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Bounding Performance of Random Walks in the Positive Orthant



- ▶ Phd from TU Delft, electrical engineering, 2010
 - information theory for wireless communication networks
 - ► in part at Ecole Polytechnique Fédérale de Lausanne, Switzerland
- ► Assistent professor at University of Twente since October 2014
 - stochastic operations research

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This talk is based on joint work with Xinwei Bai, Richard Boucherie, Yanting Chen, Tom Coenen and Jan-Kees van Ommeren.







Model: random walk in the positive orthant

- Discrete-time Markov chain on state space: $S = \{0, 1, ...\}^D$
- ▶ Partition of *S* into components $C_1, C_2, ..., C_K$ of the form

$$C_k = \prod_{d=1}^D \{b_\ell(k,d),\ldots,b_u(k,d)\}.$$

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- ▶ Discrete-time Markov chain on state space: $S = \{0, 1, ... \}^D$
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$$C_k = \prod_{d=1}^D \{b_\ell(k,d),\ldots,b_u(k,d)\}.$$

- $p_u(n)$: Probability to jump from n to n + u
- ► Translation invariant transition probabilities in each component, i.e.,

if
$$n, m \in C_k$$
, then $p_u(n) = p_u(m)$.

Transitions to neighbours only

$$p_u(n) > 0$$
, only if $u \in N = \{-1, 0, 1\}^D$.

- ► Irreducible, aperiodic and positive recurrent Markov chain
- Stationary probability distribution, π : S → [0,∞), is unique solution to balance equations π(n) = ∑_mπ(m)p_{n-m}(m).
- Performance measure
 - $\mathcal{F} = \mathbb{E}_{\pi}[F]$, where $F : S \to [0, \infty)$.
 - Component-wise linear.
 - ► Examples: First moments, tail probabilities, blocking probabilities, ...

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Goal

Bound \mathcal{F} .

Stationary Distribution of Random Walks

- ► Boundary value problem for quarter plane / two dimensions:
 - Find generating function of invariant measure
 - ► Cohen & Boxma, 1983
 - ► Fayolle, lasnogorodski & Malyshev, 1999
 - Depends on solution of conformal mapping
 - No generalization to more than two dimensions
- Matrix geometric approach:
 - ▶ Neuts, 1981
 - ► Latouche & Ramaswami, 1999
 - Depends on solution of non-linear matrix equation
 - ► No generalizations to more than two dimensions that preserve structure
- More than four dimensions:
 - ► Gamarnik, 2002: Positive recurrence is algorithmically undecidable

Geometric Product-form Random Walks

In special cases we have

$$\pi(n)=\pi(n_1,\ldots,n_D)=\prod_{d=1}^D(1-\rho_d)\rho^{n_d},$$

 $\rho_d \in (0, 1).$

- ► Examples:
 - Jackson queueing networks (routing between independent servers)
 - Networks with negative customers
 - ▶ ...
- Characterization of product-form queueing networks
 - van Dijk "Queueing Networks and Product Forms: a system's approach", 1993
 - ► Latouche & Miyazawa, QUESTA, 2014

- Construct a second/perturbed random walk with:
 - transition probabilities $\bar{p}_u(n)$.
 - known stationary distribution $\bar{\pi}$.
- Question: Can we approximate \mathcal{F} in terms of $\bar{\pi}$?

$$\bar{p}_u(n) \approx p_u(n) \stackrel{??}{\iff} \mathcal{F} \approx \sum_n \bar{\pi}(n) F(n)$$





A Bilinear Programming Approach to Error Bounds

- ▶ Interpret *F* as one-step reward function.
- Expected cumulative reward starting from state *n*:

$$F^{t}(n)=F(n)+\sum_{u}p_{u}(n)F^{t-1}(n+u),$$

for
$$t > 0$$
 and $F^0(n) = 0$.

