

Improved FPTAS *for* Multi-spin Systems

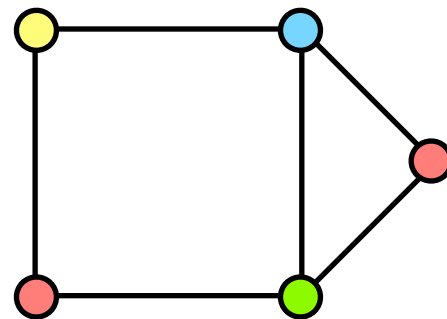
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Microsoft Research Asia

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Nanjing University

presented by:
Sangxia Huang
KTH

Colorings

instance: undirected $G(V,E)$ with max-degree $\leq \Delta$



q colors:

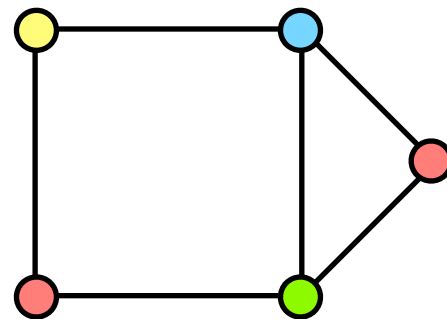


goal: counting the number of proper q -colorings for G

- exact counting is **#P-hard**
- when $q < \Delta$, decision of existence is **NP-hard**

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approximately counting the number of proper q -colorings for G when $q \geq \alpha\Delta + \beta$

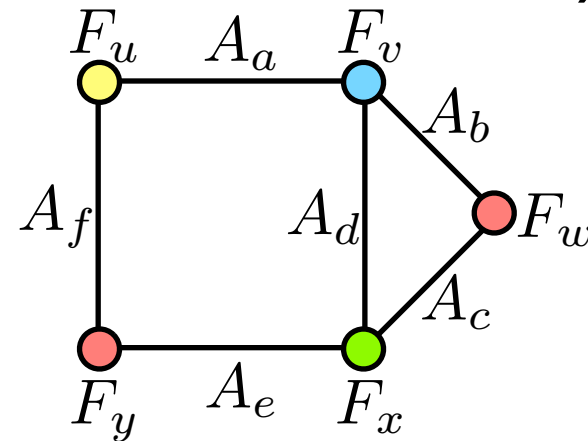
equivalent to sampling an almost uniform random q -coloring

Spin System

(pairwise Markov random field)

instance:

- undirected $G(V,E)$;
- q **states**: $[q]$;
- each edge $e \in E$ associated with an **activity**:



a symmetric nonnegative $q \times q$ matrix

$$A_e : [q] \times [q] \rightarrow \mathbb{R}_{\geq 0}$$

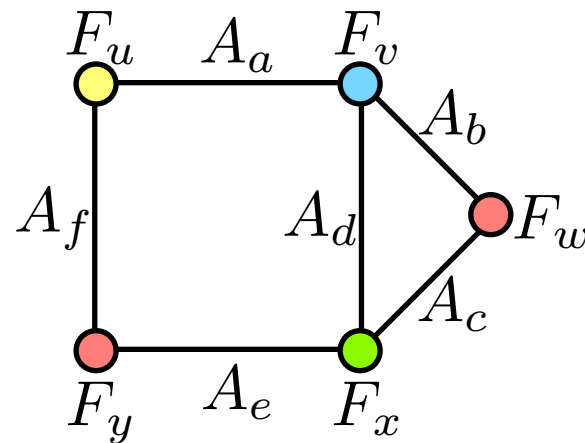
- each vertex $v \in V$ associated with an **external field**:

a nonnegative q -vector $F_v : [q] \rightarrow \mathbb{R}_{\geq 0}$

goal: computing the **partition function**:

$$Z = \sum_{\mathbf{x} \in [q]^V} \prod_{e=uv \in E} A_e(x_u, x_v) \prod_{v \in V} F_v(x_v)$$

Spin System



$$A_e : [q] \times [q] \rightarrow \mathbb{R}_{\geq 0}$$

$$F_v : [q] \rightarrow \mathbb{R}_{\geq 0}$$

partition function: count the # of solutions to an CSP

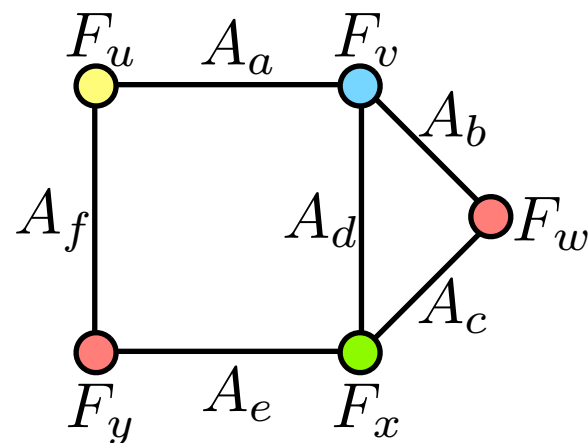
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enumerate all
configurations

binary constraints

unary constraints

Spin System



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coloring:

$$A = \begin{bmatrix} 0 & & & & \\ & 0 & & & \\ & & 1 & & \\ & 1 & & \ddots & \\ & & & \ddots & 0 \end{bmatrix}$$

$$F = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

Examples of Spin systems

- 2-spin: $q=2$
 - *hardcore* model (independent set), *Ising* model, etc.

- **multi-spin:** general q

- **coloring:**

$$A = \begin{bmatrix} 0 & & & \\ & 0 & & \\ & & 1 & \\ & 1 & & 0 \end{bmatrix} \quad F = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$$

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- **coloring**: $A = \begin{bmatrix} 0 & & & \\ & 0 & & \\ & & 1 & \\ & 1 & & 0 \end{bmatrix} \quad F = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$
- **Potts model**: **inverse temperature** β

$$A = \begin{bmatrix} e^\beta & & & \\ & e^\beta & & \\ & & 1 & \\ & 1 & \ddots & \\ & & & e^\beta \end{bmatrix} \quad \text{arbitrary } F$$

when $\beta = -\infty$ and $F = (1, 1, \dots, 1)$, it is coloring

Results

sufficient conditions for FPTAS for classes of spin systems

- **coloring**: $q \geq \alpha\Delta + \beta$
 - **randomized** algorithms: by simulating a random walk (the Glauber dynamics) over colorings
 - $\alpha=11/6$ (Jerrum'95 \rightarrow Bublely-Dyer'97 \rightarrow Vigoda'99)
 - **deterministic** algorithms: by exploiting the **correlation decay** (**spatial mixing**) property
 - $\alpha \approx 2.8432$ (Gamarnik-Katz'07)
 - just correlation decay (no FPTAS): $\alpha \approx 1.763$ (Goldberg-Martin-Paterson'05, Gamarnik-Katz-Misra'12)

this paper: deterministic FPTAS for $\alpha \approx 2.58071$

Results

sufficient conditions for FPTAS for classes of spin systems

- general **multi-spin system**:

in terms of $c = \max_{\substack{e \in E \\ w, x, y, z \in [q]}} \frac{A_e(x, y)}{A_e(w, z)}$

- Gamarnik-Katz'07: $(c^\Delta - c^{-\Delta})\Delta q^\Delta < 1$

this paper: $3\Delta(c^\Delta - 1) \leq 1$

an exponential improvement!

