

REALITY

Road Emission Activity-Link based InvenTorY

Megan Lebacque
Ecole des Ponts – parisTech
Marne la Vallée - France

Schematics of REALITY

Parking data

Network data,
dynamic traffic
volumes, and
average speeds

Complementary data:
Coefficients used in formulas
in BER calculation, weather
(temp, wind,
humidity) data, vehicle fleet,
fuel type

DYNABURBS:

Dynamic
Assignment for
Suburbs

REALITY:

Road Emission
Activity-Link
based InvenTorY

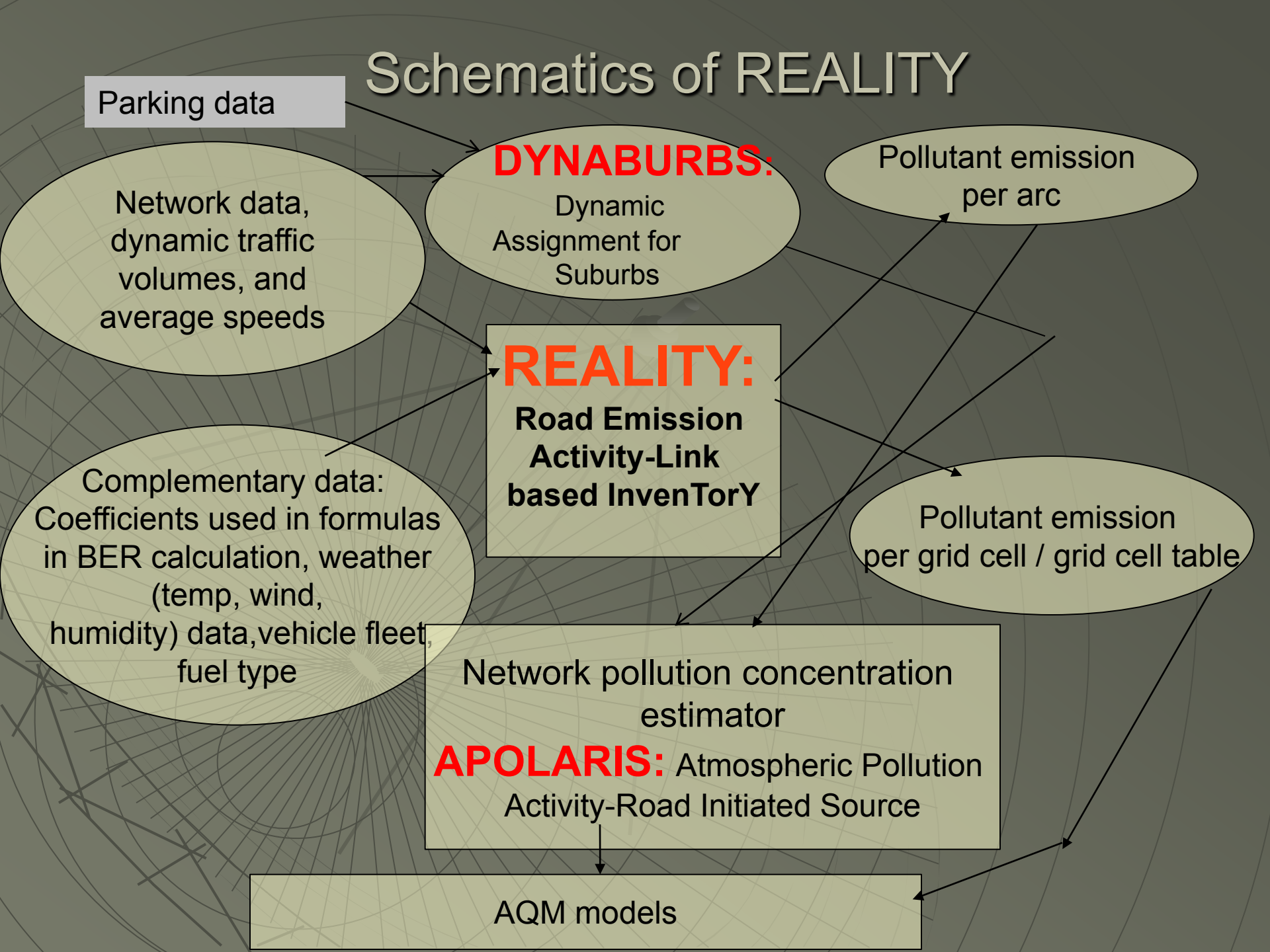
Pollutant emission
per arc

Pollutant emission
per grid cell / grid cell table

Network pollution concentration
estimator

APOLARIS: Atmospheric Pollution
Activity-Road Initiated Source

AQM models



Introducing REALITY

(sponsored by Institut Carnot Vitres)