► Bias term:

$$D_u^t(n) = F^t(n+u) - F^t(n).$$

$$F^{t}(n) = F(n) + \sum_{u} p_{u}(n)F^{t-1}(n+u),$$

 $D^{t}_{u}(n) = F^{t}(n+u) - F^{t}(n).$

Theorem (van Dijk)
Let
$$\overline{F} : S \to [0, \infty)$$
 and $G : S \to [0, \infty)$ satisfy
 $\left| \overline{F}(n) - F(n) + \sum_{u} [\overline{p}_{u}(n) - p_{u}(n)] D_{u}^{t}(n) \right| \leq G(n)$

for all $n \in S$ and $t \ge 0$. Then

$$\left|\sum_{n} \bar{\pi}(n) \bar{F}(n) - \mathcal{F}\right| \leq \sum_{n} \bar{\pi}(n) G(n).$$

A Recurrence Relation for the Bias Terms

• We will find variables c(n, u, v, w) for which

$$D_{u}^{t+1}(n) = F(n+u) - F(n) + \sum_{v,w} c(n,u,v,w) D_{v}^{t}(n+w).$$
(*)

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▶ Purpose of c(n, u, v, w):

Lemma

If
$$A_u: S \to [0,\infty)$$
 and $B_u: S \to [0,\infty)$, $u \in N$ satisfy

$$F(n+u) - F(n) + \sum_{v,w} \max\{-c(\cdot)A_v(n+w), c(\cdot)B_v(n+w)\} \le B_u(n),$$

$$F(n) - F(n+u) + \sum_{v,w} \max\{-c(\cdot)B_v(n+w), c(\cdot)A_v(n+w)\} \le A_u(n),$$

for all $n \in S$ and (*) is satisfied, then

$$-A_u(n) \leq D_u^t(n) \leq B_u(n),$$

for all $u \in N$, $n \in S$ and $t \ge 0$.

Lemma

We can formulate a set of linear constraints on c(n, u, v, w) that ensure

$$D_u^{t+1}(n) = F(n+u) - F(n) + \sum_{v,w} c(n, u, v, w) D_v^t(n+w).$$

Linear constraints in form of a flow problem

$$\blacktriangleright D^t_w(n) = F^t(n+w) - F^t(n)$$

►
$$D_u^{t+1}(n) = F(n+u) - F(n) + \sum_{v,w} c(n,u,v,w) \left[F^t(n+w+v) - F^t(n+w)\right]$$

•
$$F^{t+1}(n) = F(n) + \sum_{u} p_u(n)F^t(n+u)$$

►
$$D_u^{t+1}(n) = F(n+u) - F(n) + \sum_v p_u(n)F^t(n+u+v) - \sum_w p_w(n)F^t(n+w)$$

$$F^{t}(n) = F(n) + \sum_{u} p_{u}(n)F^{t-1}(n+u),$$

$$D^{t}_{u}(n) = F^{t}(n+u) - F^{t}(n).$$

Theorem (van Dijk) Let $\overline{F} : S \to [0, \infty)$ and $G : S \to [0, \infty)$ satisfy $\left| \overline{F}(n) - F(n) + \sum_{u} [\overline{p}_{u}(n) - p_{u}(n)] D_{u}^{t}(n) \right| \leq G(n)$

for all $n \in S$ and $t \ge 0$. Then

$$\left|\sum_{n} \bar{\pi}(n) \bar{F}(n) - \mathcal{F}\right| \leq \sum_{n} \bar{\pi}(n) G(n).$$

Let
$$q_u(n) = \bar{p}_u(n) - p_u(n)$$
.

 \blacktriangleright Upper bound ${\mathcal F}$ by

$$\begin{split} \min \ \sum_{n \in S} \left[\bar{F}(n) + G(n) \right] \bar{\pi}(n), \\ \text{subject to } \bar{F}(n) - F(n) + \sum_{u} \max \left\{ q_{u}(n) B_{u}(n), -q_{u}(n) A_{u}(n) \right\} \leq G(n), \\ F(n) - \bar{F}(n) + \sum_{u} \max \left\{ q_{u}(n) A_{u}(n), -q_{u}(n) B_{u}(n) \right\} \leq G(n), \\ F(n+u) - F(n) + \sum_{v,w} \max\{ -c(\cdot) A_{v}(n+w), c(\cdot) B_{v}(n+w) \} \leq B_{u}(n), \\ F(n) - F(n+u) + \sum_{v,w} \max\{ -c(\cdot) B_{v}(n+w), c(\cdot) A_{v}(n+w) \} \leq A_{u}(n), \\ \text{Constraints (*), linear in } c(\cdot), \\ A_{u}(n) \geq 0, B_{u}(n) \geq 0, \bar{F}(n) \geq 0, G(n) \geq 0, \quad \text{for } n \in S, u \in N. \end{split}$$

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Reduces to a finite number of constraints and variables if F, Au, Bu, G and c are constrained to be component-wise linear.

Aim

General purpose software implementation in which transition probabilities and reward function are only input parameters.

