

**Un modèle PDMP multi-classes pour  
le contrôle de congestion implémenté par  
les connexions dans un grand réseau**

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**Travail effectué avec**

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## Travail effectué avec

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1. Interacting multi-class transmissions in large stochastic networks (AAP 2009)
2. Self-adaptive congestion control for multi-class intermittent connections in a communication network (QUES 2011)

et pour une étude poussée sur les **lois invariantes** pour 1.,

**Maaïke Verloop**, **CWI Amsterdam**

3. Stability properties of networks with interacting TCP flows (Proceedings of NET-COOP 2009)

# 1 Modeling the Internet

The **Internet** is a **vast, complex**, and **ill-known** structure  
in **perpetual evolution**.

**Undisciplined users**

access it for **varied** purposes  
and have **vastly different characteristics**  
and **QoS (quality of service)** requirements.

**Connections** (initiated by **users**) have to  
**self-adapt**  
to the **packet losses** due to the **congestion** they **create**  
by **regulating** their **output**  
using **algorithms** such as the  
**congestion control** part of **TCP**  
(**T**ransmission **C**ontrol **P**rotocol).

It is important to understand better this  
**feedback loop.**

Highly **nonlinear** behavior is expected.

The situation is a **fairly well** understood for a

**single node**

with **asymptotic formulæ** for **throughput**, etc., even if

**many problems** remain **open**

depending on the level of precision and detail sought

(**delayed information ...**)

**Much less**

is understood about the **true** issue,

**Stochastic Networks,**

*i.e.*, the **coexistence**, and hence the **interaction**,



- of **varied data flows**  
(file transfers, telephony, video streaming, web browsing, MMORPG, P2P, ...)
- with **varied characteristics**  
(routes, QoS requirements, rates, ...)

each **using** in its **own fashion**

**limited** common resources

- constituted of **varied nodes**  
(links, routers, processors, buffers, ...)
- with **varied purposes.**

It is natural to regroup **data flows** into a  
**reasonable number of classes**,  
according to their **characteristics**, and we **focus** on such  
**interacting multi-class network models.**

## **One approach:**

**Introduced and studied by several distinguished authors,  
*e.g.*,**

**Kelly, Maulloo, and Tan (JORS 1998)**

**Massoulié and Roberts (INFOCOM 1999)**

**Kelly and Williams (AAP 2004)**

**Massoulié (AAP 2007)**

**Kang, Kelly, Lee, and Williams (AAP 2009)**

**Kelly, Massoulié, and Walton (QS 2009).**

**Optimisation problem:** if there are  $x_k$  **connections** of class  $k \in \{1, \dots, K\}$ , then class  $k$  obtains **throughput**  $\lambda_k^*$  such that  $(\lambda_k^*)$  achieves

$$\max_{\lambda \in \mathcal{C}} \sum_{k=1}^K x_k U_k(\lambda_k / x_k)$$

where  $U_k$  is some **utility function**, and  $\mathcal{C}$  depends on the capacity of the resources.

The **utility function** is used to **summarize** in an **idealized** fashion the **effect** of **TCP** on the **network**.

This is a very **high level macroscopic** approach,  
perhaps akin to **Thermodynamics**.

**Our approach:**

**to devise a microscopic model  
and then derive its macroscopic limit.**

The **idealization** of **TCP** takes place at the **micro** level.

**Akin to kinetic models in Statistical Physics.**

## **2 Markovian modeling of user interaction**

**We propose**

a **Markovian model** for a **stochastic network**

- constituted of  $J \geq 1$  **nodes**,
- hosting  $K \geq 1$  **classes** of **users** (or **connections**),
- with  $N_k \geq 1$  **users** in **class**  $k$  for  $1 \leq k \leq K$ .

$$N = (N_1, \dots, N_K)$$

**class size vector**,

$$|N| = N_1 + \dots + N_K$$

**total number of users.**



Each **user** (or **connection**) alternates between being

- **active**, or **on**, when it has **data** to **transmit**
- and **inactive**, or **off**, when it has **none**

and thus generates

**intermittent** (**intermittent**) **transmissions**.

# Markov model and Itô-Skorohod equations

This model corresponds to a **Markov process**

$$W^N(t) = \left( W_{n,k}^N(t), 1 \leq n \leq N_k, 1 \leq k \leq K \right), \quad t \geq 0,$$

where

$$W_{n,k}^N(t) \in \mathbb{R}_+ \cup \{-1\}$$

is the **state (output, window)** of the  $n$ -th **transmission** of **class  $k$**  at time  $t$ , with  $-1$  corresponding to an **inactive user**.