REALITY is a **dynamic** model of emission calculation of pollutants that result from traffic on a road network. Calculation can be made on precise locations (roads) or for the entire network (divided into grid cells). REALITY calculates **hot emissions** (vehicles running on hot engines).

- **dynamic** : with respect to **time**

1. all day (24 hours)
2. Per hour
3. By fraction of an hour
4. For each instant of time

- **dynamic** : with respect to **traffic volume and speed**

1. **traffic volumes** on each road (link) of a network change as a function of time
2. **average speed** on each link varies by time of day

Introducing REALITY

- **dynamic** : with respect to **basic emission rates** (BER)
 1. emission rates are calculated as non-linear functions of average speeds on each link of the network and thus change as speeds change.
 2. Basic emission rates are calculated for each arc of the network
- **dynamic** with respect to **location**:
 - **Precise locations**:
 1. Per arc or per segment of arc
 2. Per grid cell (a collection of arcs)
 3. At a network level: a collection of arcs, or (a collection of grid cells)

Dynamic Traffic Assignment and REALITY

Network equilibrium dynamic traffic assignment: (New feature of the model REALITY)

- given a **variable matrix of origin – destination** travel demand, traffic volume is distributed among the links in a network in a way that the costs of taking these roads are equal in the network (the Wardrop principal). When due to change in **activity level** or **activity type** origin - destination matrices **vary in time**, traffic volumes and average speeds which are distributed **vary** respectively.

- link **average speeds** are calculated using the **fundamental diagram**, which gives the following relationship between traffic flow , density, and speed:

$$q(t) = k(t)v(t)$$

$$q(t) = \text{traffic flow during time interval (t)} \quad v(t) = Q(k(t))$$

$$k = \text{density during time interval (t)}$$

$$v = \text{average speed during time interval (t)}$$

Calculation of Basic Emission Rates (BER): REALITY

- ◆ BERs are calculated as functions of average speed, itself calculated by a network equilibrium dynamic assignment model

- ◆ $BER = f(v(p,k,m,t,i))$

An example of a speed equation

$$f(v(p,k,m,t,i)) = a * v(p,k,m,t,i) + b * v(p,k,m,t,i)^2 + c$$

- ◆ **BER** = **basic emission rate** (gr/km) per pollutant (p), for car class (k), and fuel type (m) during time interval (t) and per link (i).
- ◆ **a,b,c** are coefficients from COPERT adjusted for use in REALITY
- ◆ Equations follow **COPERT guidelines**
- ◆ **COPERT** is a European equivalent of **MOBILE6**
- ◆ $v(p,k,m,t,i)$ = **average speed** per pollutant (p), for car class (k), and fuel type (m) during time interval (t) and per link (i).

Calculating link pollutant emissions in REALITY

- ◆ Pollutant emissions are calculated on each link (i), for car class (k), and fuel type (m), during interval (t).

- ◆ **Pollutant emissions per link :**

$$E(p, i, k, m, t) = y(p, i, k, m, t) * v(p, i, k, m, t) * l(i)$$

$E(p, i, k, m, t)$ = is the emission of pollutant (p), on link (i), for car class (k), and fuel type (m) during time interval (t).

- ◆ $y(p, i, k, m, t)$ = is the emission factor for pollutant (p), link (i), car class (k), fuel type (m), and time interval (t).

- ◆ $v(p, i, k, m, t)$ = is the volume of car class (k) differentiated by fuel type (m) on link (i) and time interval (t).

- $l(i)$ = length of link (i) traveled by vehicles

Calculation of pollutant emissions by grid cell in REALITY

Total emission is calculated as the sum of link emissions multiplied by the fraction of each link in each cell.