- on **Potts model** (with inverse temperature β):

it implies: $3\Delta(e^{|\beta|} - 1) \leq 1$

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- confirming the conjecture of $|\beta| = O\left(\frac{1}{\Delta}\right)$ in [GK'07]
- asymptotically matching the $e^\beta < 1 - \frac{q}{\Delta}$ inapproximability bound for $\beta < 0$ in [Galanis-Stefankovic-Vigoda'13]

The standard first step:
reducing to the computing of marginal probability

$$Z = \sum_{\mathbf{x} \in [q]^V} \prod_{e=uv \in E} A_e(x_u, x_v) \prod_{v \in V} F_v(x_v)$$

for any configuration $\mathbf{x} \in [q]^V$

Gibbs measure:

$$\mathbb{P}[\mathbf{X} = \mathbf{x}] = \frac{\prod_{e=uv \in E} A_e(x_u, x_v) \prod_{v \in V} F_v(x_v)}{Z}$$

marginal probability: $\mathbb{P}[X_v = x_v]$

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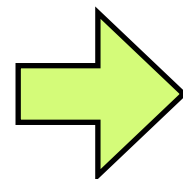
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marginal probability: $\mathbb{P}[X_v = x_v]$

Jerrum-Valiant-Vazirani'86

for **self-reducible** class of spin-systems:

efficient approximation
of marginal probability
(with additive error)

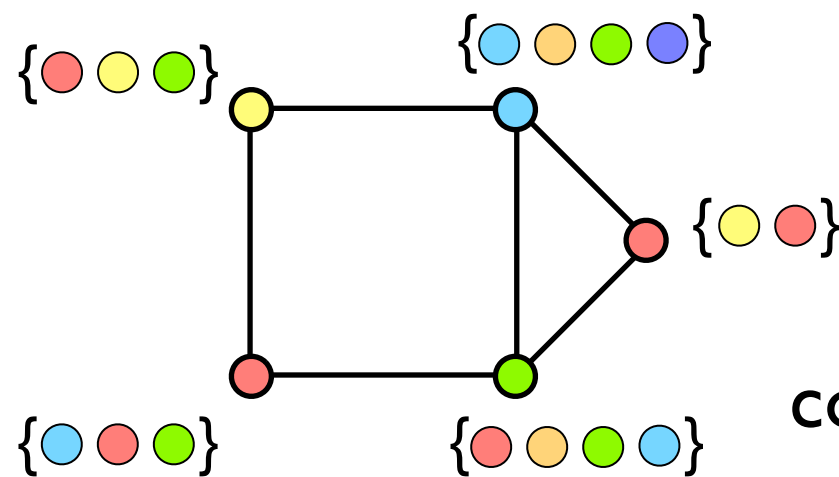


FPTAS for Z

The standard first step:

- **self-reducible**: general spin systems, Potts models
- **not self-reducible**: coloring
- **self-reducible** superclass of coloring: **list-coloring**

instance: undirected $G(V,E)$



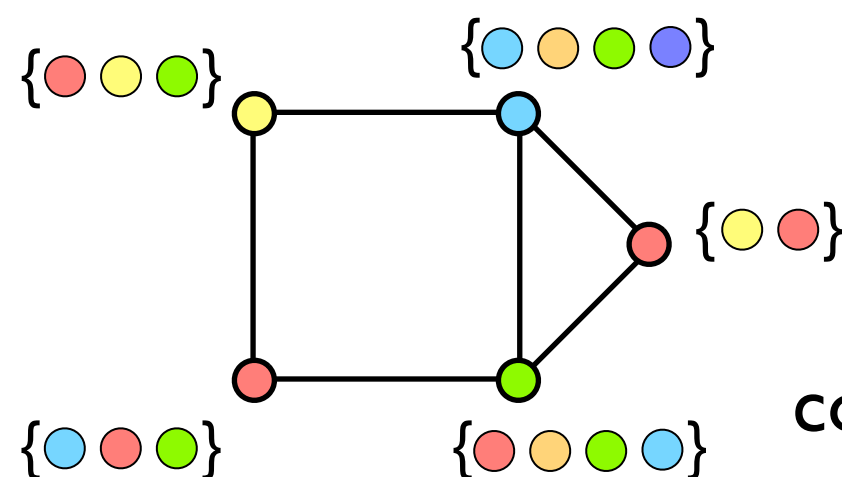
each vertex v associated
with a list L_v of colors
allowed to use on v

coloring \subset list-coloring \subset Potts \subset multi-spin
self-reducible

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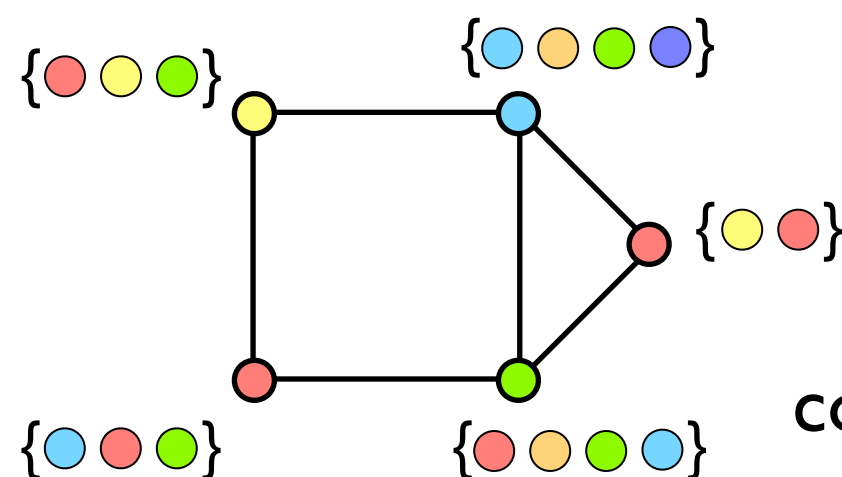
new goal:

for $\left\{ \begin{array}{l} \text{multi-spin system} \\ \text{Potts model} \\ \text{list-coloring} \end{array} \right\}$ approximate the **marginal**
 $\mathbb{P}[X_v = x]$ (with additive error)