We choose to represent it (in law) by the following

**Itô-Skorohod SDE:**

$$\begin{aligned}
dW_{n,k}^N(t) = & \mathbb{1}_{\{W_{n,k}^N(t-) = -1\}} \int (1+w) \mathbb{1}_{\{0 < z < \lambda_k(\mathbf{U}^N(t-))\}} \mathcal{A}_{n,k}(dw, dz, dt) \\
& + \mathbb{1}_{\{W_{n,k}^N(t-) \geq 0\}} \left[ a_k(W_{n,k}^N(t-), \mathbf{U}^N(t-)) dt \right. \\
& - (1-r_k) W_{n,k}^N(t-) \int \mathbb{1}_{\{0 < z < b_k(W_{n,k}^N(t-), \mathbf{U}^N(t-))\}} \mathcal{N}_{n,k}(dz, dt) \\
& \left. - (1+W_{n,k}^N(t-)) \int \mathbb{1}_{\{0 < z < \mu_k(W_{n,k}^N(t-), \mathbf{U}^N(t-))\}} \mathcal{D}_{n,k}(dz, dt) \right]
\end{aligned}$$

with  $\mathbf{U}^N(t) = (U_j^N(t), 1 \leq j \leq J)$ ,  $U_j^N(t) = \sum_{k=1}^K A_{jk} \sum_{n=1}^{N_k} W_{n,k}^N(t)^+$ ,

where the  $\mathcal{A}_{n,k}$  are **Poisson point processes** with intensity  $\alpha_k(dw)dzdt$ , and the  $\mathcal{N}_{n,k}$  and  $\mathcal{D}_{n,k}$  are **Poisson point processes** with intensity  $dzdt$ .

The following can be proved using **standard arguments**.

If the functions  $a_k$  are **Lipschitz**

and  $b_k$ ,  $\lambda_k$ , and  $\mu_k$  are **locally bounded**,

then there is **pathwise existence and uniqueness**

of solution for the **SDE**,

and the corresponding **Markov process** is **well defined**.

# The mean-field asymptotic regime

This **SDE** constitutes

a **coupled** system in **very high dimension**

and an **asymptotic study** is performed to render it

**tractable**, which **reduces**

the **dimension** to the **number of classes**  $K$ .

For  $1 \leq k \leq K$ , we **assume** that

$$N_k \rightarrow \infty, \quad \frac{N_k}{|N|} := \frac{N_k}{N_1 + \dots + N_K} \rightarrow p_k,$$

and, so as to **scale** the resource **capacities** adequately, that a factor  $\frac{1}{|N|}$  is **introduced** inside the **coefficients** by

replacing  $U^N$  by  $\overline{U}^N = \frac{1}{|N|} U^N$ .

This yields the following **mean-field rescaled SDE**:

$$\begin{aligned}
dW_{n,k}^N(t) = & \mathbb{1}_{\{W_{n,k}^N(t-) = -1\}} \int (1+w) \mathbb{1}_{\{0 < z < \lambda_k(\bar{U}^N(t-))\}} \mathcal{A}_{n,k}(dw, dz, dt) \\
& + \mathbb{1}_{\{W_{n,k}^N(t-) \geq 0\}} \left[ a_k(W_{n,k}^N(t-), \bar{U}^N(t-)) dt \right. \\
& - (1-r_k) W_{n,k}^N(t-) \int \mathbb{1}_{\{0 < z < b_k(W_{n,k}^N(t-), \bar{U}^N(t-))\}} \mathcal{N}_{n,k}(dz, dt) \\
& \left. - (1+W_{n,k}^N(t-)) \int \mathbb{1}_{\{0 < z < \mu_k(W_{n,k}^N(t-), \bar{U}^N(t-))\}} \mathcal{D}_{n,k}(dz, dt) \right]
\end{aligned}$$

with  $\bar{U}^N(t) = \frac{1}{|N|} U^N(t) = \left( \bar{U}_j^N(t), 1 \leq j \leq J \right),$

$$\bar{U}_j^N(t) = \frac{1}{|N|} U_j^N(t) = \sum_{k=1}^K A_{jk} \frac{N_k}{|N|} \bar{W}_k^N(t), \quad \bar{W}_k^N(t) = \frac{1}{N_k} \sum_{n=1}^{N_k} W_{n,k}^N(t)^+.$$

This system is in **multi-class mean-field interaction** through

$$\left( \overline{W}_k^N(t), 1 \leq k \leq K \right) \text{ via } \overline{U}^N(t) = \frac{\mathbf{1}}{|N|} U^N(t),$$

the **vector of the empirical means of the class outputs**

via

the **rescaled throughput vector**.



For  $1 \leq k \leq K$ , a natural quantity is the **class  $k$  empirical measure**

$$\Lambda_k^N = \frac{1}{N_k} \sum_{n=1}^{N_k} \delta_{(W_{n,k}^N(t), t \geq 0)}$$

where  $\delta_{(x(t), t \geq 0)}$  denotes the Dirac mass at the **sample path**  $(x(t), t \geq 0)$ .