◆ Emissions per grid cell :

$$\Phi_t^{p,k,j,m} = \sum E_t^{p,k,i,m} \times \alpha^{ij}$$

$\Phi_t^{p,j,k,m}$ = is the total emission of pollutant (p) for car class (k) with fuel intake of type (m) during interval (t) in grid cell (j); $j = 1, \dots, M$

$E_t^{p,i,k,m}$ = is link emission of pollutant (p), for car class ($k = 1, \dots, L$), with fuel intake of type (m)

$\alpha^{i,j}$ = is the fraction of link (i) in cell (j)

car class includes: type and age

Application of REALITY

- Case study: Ile de France (Paris metropolitan area)
- hot pollutant emission calculation for urban and non-urban (highways, expressways) on the Ile de France network
- hot pollutant emission calculation on grid level, where each grid contains a collection of arcs of the network
- graphical representation of the model application for CO, and NO_x

Île de France (Paris metropolitan) network

- Network size : **36583** arcs
- Network equilibrium dynamic assignment output: link flows , and average speeds **per time interval**

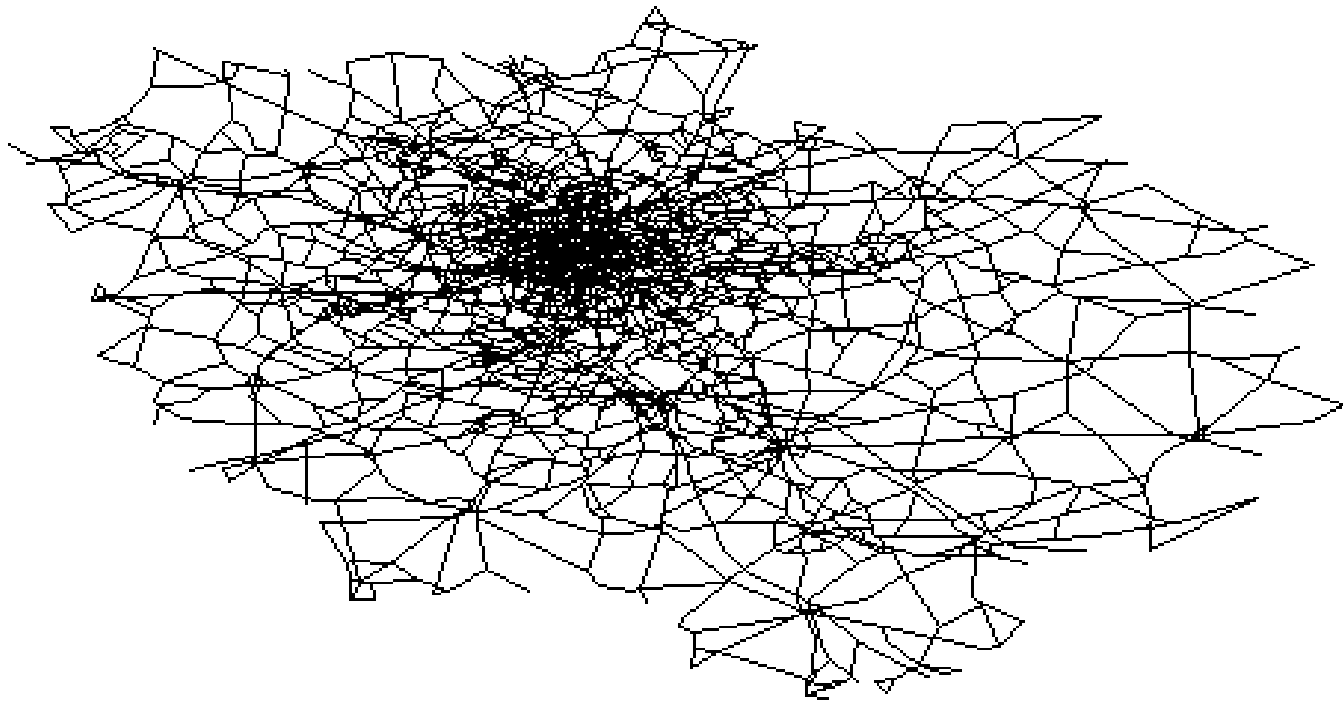
Time interval: hourly for 24 hours

- Pollutant emissions are calculated for each arc of the network of l'Ile-de-France by grid cell: **grid cells of size 0.5 degrees** longitude and latitude: total number of grid cells: (43 x 24 grid cells)
- Each grid cell contains several links
- The links are either entirely within a grid cell or pass by 2 or more grid cells.

