + |\mathcal{L}(R_1(p^1))|$

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Hence, given that  $N(R_b(p^1))$  is a  $(1 \pm \kappa^{-2})$ -approximation of  $|\mathcal{L}(R_b(p^1))|$  for  $b = 0, 1$ :

$N(R_0(p^1)) + N(R_1(p^1))$  is a  $(1 \pm \kappa^{-2})$ -approximation of  $N(p^1)$

## The use of the main property: a summary

$\mathcal{E}(0)$  holds and  $N(p^0)$  is a  $(1 \pm \kappa^{-2})^0$ -approximation of  $|\mathcal{L}(p^0)|$  for every  $p \in Q$



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$N(p^2) = N(R_0(p^2)) + N(R_1(p^2))$  is a  $(1 \pm \kappa^{-2})^2$ -approximation of  $N(p^2)$  for every  $p \in Q$

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# The final result

## Proposition

*If  $\mathcal{E}(i)$  holds for every  $i \in \{0, 1, \dots, n\}$ , then  $N(F^n)$  is a  $(1 \pm \varepsilon)$ -approximation of  $|\mathcal{L}(F^n)|$*



# The final result

## Proposition

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The issue then is to maintain property  $\mathcal{E}(i)$

- ▶ Multisets  $S(q^i)$  of uniform samples play a central role on this

# Sampling from a state

We need to construct the multiset  $S(q^i)$  of uniform samples

Recall that:

- ▶  $S(q^i)$  contains  $2\kappa^7$  words from  $\mathcal{L}(q^i)$
- ▶  $S(q^i)$  is computed assuming that  $N(q^i)$  and  $\text{sketch}[i - 1] = \{N(q^j), S(q^j) \mid 0 \leq j \leq i - 1\}$  have already been constructed

# To recall

1. Construct  $A_{unroll}$  from  $A$
2. For each state  $q \in I$ , set  $N(q^0) = |\mathcal{L}(q^0)| = 1$  and  $S(q^0) = \mathcal{L}(q^0) = \{\lambda\}$
3. For each  $i = 1, \dots, n$  and state  $q \in Q$ :
  - (a) Compute  $N(q^i)$  given  $\text{sketch}[i - 1]$
  - (b) Sample polynomially many uniform elements from  $\mathcal{L}(q^i)$  using  $N(q^i)$  and  $\text{sketch}[i - 1]$ , and let  $S(q^i)$  be the multiset of uniform samples obtained
4. Return an estimation of  $|\mathcal{L}(F^n)|$  given  $\text{sketch}[n]$

# Sampling from $q^i$

To generate a sample in  $\mathcal{L}(q^i)$ , we construct a sequence  $w^i, w^{i-1}, \dots, w^1, w^0$  such that

- ▶  $w^i = \lambda$
- ▶  $w^j = b_j w^{j+1}$  with  $b_j \in \{0, 1\}$
- ▶  $w^0 \in \mathcal{L}(q^i)$

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To choose  $w^{i-1} = b w^i$ , construct for  $b = 0, 1$ :

$$P_b^i = \{p^{i-1} \mid (p^{i-1}, b, q^i) \text{ is a transition in } A_{\text{unroll}}\}$$

# Sampling from $q^i$

$P_0^i$  and  $P_1^i$  are sets of states at layer  $i - 1$

We can use the following estimations:

$$N(X^{i-1}) = \sum_{p \in X} N(p^{i-1}) \frac{|S(p^{i-1}) \setminus \bigcup_{q \in X: q < p} \mathcal{L}(q^{i-1})|}{|S(p^{i-1})|}$$

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We choose  $b \in \{0, 1\}$  with probability:

$$\frac{N(P_b^i)}{N(P_0^i) + N(P_1^i)}$$

## We could have started from a set of states

The previous procedure works for every set of states  $P^i$ :

$$P_b^i = \{p^{i-1} \mid \exists r^i \in P^i : (p^{i-1}, b, r^i) \text{ is a transition in } A_{\text{unroll}}\}$$