Notably

$$\overline{W}_k^N(t) = \langle w^+, \Lambda_k^N(t)(dw) \rangle.$$

# **3 Limit nonlinear Markov process and class interaction**

**In the mean-field asymptotic regime,**  
**under adequate assumptions for the initial conditions,**  
**a propagation of chaos phenomenon is expected:**

- the **processes**  $W_{n,k}^N$  should become **independent**, and each **converge in law** to a **process**  $W_k$ ,
- the **empirical measures**  $\Lambda_k^N$  (**random laws on path space**) should **converge in law** (and in probability) to the **law** of the **same process**  $W_k$ ,

where the **limit** process

$$W(t) = (W_k(t), 1 \leq k \leq K), \quad t \geq 0,$$

solves the (**adequately started**) following equation:

## McKean-Vlasov Itô-Skorohod SDE (Nonlinear Markov process)

$$\begin{aligned}
 dW_k(t) = & \mathbb{1}_{\{W_k(t-) = -1\}} \int (1+w) \mathbb{1}_{\{0 < z < \lambda_k(\mathbf{u}_W(t))\}} \mathcal{A}_k(dw, dz, dt) \\
 & + \mathbb{1}_{\{W_k(t-) \geq 0\}} \left[ a_k(W_k(t-), \mathbf{u}_W(t)) dt \right. \\
 & - (1-r_k)W_k(t-) \int \mathbb{1}_{\{0 < z < b_k(W_k(t-), \mathbf{u}_W(t))\}} \mathcal{N}_k(dz, dt) \\
 & \left. - (1+W_k(t-)) \int \mathbb{1}_{\{0 < z < \mu_k(W_k(t-), \mathbf{u}_W(t))\}} \mathcal{D}_k(dz, dt) \right]
 \end{aligned}$$

with

$$\mathbf{u}_W(t) = (u_{W,j}(t), 1 \leq j \leq J), \quad u_{W,j}(t) = \sum_{k=1}^K A_{jk} p_k \mathbb{E}(W_k(t)^+).$$

The **interaction** between coordinates depends on  
the **mean throughput** vector

$$(u_W(t), t \geq 0)$$

which is a **linear** functional of  
the **mean class output** vector

$$\mathbb{E}(W(t)^+) = \mathbb{E}(W_k(t)^+, 1 \leq k \leq K) = \langle w^+, \mathcal{L}(W(t)) \rangle.$$

Notably, the **infinitesimal generator** of the Markov process  $(W(t), t \geq 0)$  depends, at time  $t$ , on the

**law**  $\mathcal{L}(W(t))$  of  $W(t)$  itself

and not only on the **value** of the sample path.

**Actually**, only on the **class marginals** of  $\mathcal{L}(W(t))$ .



The **Kolmogorov** equations are  
**nonlinear** integro-differential equations  
and a  
**nonlinear martingale problem**

can be associated to this process.

This is why it is called a

**Nonlinear Markov process.**

Using this **SDE representation**, we have adapted  
**contraction** and **coupling** techniques developed  
for **exchangeable** systems by **Sznitman (1980's, 1989)**  
to this **multi-class** setting (**technical**), and  
**obtained existence and uniqueness**  
for this **fixed-point** problem,  
as well as **propagation of chaos**.

This was done in **G. and Robert (AAP 2009)** for

**persistent transmissions**

and has been **extended** to the **general case** of

**on-off users with intermittent transmissions**

in **G. and Robert (QUES 2011)**

There were many **difficulties**, e.g.,

- **lack** of **symmetry** of **multi-class** systems w.r.t. **exchangeable** ones:

$$N_1! \cdots N_K! \ll |N|! = (N_1 + \cdots + N_K)!$$

and **rigorous** results are **rare** for **multi-class** systems,

- **quadratic behavior** of  $b_k(w, u) = w \beta_k(u)$ ,
- **on-off users** introduce **discontinuities**. Smart choices may simplify computations ( $-1$  as **cemetery state**, ...).

The **assumptions**, notably on **initial conditions**, reflect this.

**These** must not be too **stringent** since

**long-time** and **stationary** behavior  
are **essential**.

**Gaussian moment assumptions** were used.

# **4 Fixed points and invariant laws for the limit**

Recall that **solution**  $(W(t), t \geq 0)$  has in general

an **infinitesimal generator** at time  $t$

**depending** not only on the **state**,

but also, through  $u_W(t)$ , on  $\mathbb{E}(W(t)^+)$

and thus **on the law**  $\mathcal{L}(W(t))$  itself.