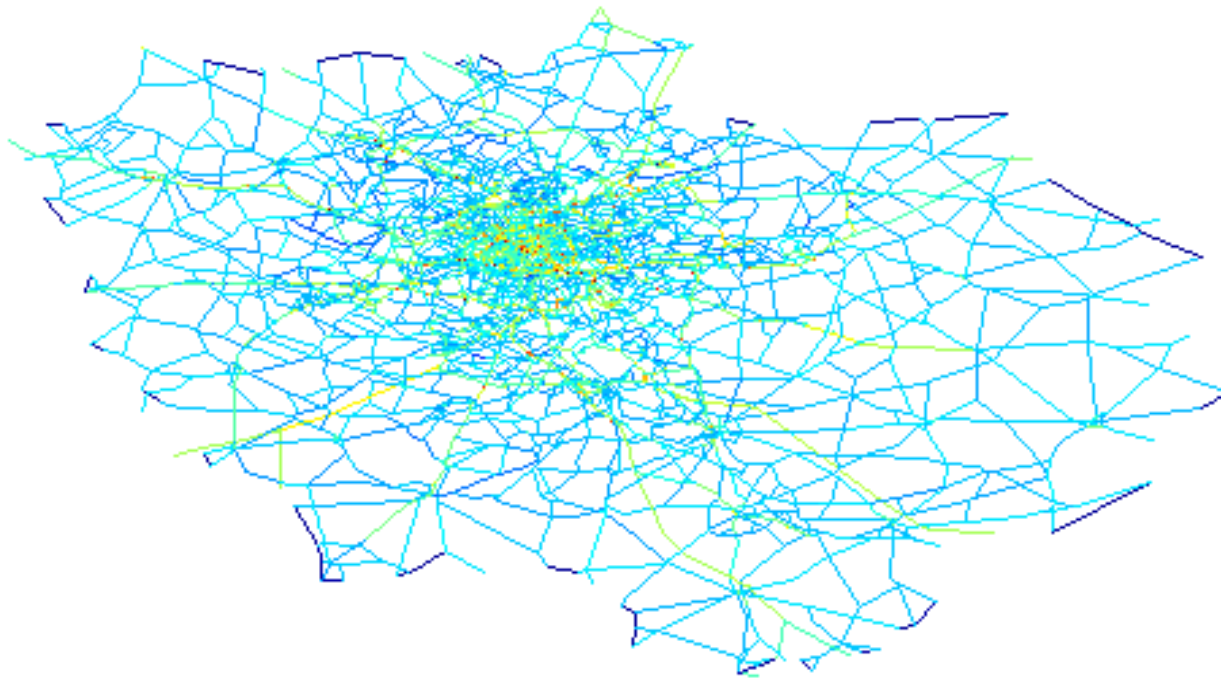
Grid cell emission is calculated by multiplying link emissions by the fraction of links in each grid cell and then added up

Color codes: blue (low emission), red (high emission)