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General purpose software implementation in which transition probabilities and reward function are only input parameters.

- ► Pyomo: a Python-based optimization modelling framework.
- Current implementation:
 - Additional input: perturbed random walk and its stationary distribution $\bar{\pi}$.
 - ► Find a feasible (not necessarily optimal) solution by first solving for (any) c(n, u, v, w) and then the remaining linear program.

Example 1: Coupled Queue with a Finite Buffer



Example 1: Perturbed Random Walk



 $\pi(n) = \alpha \left(\frac{\lambda_1}{\mu_1}\right)^{n_1} \left(\frac{\lambda_2}{\mu_2}\right)^{n_2}$, where α is the normalization constant.

 $F_1 = n_1$: average number of customers in the first queue







Perturbations for Two-dimensional Random Walks

• Geometric product form measure: $m(i,j) = \rho^i \sigma^j$, $(\rho, \sigma) \in (0,1)^2$.

- Geometric product form measure: m(i,j) = ρⁱσ^j, (ρ, σ) ∈ (0,1)².
 Balance equation in interior: m(i,j) = Σ¹_{s=-1}Σ¹_{t=-1}m(i-s,j-t)p_{s,t}.
- **Balance** equations

$$\rho\sigma\left(1-\sum_{s=-1}^{1}\sum_{t=-1}^{1}\rho^{-s}\sigma^{-t}p_{s,t}\right)=0\quad,$$

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- ► Balance equations induce algebraic curves Q

$$Q = \left\{ (\rho, \sigma) \mid \rho\sigma \left(1 - \sum_{s=-1}^{1} \sum_{t=-1}^{1} \rho^{-s} \sigma^{-t} p_{s,t} \right) = 0 \right\},$$

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- Balance equations induce algebraic curves Q, H and V,

$$Q = \left\{ (\rho, \sigma) \middle| \rho\sigma \left(1 - \sum_{s=-1}^{1} \sum_{t=-1}^{1} \rho^{-s} \sigma^{-t} \rho_{s,t} \right) = 0 \right\},$$

$$H = \left\{ (\rho, \sigma) \middle| \rho \left(1 - \sum_{s=-1}^{1} \rho^{-s} \sigma \rho_{s,-1} - \sum_{s=-1}^{1} \rho^{-s} h_s \right) = 0 \right\},$$

$$V = \left\{ (\rho, \sigma) \middle| \sigma \left(1 - \sum_{t=-1}^{1} \rho\sigma^{-t} \rho_{-1,t} - \sum_{t=-1}^{1} \sigma^{-t} v_t \right) = 0 \right\}.$$

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• $m(i, j) = \rho^i \sigma^j$ is invariant measure iff $(\rho, \sigma) \in Q \cap H \cap V$.

Product-form random walks (cont'd)

- $m(i,j) = \rho^i \sigma^j$ is invariant measure iff $(\rho, \sigma) \in Q \cap H \cap V$.
- Example of a product-form random walk:



 Example continued: Not a product form for other boundary transition rates



Characterization of product-form queueing networks

- van Dijk "Queueing Networks and Product Forms: a system's approach", 1993
- ► Latouche & Miyazawa, QUESTA, 2014

Extending the class of 'tractable' random walks

- Very limited number of random walks have a product-form invariant measure
- ► Goal: extend class of 'tractable' performance measures

Consider $m(i,j) = \sum_{(\rho,\sigma)\in\Gamma} \alpha(\rho,\sigma) \rho^i \sigma^j$

- \blacktriangleright Sum of geometric terms induced by Γ
- Assumptions:
 - $\blacktriangleright \ |\Gamma| < \infty,$
 - Γ ⊂ (0, 1)²,

Compensation approach

► Countably many terms with pairwise-coupled structure



- ► Sum of geometric terms: $m(i,j) = \sum_{(\rho,\sigma) \in \Gamma} \alpha(\rho,\sigma) \rho^i \sigma^j$
- ► When does a random walk have an invariant measure that can be expressed as a sum of geometric terms?
 - Which elements (ρ, σ) can occur in Γ ?
 - Does Γ need to have a specific structure?
 - What are the values of $\alpha(\rho, \sigma)$?
- Can we construct a random walk that has a prespecified invariant measure that is a sum of geometric terms?

Example 2







Future Work

- Improve solution method for bilinear program
- Extend class of perturbed random walks
- Make software available

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Thank you for your attention.