This **nonlinearity** is an important **complication** in **studying** the behavior of  $(W(t), t \geq 0)$ , in particular for

**existence and uniqueness of invariant laws.**

Note that, starting from an **invariant law**,

$(\mathbb{E}(W(t)^+), t \geq 0)$  and hence  $(u_W(t), t \geq 0)$  are **constant**,

and  $(W(t), t \geq 0)$  corresponds to a

**homogeneous Markov process**  
**in equilibrium.**



Interesting results on the **invariant laws** were obtained in **G. and Robert (AAP 2009, QUES 2011)** by reducing the corresponding

**infinite**-dimensional **fixed-point** problem  
to a **finite**-dimensional one.

The **results** for **on-off users** being still **preliminary**,  
we concentrate now on

**users or connections**

which emit **persistently**

and thus can be identified with their **transmissions**.

**Assume** that, for  $1 \leq k \leq K$ ,

$$a_k(w, u) = a_k(u), \quad b_k(w, u) = w \beta_k(u), \quad w \in \mathbb{R}_+, \quad u \in \mathbb{R}_+^J,$$

**where**

$a_k : \mathbb{R}_+^J \rightarrow \mathbb{R}_+$  is **Lipschitz bounded**

$\beta_k : \mathbb{R}_+^J \rightarrow \mathbb{R}_+$  is **Lipschitz**.

Then, the **invariant laws** for the **nonlinear SDE** are in

**one-to-one** relation with

the **solutions** of the

**finite-dimensional fixed-point problem**

$$u = (u_j, 1 \leq j \leq J) \in \mathbb{R}_+^J, \quad u_j = \sum_{k=1}^K A_{jk} p_k \psi(r_k) \sqrt{\frac{a_k(u)}{\beta_k(u)}},$$

where

$$\psi(r) = \sqrt{\frac{2}{\pi}} \prod_{n=1}^{\infty} \frac{1 - r^{2n}}{1 - r^{2n-1}}.$$

Such a **solution**  $u^*$  corresponds to a **product-form invariant law** with density

$$\prod_{k=1}^K H_{r_k, \rho_k}(w_k), \quad w = (w_k, 1 \leq k \leq K) \in \mathbb{R}_+^K,$$

where  $\rho_k = \frac{a_k(u^*)}{\beta_k(u^*)}$  and, for  $x \in \mathbb{R}_+$ ,

$$H_{r, \rho}(x) = \frac{\sqrt{2\rho/\pi}}{\prod_{n=0}^{\infty} (1 - r^{2n+1})} \sum_{n=0}^{\infty} \frac{r^{-2n}}{\prod_{k=1}^n (1 - r^{-2k})} e^{-\rho r^{-2n} x^2/2}.$$

Note that the **decay** is appropriate for

**Gaussian moment conditions.**

The **limit equilibrium throughput** for **class  $k$  users** is given by

$$\begin{aligned}\lambda_k &= \mathbb{E}(\overline{W}_k) = \int_{\mathbb{R}_+} x H_{r_k, \rho_k}(x) dx \\ &= \psi(r_k) \sqrt{\rho_k} = \psi(r_k) \sqrt{\frac{a_k(u^*)}{\beta_k(u^*)}}\end{aligned}$$

with  $u^*$  solving a **fixed-point problem**

which can be written as

$$H(u^*) = 0.$$

Back to an **optimisation problem** ?

Class  $k \in \{1, \dots, K\}$  receives **throughput**  $\lambda_k^*$ , such that  $(\lambda_k^*)$  achieves

$$\max_{\lambda \in \mathcal{C}} \sum_{k=1}^K p_k U_k(\lambda_k / p_k)$$

which can be put (under **some conditions**) in the form

$$\nabla G(\lambda^* / p) = 0.$$



**How many solutions for the fixed-point problem ?**

**G., Robert, and Verloop (NETCOOP 2009) prove**

**existence and uniqueness of the invariant law**

**by contraction and monotonicity methods,**

**under some assumptions, for topologies such as**

- **One node, several classes**
- **Linear networks with cross-traffic**
- **Trees**
- **Rings and toruses.**

**Assume** moreover that

$$A_{jk} = \begin{cases} 1 & \text{if node } j \text{ is used by a class } k \text{ user,} \\ 0 & \text{otherwise,} \end{cases}$$

$$\beta_k(\mathbf{u}) = \beta_k \left( \sum_{j=1}^J A_{jk} \mathbf{u}_j \right) \cdot$$

**(With slight abuse of notations.)**

The **fixed-point equation** becomes

$$u_j = \sum_{k=1}^K A_{jk} p_k \psi(r_k) \frac{\sqrt{a_k(u)}}{\sqrt{\beta_k \left( \sum_{j=1}^J A_{jk} u_j \right)}}, \quad 1 \leq j \leq J.$$

We **assume** that  $u \rightarrow \beta_k(u)$  is **strictly increasing** and **Lipschitz** and that  $a_k(u) \equiv a_k$  is **constant**.