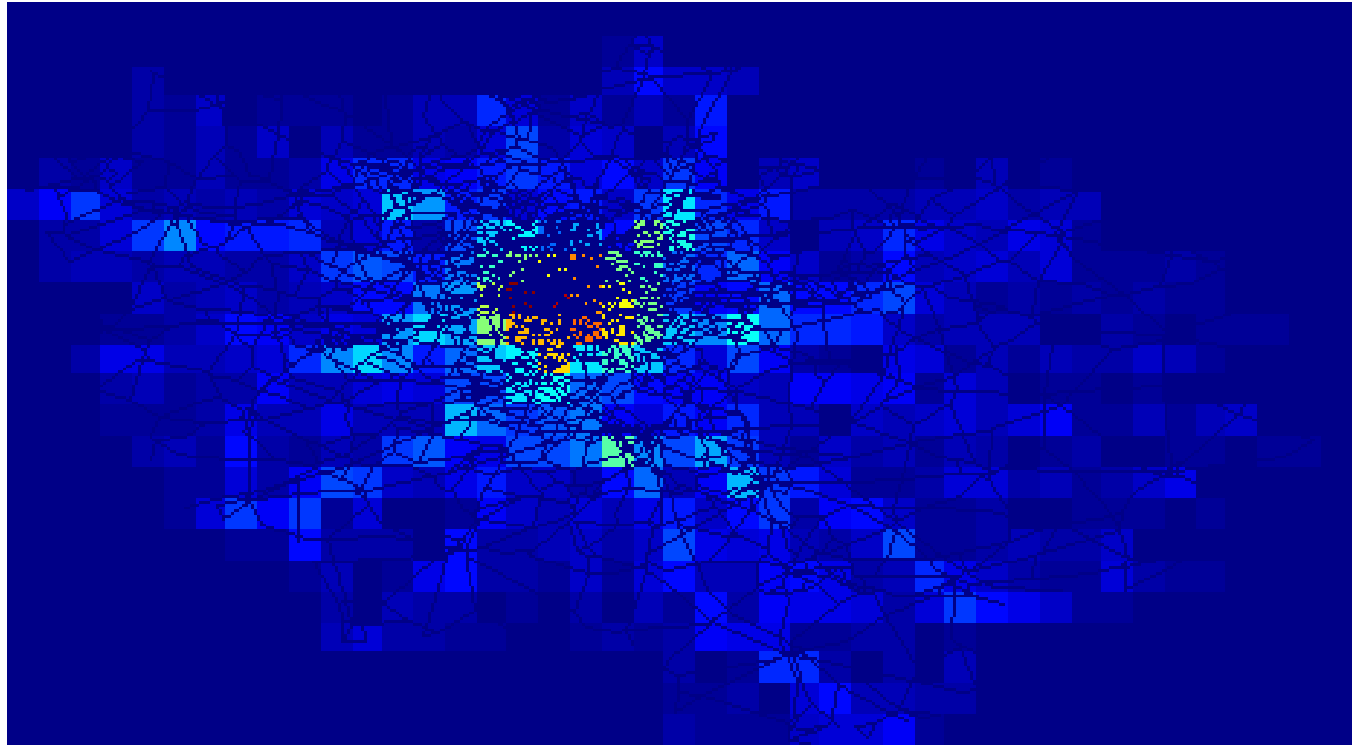
Road network - Île de France (Paris and all suburbs)



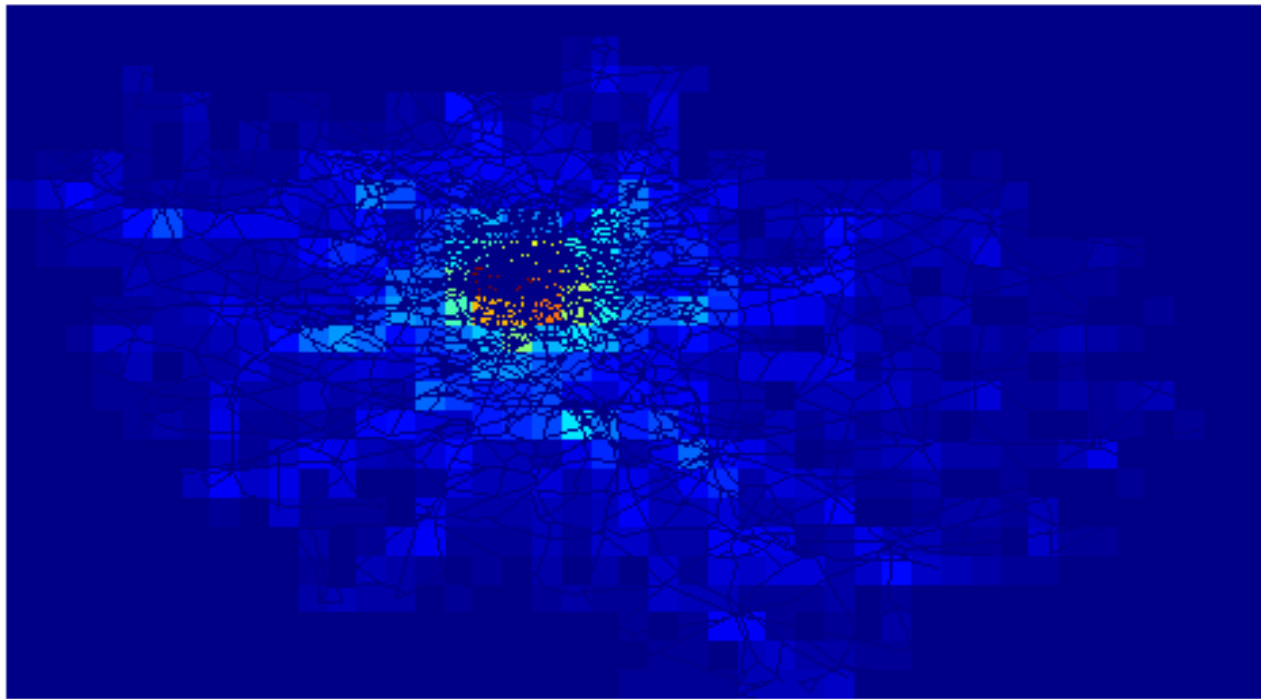
CO emissions – grams- private cars –
gasoline- Île de France – at 8h00 a.m.- by
link