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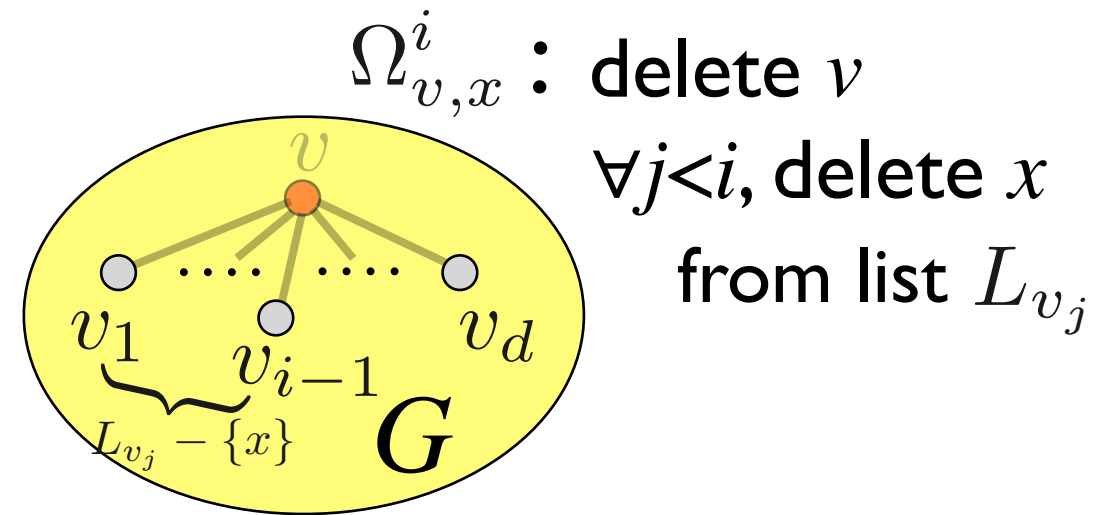
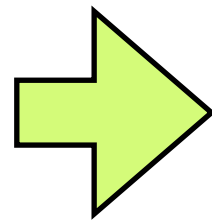
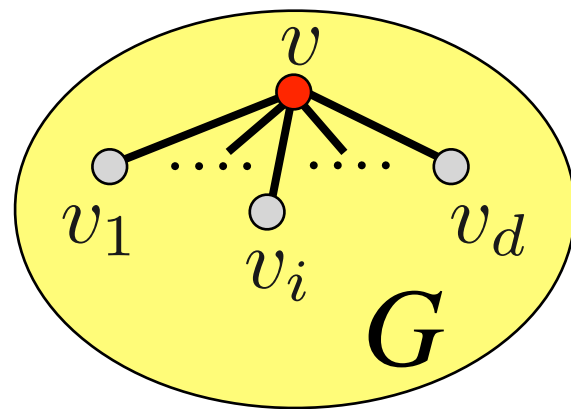
classic way: random walk new way: **correlation decay**

Recursion for List-Coloring

list-coloring instance Ω

v 's neighbors: v_1, v_2, \dots, v_d

color: x



Gamarnik-Katz'07:

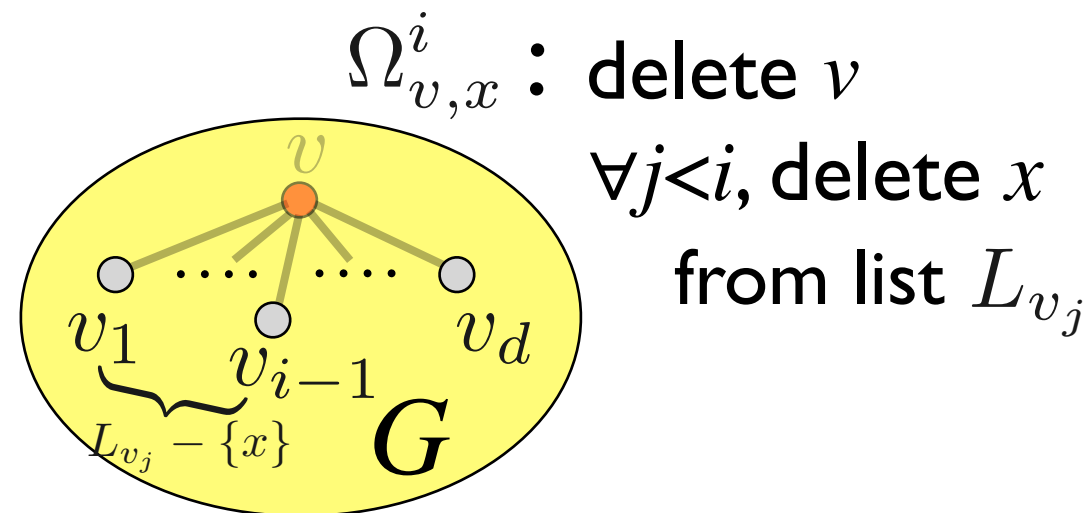
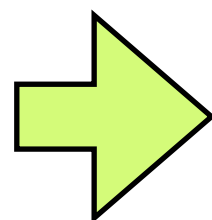
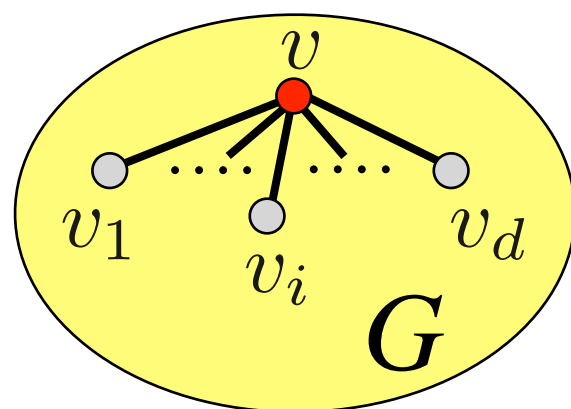
$$\begin{aligned} \mathbb{P}_{\Omega}(X_v = x) &= \frac{\prod_{i=1}^d \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} \neq x)}{\sum_{y \in L_v} \prod_{i=1}^d \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} \neq y)} \\ &= \frac{\prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = x)\right)}{\sum_{y \in L_v} \prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = y)\right)} \end{aligned}$$

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telescopic products

$$= \frac{\prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = x)\right)}{\sum_{y \in L_v} \prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = y)\right)}$$

Recursion for general multi-spin system

a natural generalization of list-coloring:

multi-spin system Ω

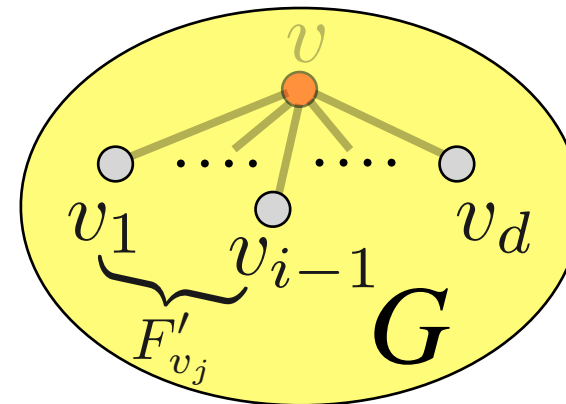
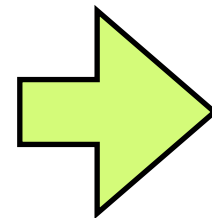
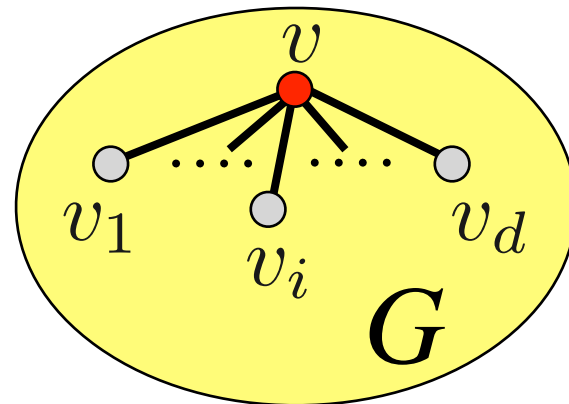
v 's neighbors: v_1, v_2, \dots, v_d