## One node, several classes

There is a unique solution  $u = u^*$  for the **fixed-point equation**

$$u = \sum_{k=1}^K \psi(r_k) p_k \sqrt{\frac{a_k}{\beta_k(u)}},$$

and a unique **invariant law** with density  $H_{r,\rho_k}$  and expectation

$$\psi(r_k) \sqrt{\frac{a_k}{\beta_k(u^*)}}$$

yielding the **mean equilibrium output**.

**If** moreover the classes vary only in their RTT's, *i.e.*,

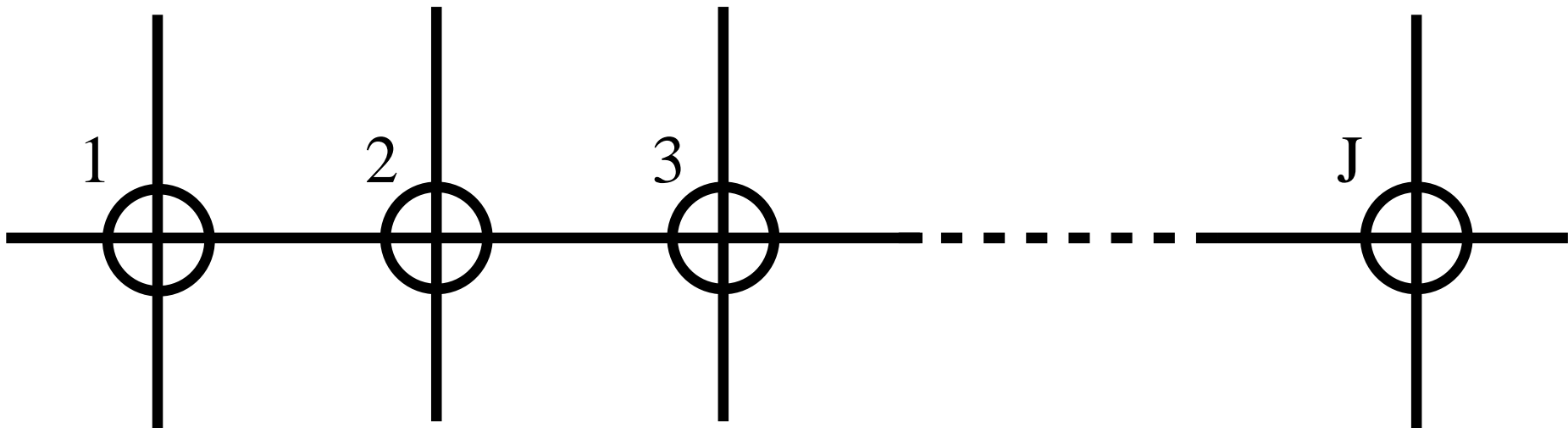
$$\beta_k \equiv \beta, \quad r_k \equiv r, \quad 1 \leq k \leq K,$$