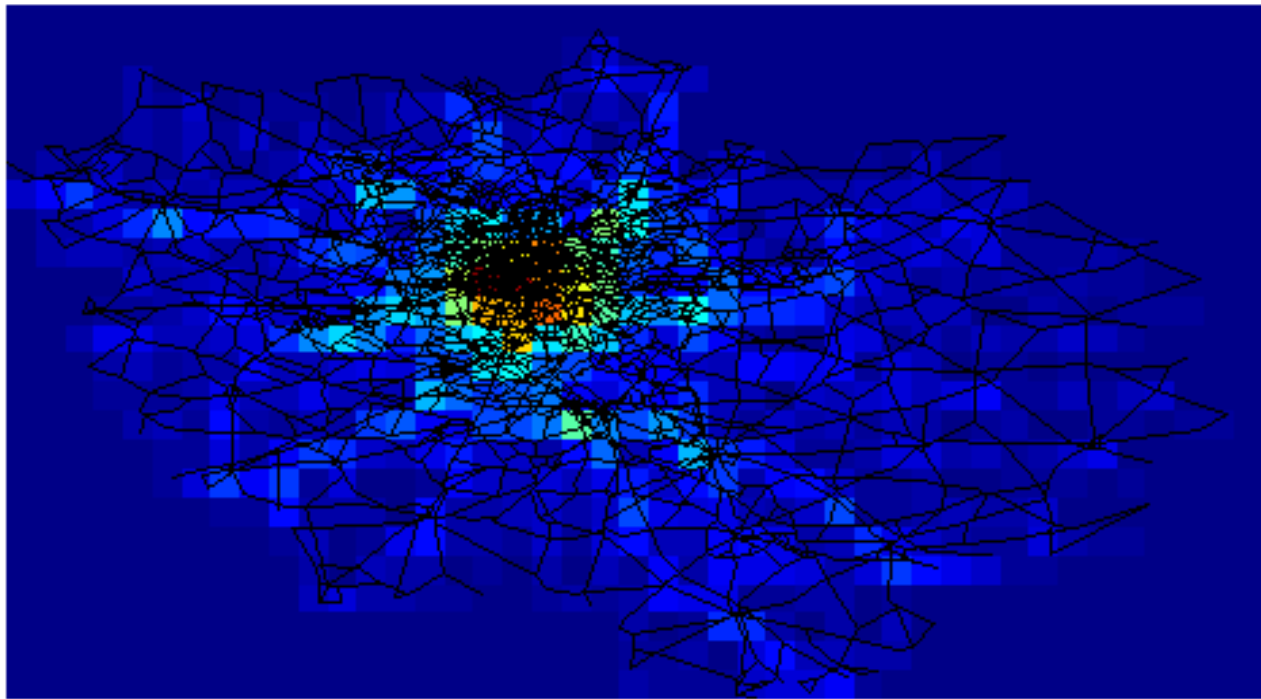
CO emissions – in grams- trucks – diesel- Île de France – at 7h00 a.m.- by grid cell