state: x

$\Omega_{v,x}^i$: delete v

$\forall j < i$, new external field

$$F'_{v_j}(y) = A_{vv_j}(x, y) F_{v_j}(y)$$



augmented
by edge
activity

$$\mathbb{P}_{\Omega}(X_v = x) = \frac{F_v(x) \prod_{i=1}^d \left(A_{vv_i}(x, x) - \sum_{z \neq x} (A_{vv_i}(x, x) - A_{vv_i}(x, z)) \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = z) \right)}{\sum_{y \in [q]} F_v(y) \prod_{i=1}^d \left(A_{vv_i}(y, y) - \sum_{z \neq y} (A_{vv_i}(y, y) - A_{vv_i}(y, z)) \mathbb{P}_{\Omega_{v,y}^i}(X_{v_i} = z) \right)}$$

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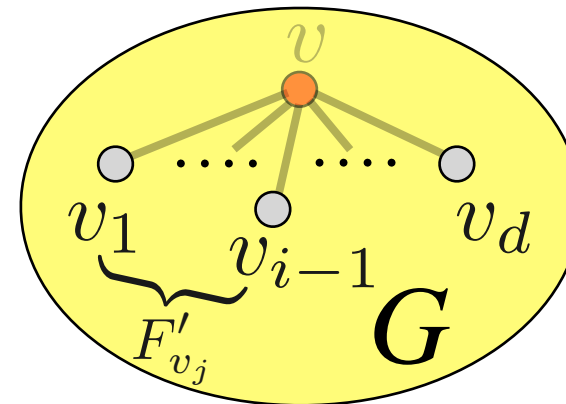
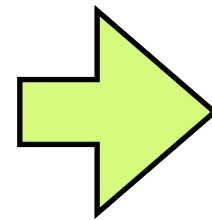
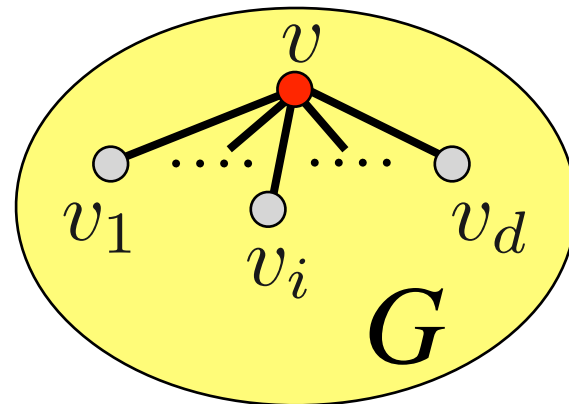
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for list-coloring: special case

$$\mathbb{P}_{\Omega}(X_v = x) = \frac{\prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = x) \right)}{\sum_{y \in L_v} \prod_{i=1}^d \left(1 - \mathbb{P}_{\Omega_{v,x}^i}(X_{v_i} = y) \right)}$$

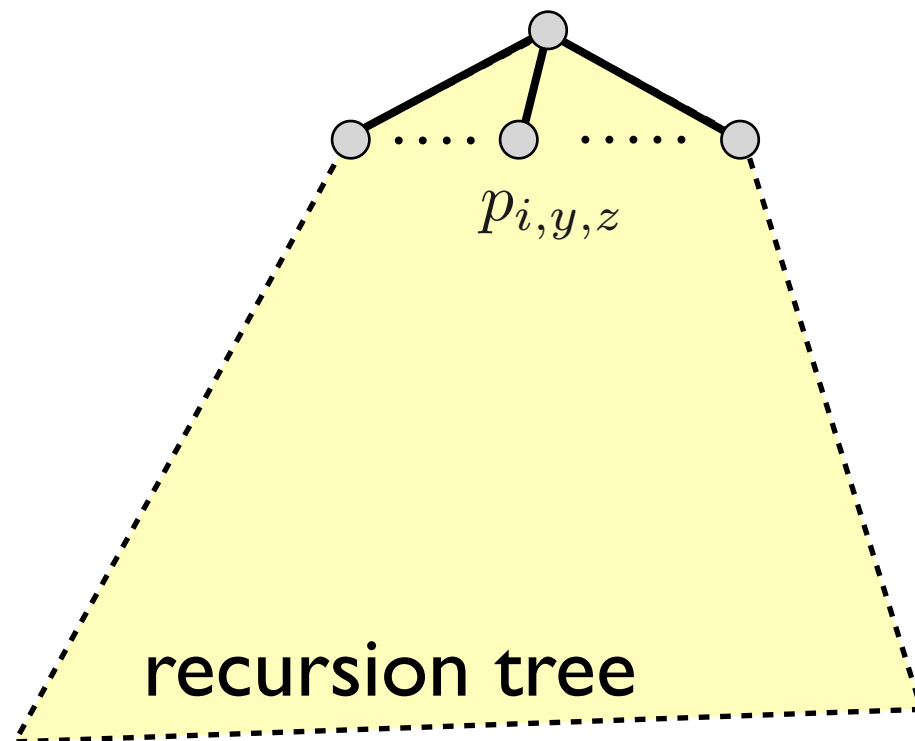
Correlation Decay

vector $\mathbf{p} = (p_{i,y,z})_{1 \leq i \leq d; y,z \in [q]; y \neq z}$ where $p_{i,y,z} = \mathbb{P}_{\Omega_{v,y}^i}(X_{v_i} = z)$

$$f(\mathbf{p}) = \frac{F_v(x) \prod_{i=1}^d \left(A_{vv_i}(x,x) - \sum_{z \neq x} \left(A_{vv_i}(x,x) - A_{vv_i}(x,z) \right) p_{i,x,z} \right)}{\sum_{y \in [q]} F_v(y) \prod_{i=1}^d \left(A_{vv_i}(y,y) - \sum_{z \neq y} \left(A_{vv_i}(y,y) - A_{vv_i}(y,z) \right) p_{i,y,z} \right)}$$