then the class **outputs** differ **only** by the factors

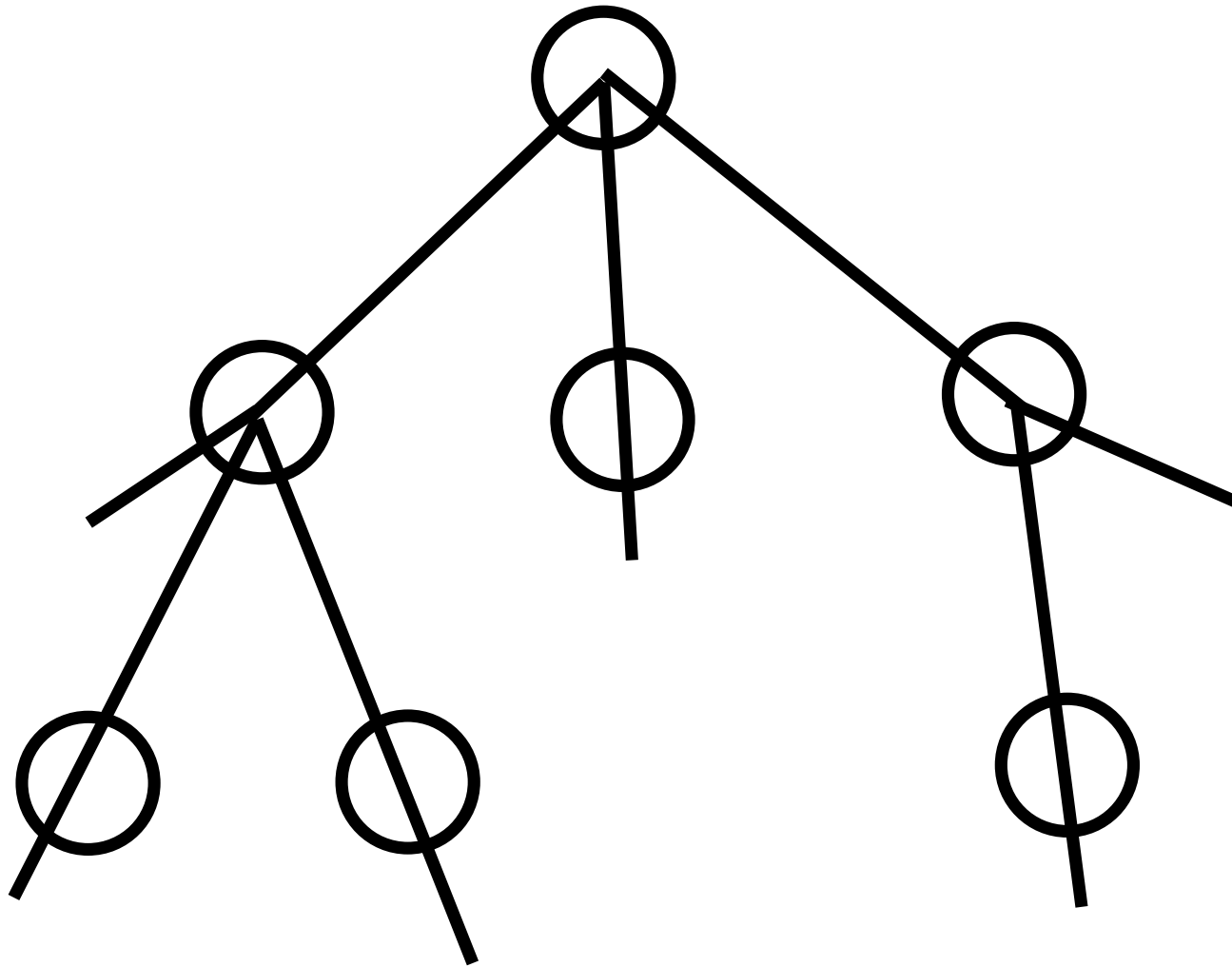
$$\sqrt{a_k} = 1 / \sqrt{\text{RTT}_k}.$$

## Linear network with cross-traffic

There are  $J$  nodes and  $K = J + 1$  classes. For  $1 \leq j \leq J$ , class  $j$  transmissions use only node  $j$ . The transmissions of class  $J + 1$  use all  $J$  nodes.