In particular, we applied the procedure for  $P^i = \{q^i\}$



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The following recursive procedure summarizes the previous idea:

**Sample** $(i, \{q^i\}, \lambda, \varphi_0)$

It uses sets of states  $P^i = \{q^i\}, P^{i-1}, \dots, P^1, P^0$  and an initial probability  $\varphi_0$

# The sampling algorithm

**Sample**( $j, P^j, w^j, \varphi$ )

1. If  $j = 0$ , then with probability  $\varphi$  return  $w^0$ , otherwise return **fail**
2. Compute  $P_b^j = \{p^{j-1} \mid \exists r^j \in P^j : (p^{j-1}, b, r^j) \text{ is a transition in } A_{\text{unroll}}\}$  for  $b = 0, 1$
3. Choose  $b \in \{0, 1\}$  with probability  $p_b = \frac{N(P_b^j)}{N(P_0^j) + N(P_1^j)}$
4. Set  $P^{j-1} = P_b^j$  and  $w^{j-1} = bw^j$
5. Return **Sample**( $j - 1, P^{j-1}, w^{j-1}, \frac{\varphi}{p_b}$ )

# The key observation

Let  $x = x_1 \cdots x_i$  be word in  $\mathcal{L}(q^i)$

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We have that:

**Pr**(the output of **Sample** is  $x$ )

= **Pr**( $w^0 = x \wedge$  the last call to **Sample** does not fail)

= **Pr**(the last call to **Sample** does not fail |  $w^0 = x$ )  $\cdot$  **Pr**( $w^0 = x$ )

$$= \left( \left( \prod_{j=1}^i \frac{N(P_{x_j}^j)}{N(P_0^j) + N(P_1^j)} \right)^{-1} \cdot \varphi_0 \right) \cdot \left( \prod_{j=1}^i \frac{N(P_{x_j}^j)}{N(P_0^j) + N(P_1^j)} \right)$$

$$= \varphi_0$$

# The value of the initial probability $\varphi_0$

## Proposition

Assume that  $\mathcal{E}(j)$  holds for each  $j < i$ . If  $w$  is the output of **Sample** $(i, \{q^i\}, \lambda, \frac{e^{-5}}{N(q^i)})$ , then

- ▶  $\varphi \in (0, 1)$  in every recursive call to **Sample**
- ▶  $\Pr(w = \text{fail}) \leq 1 - e^{-9}$
- ▶  $\Pr(w = x) = \frac{e^{-5}}{N(q^i)}$  for every  $x \in \mathcal{L}(q^i)$

# The value of the initial probability $\varphi_0$

## Proposition

Assume that  $\mathcal{E}(j)$  holds for each  $j < i$ . If  $w$  is the output of  $\mathbf{Sample}(i, \{q^i\}, \lambda, \frac{e^{-5}}{N(q^i)})$ , then

- ▶  $\varphi \in (0, 1)$  in every recursive call to **Sample**
- ▶  $\Pr(w = \mathbf{fail}) \leq 1 - e^{-9}$
- ▶  $\Pr(w = x) = \frac{e^{-5}}{N(q^i)}$  for every  $x \in \mathcal{L}(q^i)$

Hence, conditioned on not failing,  $\mathbf{Sample}(i, \{q^i\}, \lambda, \frac{e^{-5}}{N(q^i)})$  returns a uniform sample from  $\mathcal{L}(q^i)$

## The last step: bounding the probability of breaking the main assumption

Recall that  $\mathcal{E}(i)$  holds if for every  $q \in Q$  and  $X \subseteq Q$ :

$$\left| \frac{|\mathcal{L}(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|\mathcal{L}(q^i)|} - \frac{|S(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|S(q^i)|} \right| < \frac{1}{\kappa^3}$$