$$\mathbb{P}_{\Omega}[X_v = x] = f(\mathbf{p})$$

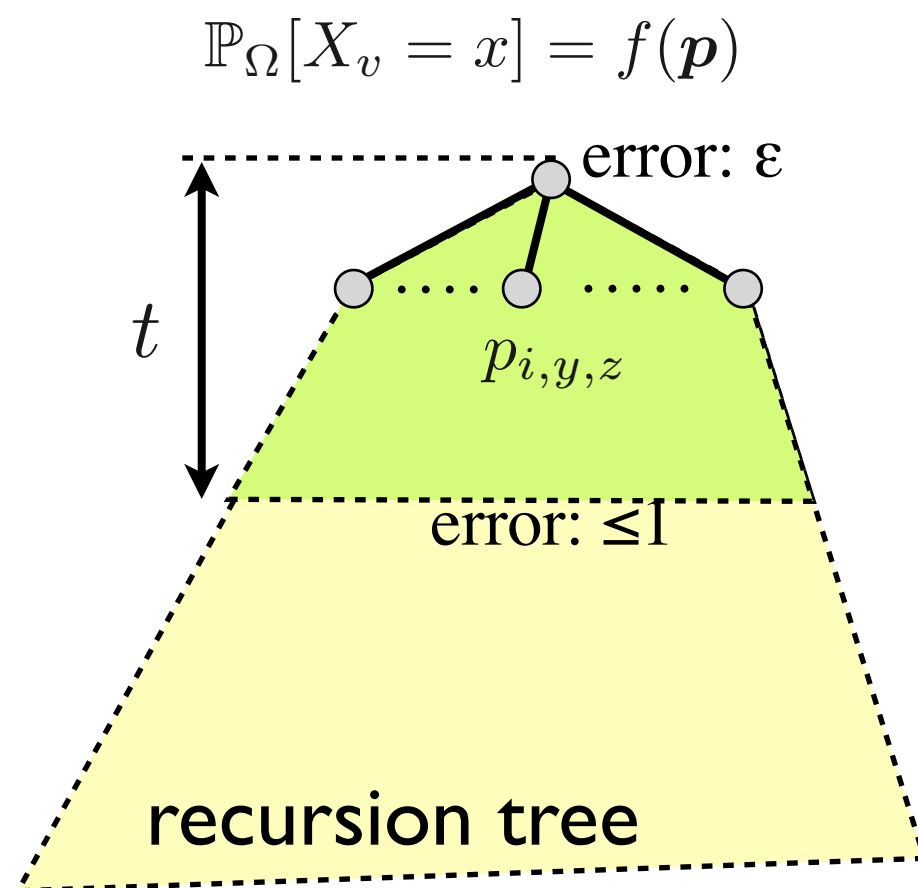
- an exponential-time exact algorithm



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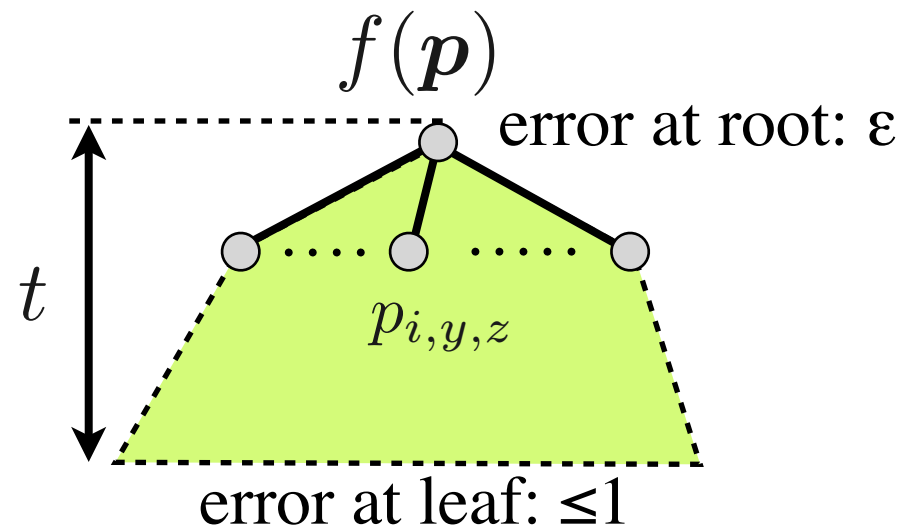


- an exponential-time exact algorithm
- **truncation:**
 - compute up-to level t
 - use arbitrary estimation at level t

correlation decay:

error at root $\varepsilon = \exp(-\Omega(t))$

Correlation Decay



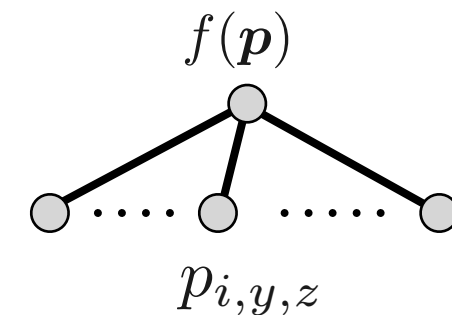
correlation decay:

error at root $\epsilon = \exp(-\Omega(t))$

if running the recursion up-to level t

a sufficient condition: at any step

$$\kappa \triangleq \sum_{i,y,z} \left| \frac{\partial f(p)}{\partial p_{i,y,z}} \right| < 1 \quad (\text{stepwise decay})$$

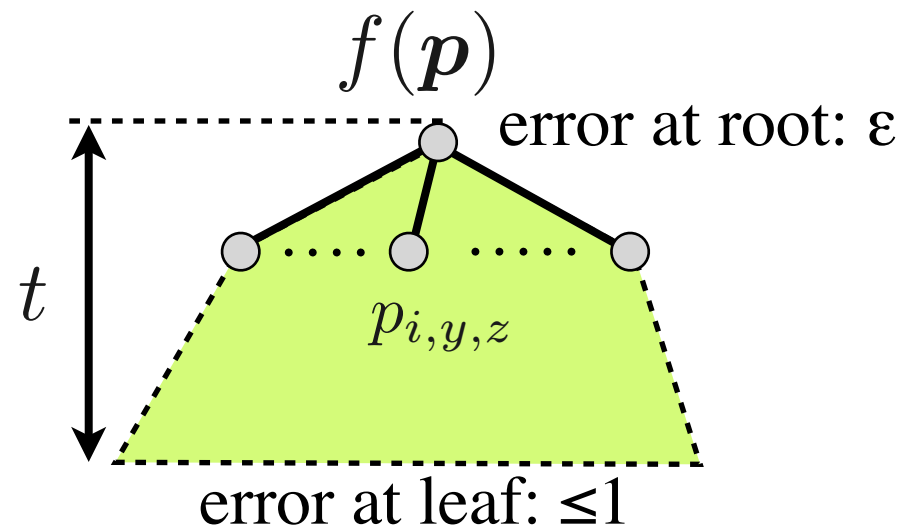


then due to the Mean Value Theorem

$$\epsilon \leq \sum_{i,y,z} \left| \frac{\partial f(p)}{\partial p_{i,y,z}} \right| \epsilon_{i,y,z} \leq \kappa \cdot \max_{i,y,z} \epsilon_{i,y,z}$$

induction!