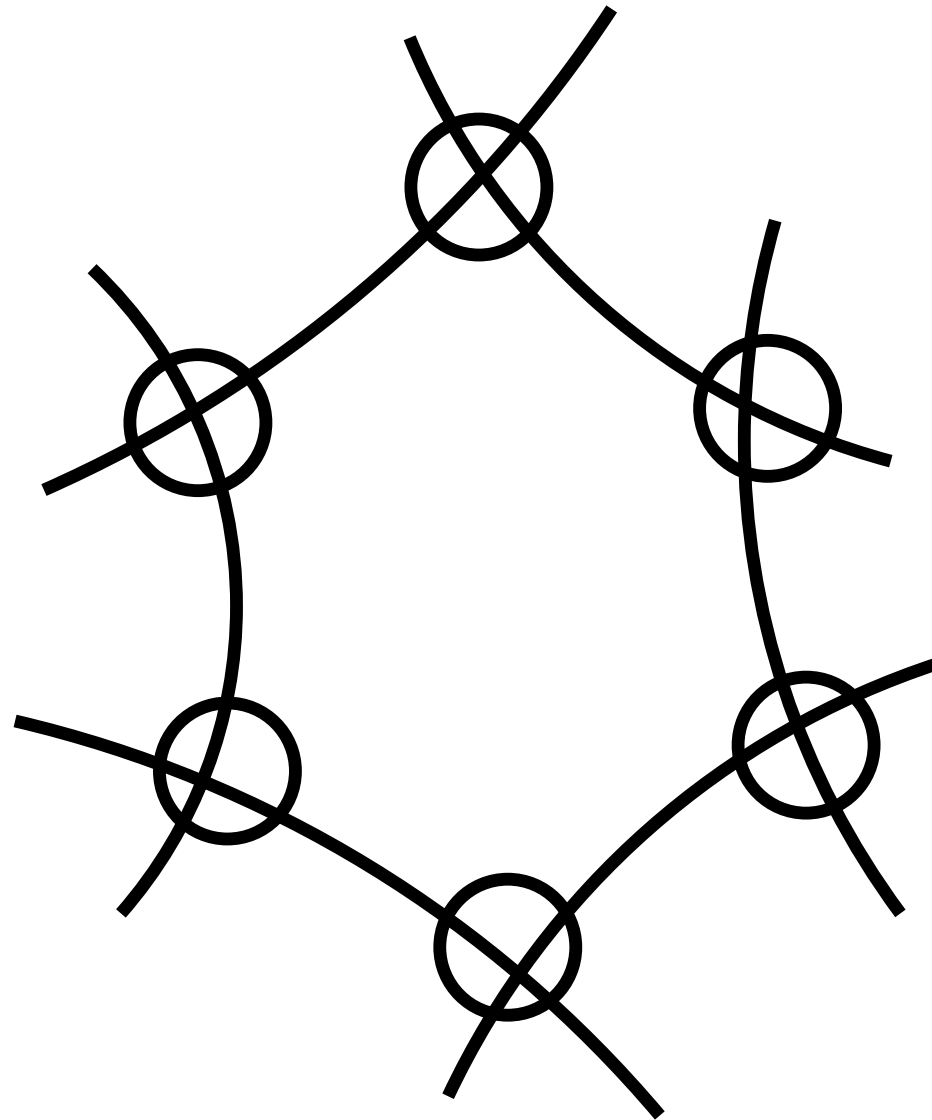


# Trees

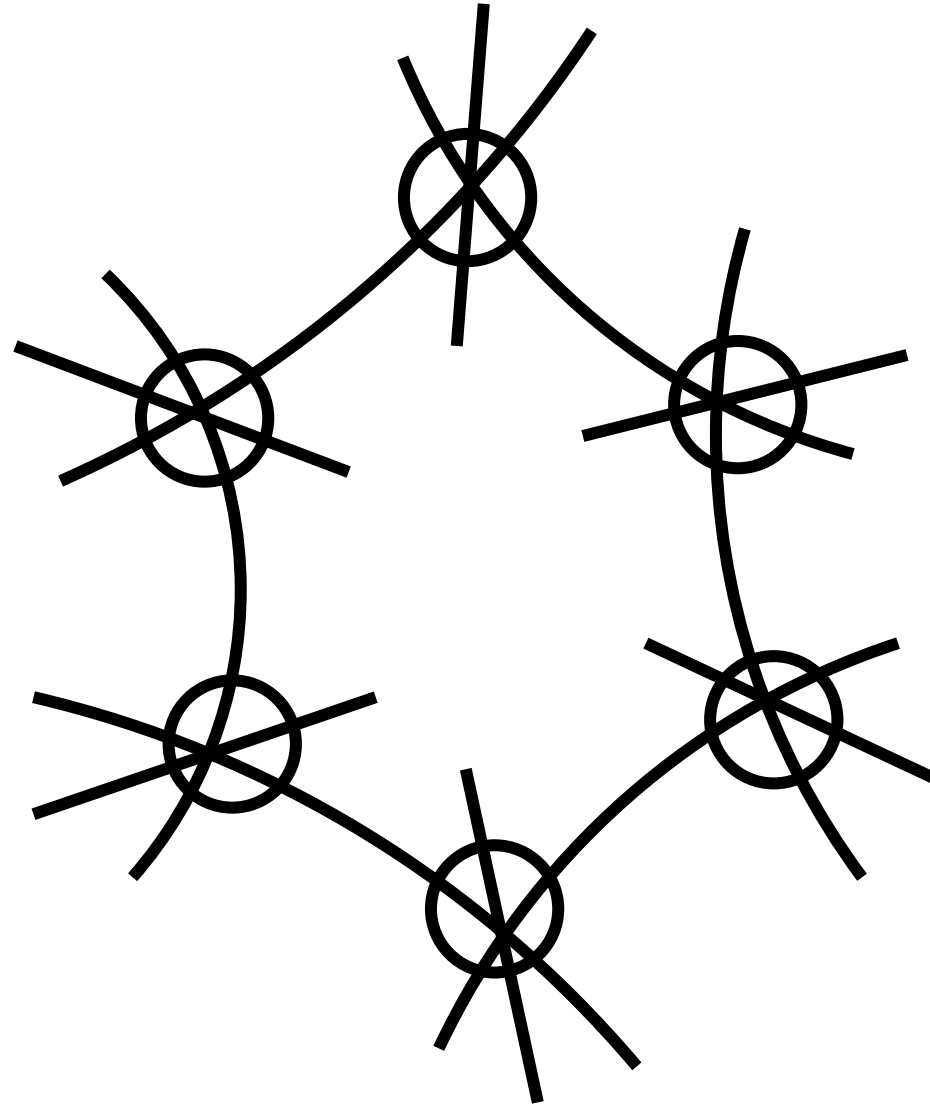




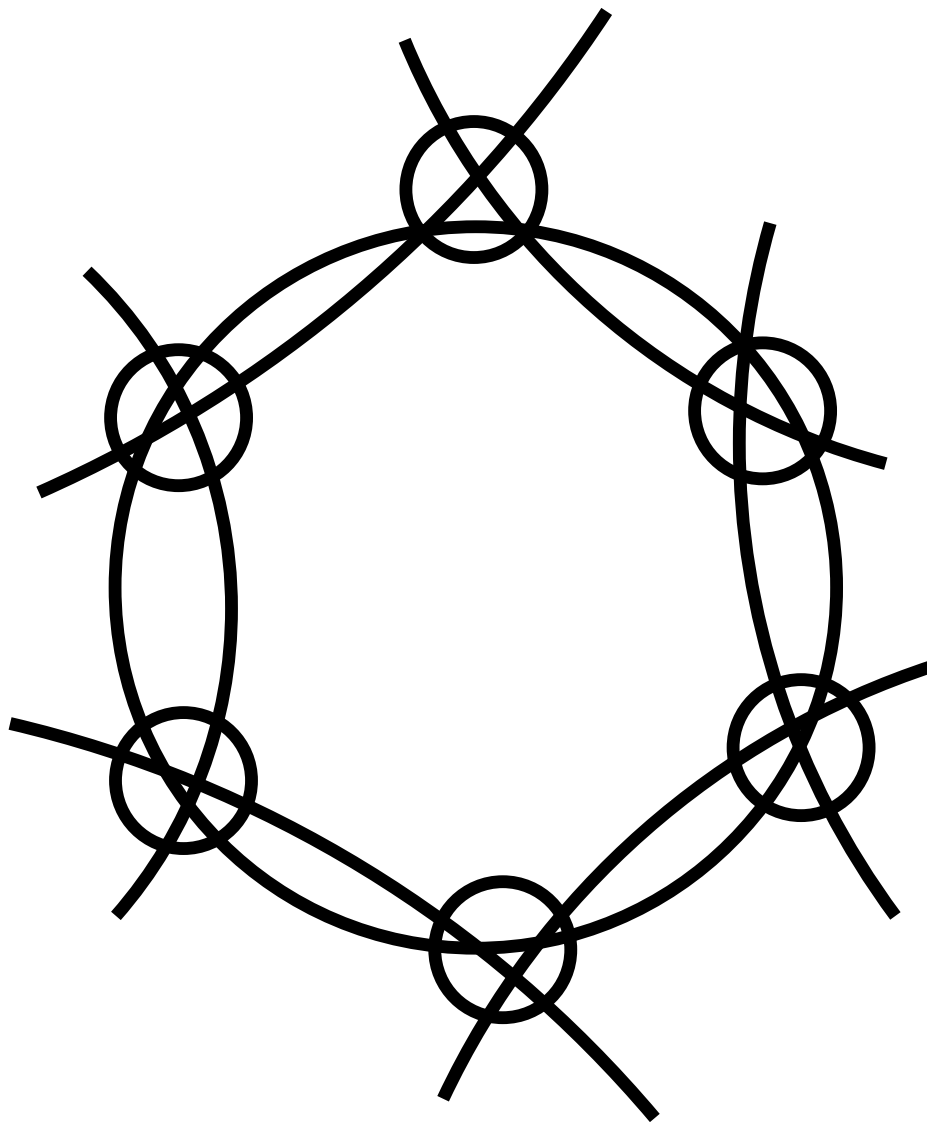
# Ring topology, 1



# Ring topology, 2



## Ring topology, 3



**Is uniqueness** always **true** if

$$u \rightarrow \beta_k(u)$$

is **increasing** ?

There exist **examples** when

**not:**

**Raghunathan and Kumar (2007)** in a wireless context.

**Are there**  
**meta-stability** phenomena  
**in these networks ?**

## Known for

- **Loss Networks: Gibbens *et al.* (1990), Marbukh (1993),**
- **Wireless Networks: Antunes *et al.* (2008).**

Are there **oscillations** ?



Such possibilities have been suggested by the literature on the **Internet** and by **simulations**:

## **cyclic behavior in congested networks**

e.g., **mean-field** study for a **single node** and a **single class** and **delay equations** in

**Bacelli, McDonald, and Reynier (2002)**

# **5 A Conclusion**

## Representation of the interaction of flows

- Using an instantaneous **fluid picture** yields:  
An **optimisation** problem  
Data for Model : **Utility function.**
- Starting from **microscopic dynamics** yields:  
A **fixed point** equation  
Data for Model : **Equation coefficients.**

# Perspectives

- Modeling **transmissions arriving** from the **outside** world and **disappearing** after **completion**. A **different scaling**.
- More on the **invariant laws** for **on-off users**. **Transient behavior** of **on** periods needs to be assessed.
- Convergence of **invariant laws**:

$$\lim_{N \rightarrow \infty} \lim_{t \rightarrow \infty} = \lim_{t \rightarrow \infty} \lim_{N \rightarrow \infty} ?$$

Always **difficult**, especially with **multiple equilibria**.

- Relations with the **optimization problem** approach.

**Merci**  
**pour votre attention !**