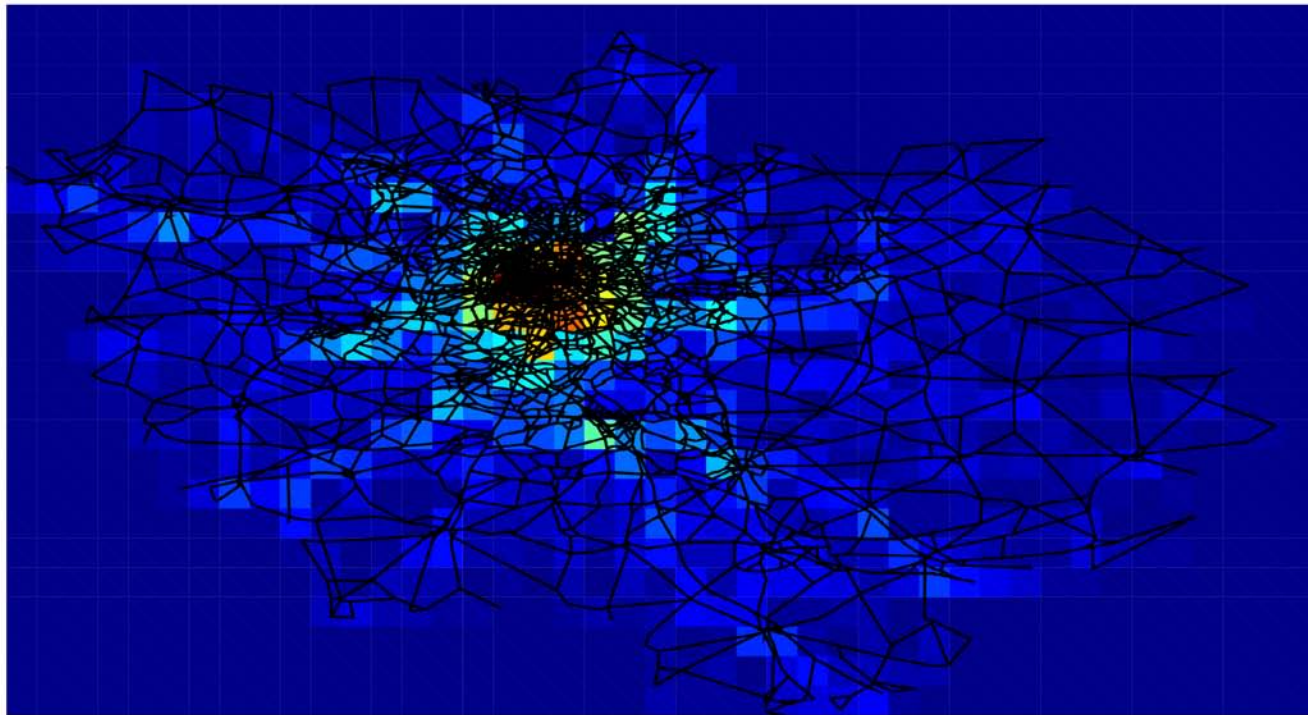
CO emissions – in grams- private cars –
gasoline – Île de France – at 7h00 a.m.– by
grid cell



NOx emission— grams- trucks— diesel - Île de France— à 7h00 - by grid cell



NOx emission – grams- cars – gasoline - Île de France –
at 7h00 a.m. by grid cell



Dynamic assignment, trip chaining, parking and cold emissions : **DYNABURBS**

DYNABURBS : Dynamic Assignment for Suburbs

A dynamic assignment model with trip chaining and parking option.

trip chaining is defined as the **number of stops** a road user makes between an origin and destination due to **non-work activities** (example: dropping kids to school, shopping, doctor's appointment, or cultural and recreational activities).

The output of the **Dynamic Assignment** coupled **with trip chaining** and **parking option** model is used in **cold emission** estimation

DYNABURBS : Dynamic Assignment for Suburbs

Network characteristics:

DYNABURBS is designed for networks that connect a **small** number of origins and destinations such as networks that connects **suburbs** to suburbs or suburbs to **city centers**.

The **arcs** of such networks are usually **urban roads** that allow road side and /or garage parking

DYNABURBS : Dynamic Assignment for Suburbs

An example: origin (a) and destination (b)

Origin (a) is connected to destination (b) by two arcs (1) et (2).