Correlation Decay



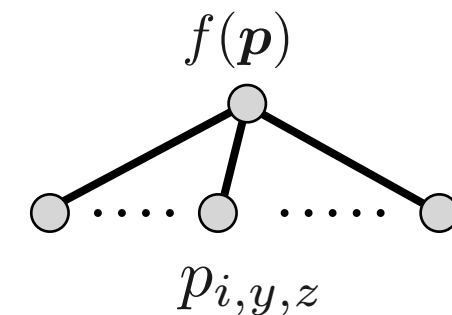
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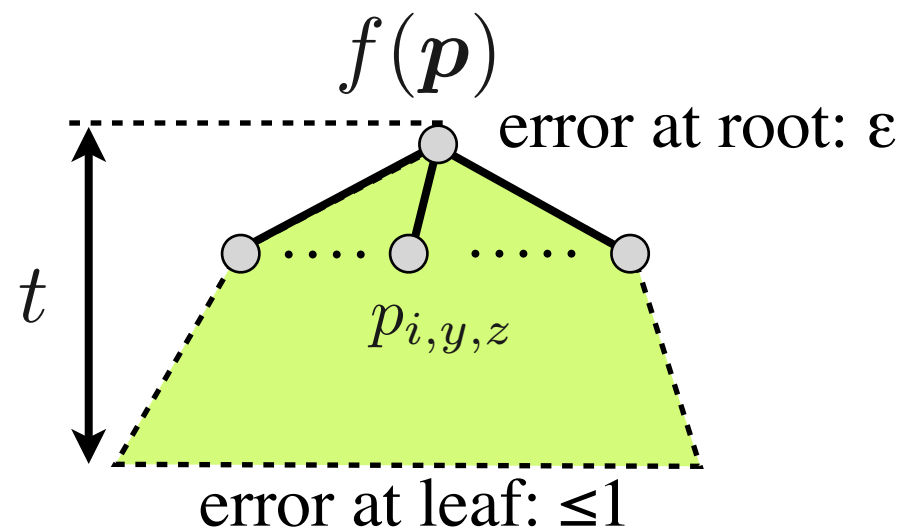


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induction!

Correlation Decay



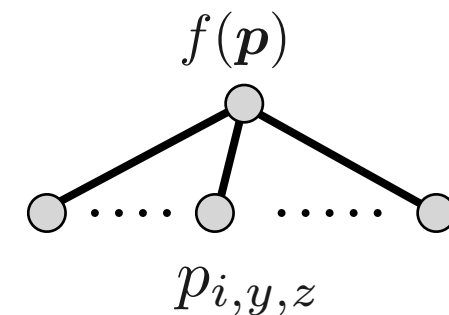
correlation decay:

$$\text{error at root } \epsilon = \exp(-\Omega(t))$$

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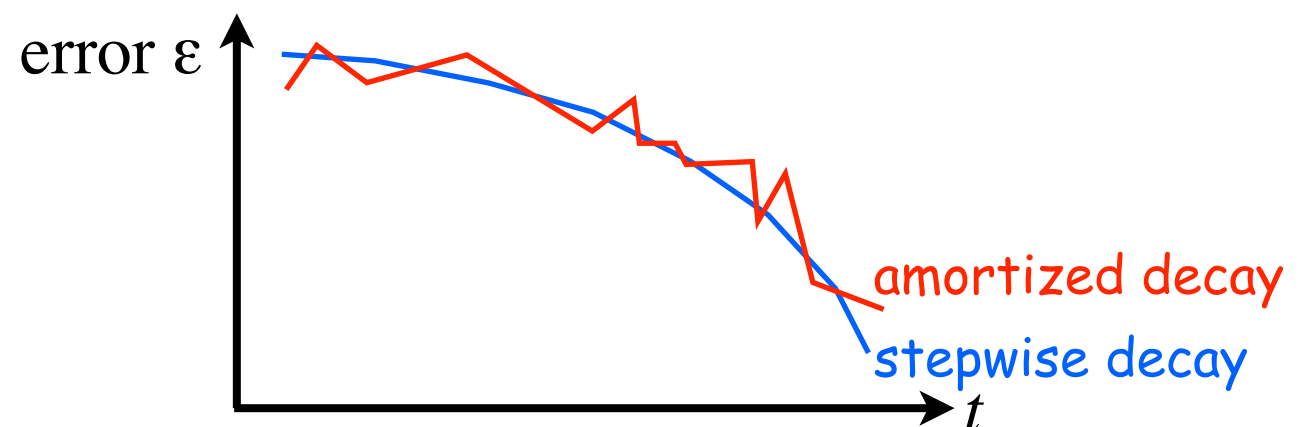


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induction!

**amortized behavior
correlation decay?**

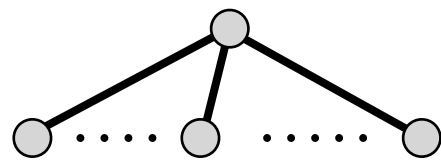


The Potential Method

original:

$$p = f(\vec{p})$$

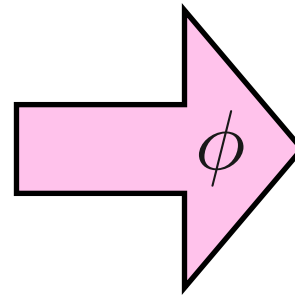
error ϵ



$$p_{i,y,z}$$

error $\epsilon_{i,y,z}$

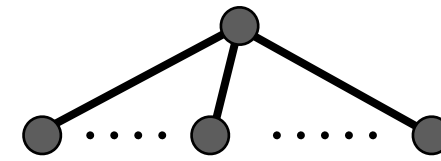
$$\xi = \phi(p)$$



potential:

$$\xi = g(\vec{\xi})$$

error δ

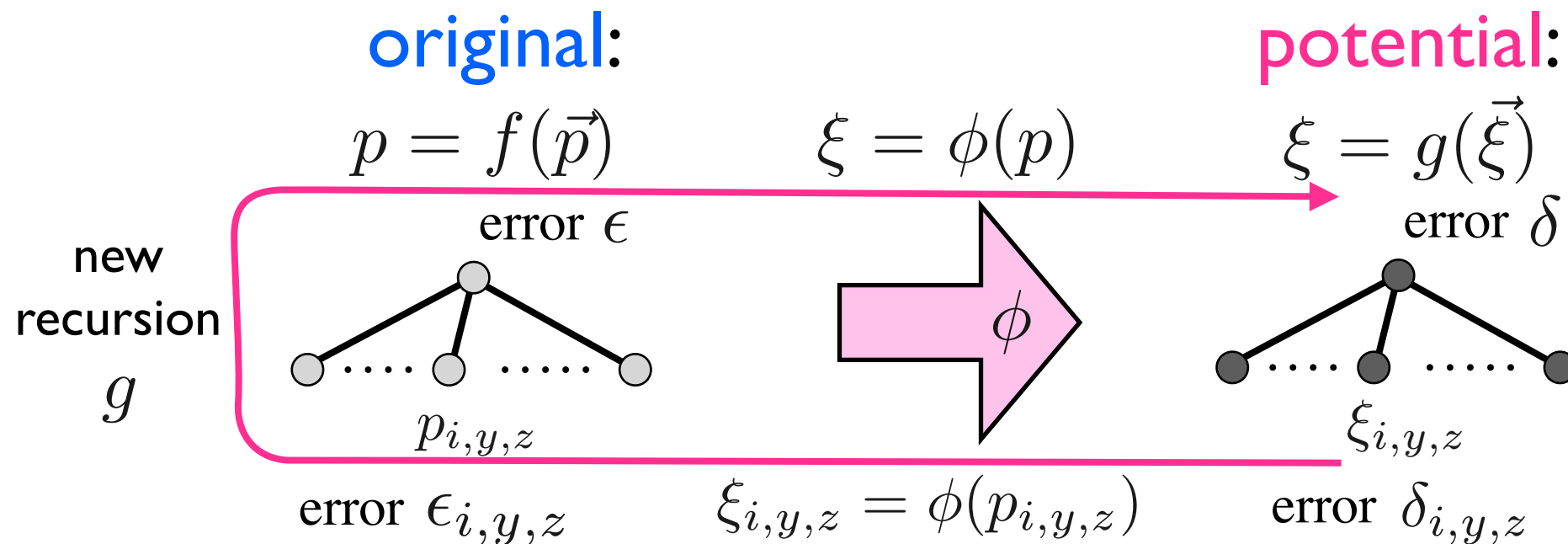


$$\xi_{i,y,z}$$

error $\delta_{i,y,z}$

$$\xi_{i,y,z} = \phi(p_{i,y,z})$$

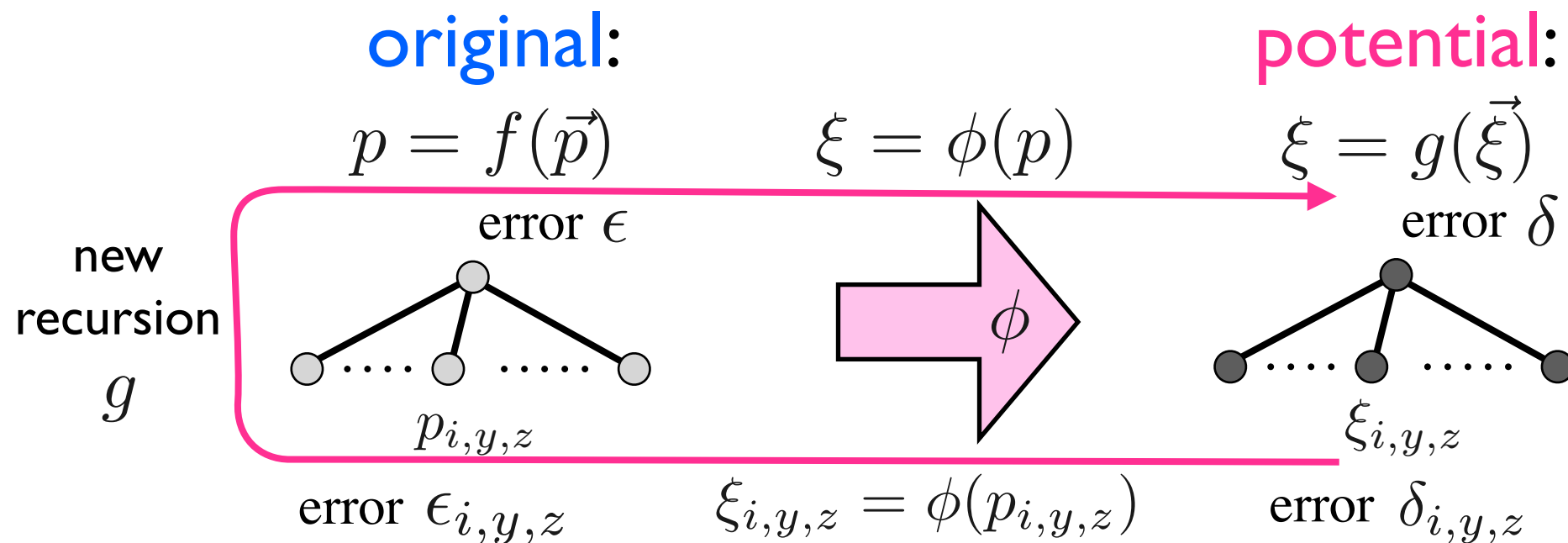
The Potential Method



$$\xi = g(\vec{\xi}) = \phi(f(\phi^{-1}(\xi_{i,y,z}), \forall i, y, z)))$$

let $\Phi(x) = \frac{d\phi(x)}{dx}$ by Mean Value Thm: $\delta \leq \sum_{i,y,z} \left| \frac{\partial f(p)}{\partial p_{i,y,z}} \right| \frac{\Phi(f(p))}{\Phi(p_{i,y,z})} \delta_{i,y,z}$