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We know that  $\mathcal{E}(0)$  holds. We need to compute a lower bound for:

$$\Pr\left(\bigwedge_{j=0}^n \mathcal{E}(j)\right)$$



# Bounding the probability of breaking $\mathcal{E}(i)$

Assume that  $\bigwedge_{j=0}^{i-1} \mathcal{E}(j)$  holds

Let  $q \in Q$  and  $S(q^i)$  be a multiset of  $2\kappa^7$  samples from  $\mathcal{L}(q^i)$  computed by calling **Sample** $(i, \{q^i\}, \lambda, \frac{e^{-5}}{N(q^i)})$

- ▶ Each element of  $S(q^i)$  is obtained by repeatedly calling **Sample** until the output is different from **fail**

Assume that  $S(q^i) = \{w_1, \dots, w_t\}$  with  $t = 2\kappa^7$

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We have that:

$$\begin{aligned} \mathbb{E}[Y_i] &= \frac{|\mathcal{L}(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|\mathcal{L}(q^i)|} \\ \sum_{j=1}^t Y_j &= |S(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)| \\ t &= |S(q^i)| \end{aligned}$$

By using Hoeffding's inequality

$$\Pr\left(\left|\frac{|S(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|S(q^i)|} - \frac{|\mathcal{L}(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|\mathcal{L}(q^i)|}\right| \geq \frac{1}{\kappa^3} \mid \bigwedge_{j=0}^{i-1} \mathcal{E}(j)\right) \leq 2e^{-4\kappa}$$

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By taking the union bound:

$$\Pr\left(\exists q \in Q \exists X \subseteq Q \left|\frac{|S(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|S(q^i)|} - \frac{|\mathcal{L}(q^i) \setminus \bigcup_{p \in X} \mathcal{L}(p^i)|}{|\mathcal{L}(q^i)|}\right| \geq \frac{1}{\kappa^3} \mid \bigwedge_{j=0}^{i-1} \mathcal{E}(j)\right) \leq 2e^{-2\kappa}$$

# The conclusion

Rewriting the previous result:

$$\Pr\left(\mathcal{E}(i) \mid \bigwedge_{j=0}^{i-1} \mathcal{E}(j)\right) \geq 1 - e^{-2\kappa}$$

We conclude that:

$$\Pr\left(\bigwedge_{j=0}^n \mathcal{E}(j)\right) \geq 1 - e^{-\kappa}$$

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3. Remove each state  $q^i$  from  $A_{unroll}$  that is not reachable from an initial state in  $I^0$
4. For each  $q^0 \in I^0$ , set  $N(q^0) = 1$  and  $S(q^0) = \{\lambda\}$

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    - 5.3.1 Run **Sample** $(i, \{q^i\}, \lambda, \frac{e^{-5}}{N(q^i)})$  until it returns  $w \neq \mathbf{fail}$ , and at most  $c(\kappa) \in \Theta(\log(\kappa))$  times



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6. Return  $N(F^n)$  as an estimation of  $|\mathcal{L}_n(A)|$

# The complete algorithm: final comments

The probability that the algorithm returns a wrong estimate is at most  $\frac{1}{4}$

▶ Considering  $c(\kappa) = \lceil \frac{2 + \log(4) + 8 \log(\kappa)}{\log(1 - e^{-9})^{-1}} \rceil$

The algorithm runs in time  $\text{poly}(m, n, \frac{1}{\epsilon})$

# Final remarks

- ▶ The algorithm also provides a randomized polynomial-time algorithm for GEN
  - ▶ Such an algorithm can also be obtained from [Jerrum, Valiant & Vazirani 1986]
- ▶ COUNT is SpanL-complete under parsimonious reductions. We conclude that each function in SpanL admits an FPRAS
  - ▶ SpanL is the class of functions computable as  $|S|$ , where  $S$  is the set of output values returned by an NL Turing machine

The complete version of the paper can be found at  
<https://arxiv.org/abs/1906.09226>

Thanks!