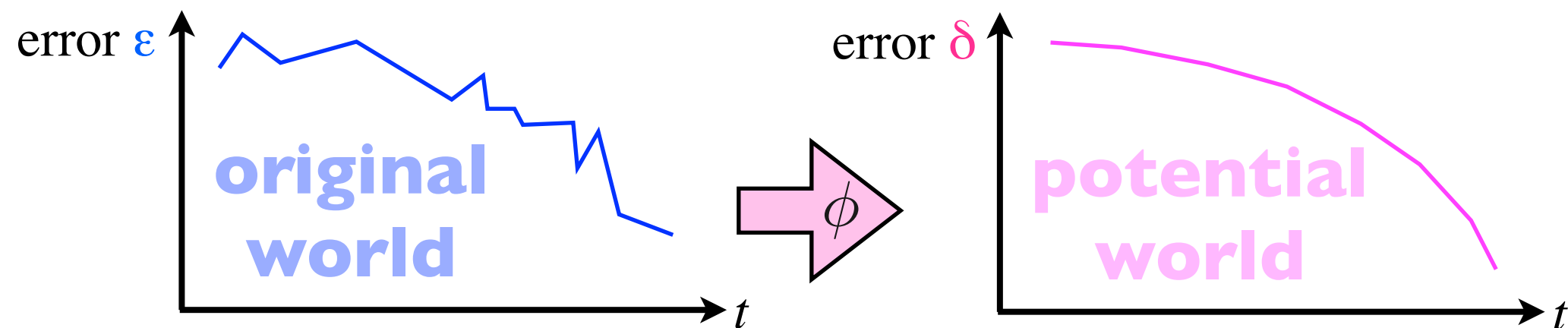
The Potential Method



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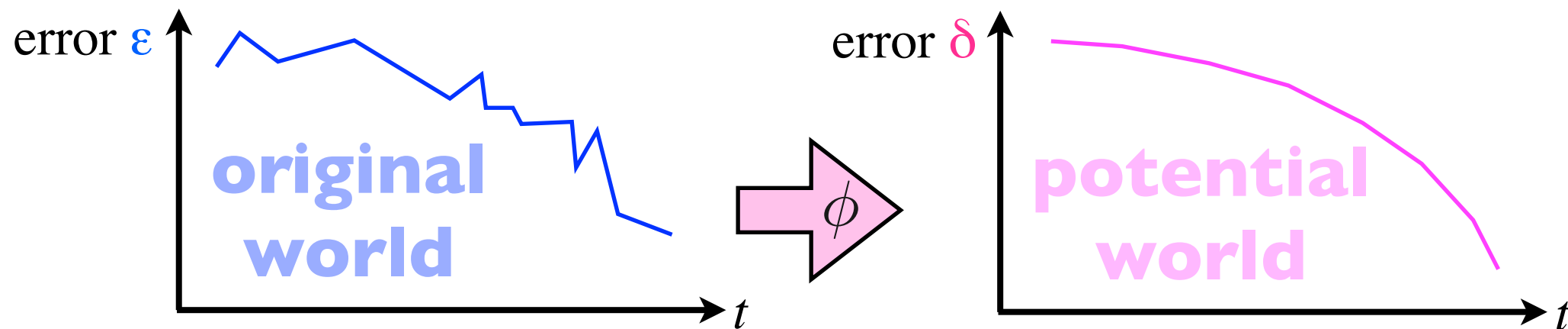
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with good choice of potential function ϕ :



The Potential Method

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- The potential method has been used for analyzing the correlation decay in 2-spin systems (Restrepo-Shin-Tetali-Vigoda-Yang'11, Sinclair-Srivastava-Thurley'12, Li-Lu-Yin'12, Li-Lu-Yin'13, Sinclair-Srivastava-Yin'13).
- This is the first time it is used for multi-spin systems.

Amortized Correlation Decay

amortized decay condition:

\exists a positive-valued function $\Phi(p)$, s.t.

- at any step, we have $\sum_{i,y,z} \left| \frac{\partial f(p)}{\partial p_{i,y,z}} \right| \frac{\Phi(f(p))}{\Phi(p_{i,y,z})} < 1$
- the values of $\Phi(p)$ and $\frac{1}{\Phi(p)}$ are bounded over domain

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control the costs of translating initially from
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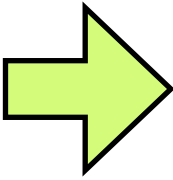
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by induction: for the considered classes of spin systems

amortized
decay condition  exponential
correlation decay

Amortized Correlation Decay

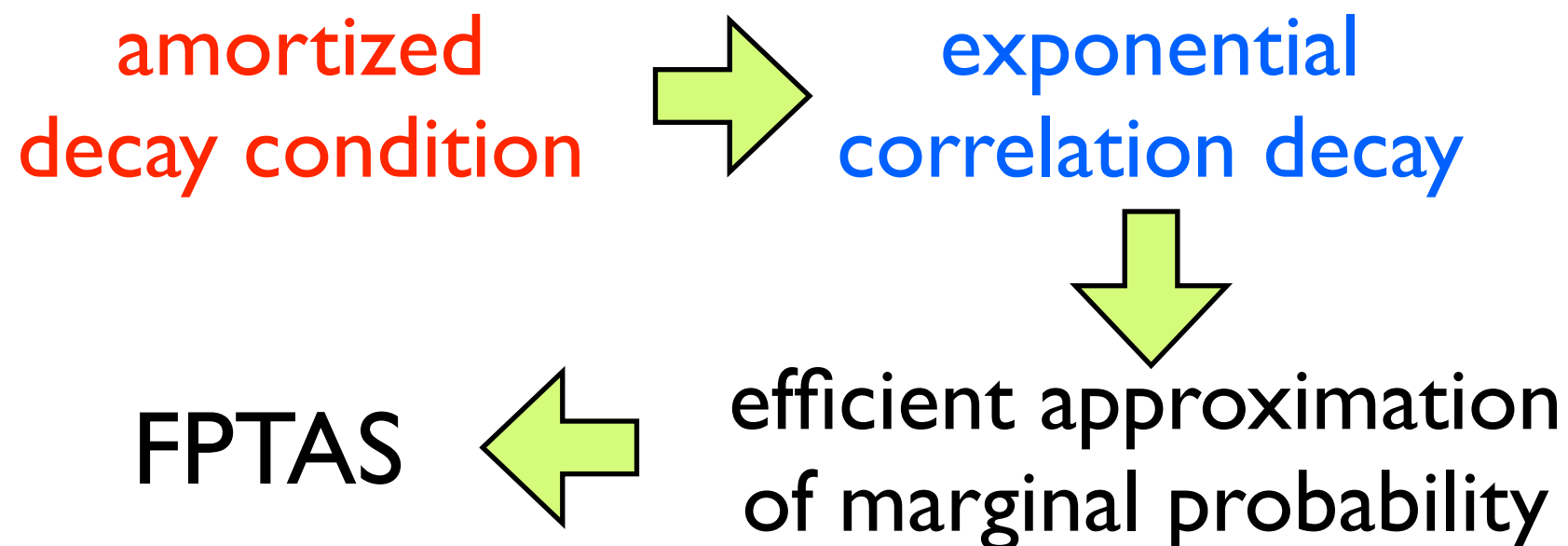
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Establishing the Decay

for **general multi-spin systems**: with max-degree Δ

choose $\Phi(p) = \frac{1}{p + \eta}$ with small enough $\eta > 0$

denoted $c = \max_{\substack{e \in E \\ w, x, y, z \in [q]}} \frac{A_e(x, y)}{A_e(w, z)}$

$$3\Delta(c^\Delta - 1) \leq 1 \quad \Rightarrow \quad \text{amortized decay condition}$$

(by easy calculation)

for **Potts model** (with inverse temperature β):

directly translated to $3\Delta(e^{|\beta|} - 1) \leq 1$

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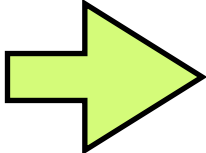
* Other potential functions may further improve the constant factor (but may be harder to analyze).

Establishing the Decay

for **list-coloring**: with max-degree Δ ,
each vertex v with color list L_v

choose $\Phi(p) = \frac{1}{(1-p)\sqrt{p}}$

observing that for list-coloring satisfying the condition,
marginals are always bounded away from both 0 and 1

$\forall v, \quad |L_v| \geq \alpha\Delta + 1$
 $\alpha \approx 2.58071$  **amortized
decay condition**
(by more involved calculation)

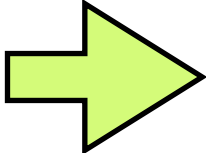
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* The potential functions are chosen in an ad hoc way.

Open Problem

- Find a more systematic way for designing good potential functions.
- Further improve the bounds for correlation decay and FPTAS for multi-spin systems.
- For coloring: $\alpha=2$ is a barrier for the approach due to the overheads caused by total differentiation. Overcome this barrier.