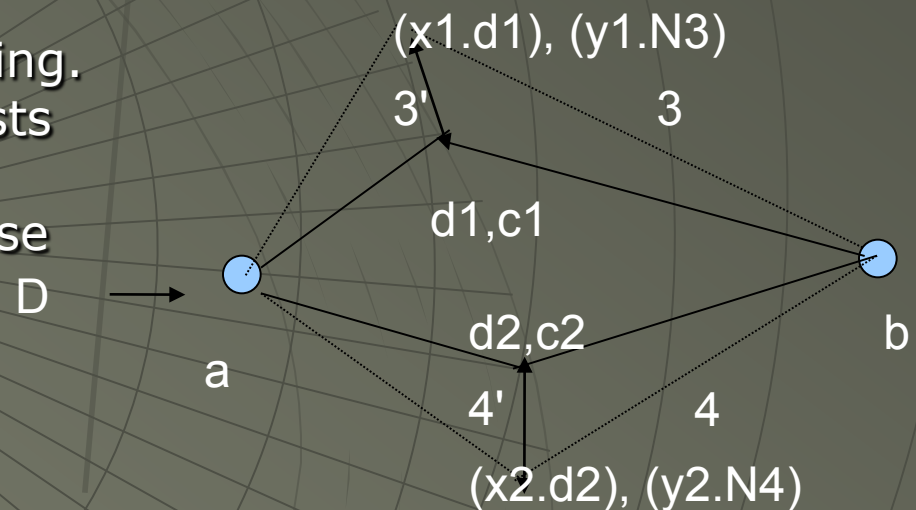
The two auxiliary arcs (3) and (4) represent parking (either curb side parking or garage parking)

arcs (3'), and (4') are « dummy » links and represent access to parking. No travel time costs or parking costs are associated with these dummy links. Users can enter and exit these arcs free of charge.

Total demand = D

$$D = d1 + d2$$

$c1(d1)$ = cost of driving on arc (1) which is the function of demand on that arc.



DYNABURBS : Dynamic Assignment for Suburbs

$c2(d2)$ = Cost of traveling on arc (2)

$x1$ = Fraction of users that exit the main traffic on arc (1) and park on link (3) ($0 < x1 \leq D$)

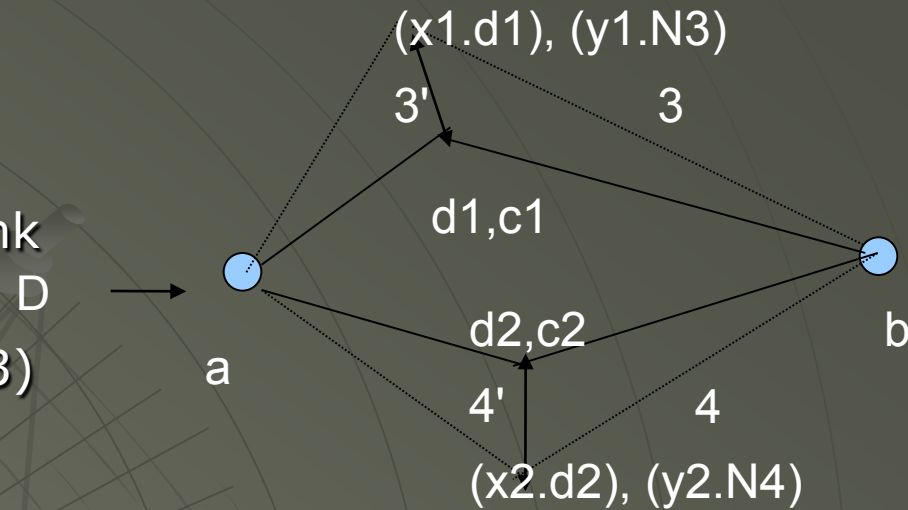
$y1$ = Fraction of users that exit arc (3) and enter the main traffic on arc (1) ($0 < y1 \leq D$)

$N3$ = Number of parking spots occupied on arc (3)

$x2$ = Fraction of users that exit arc (4) and enter the main traffic on arc (2) ($0 < x2 \leq D$)

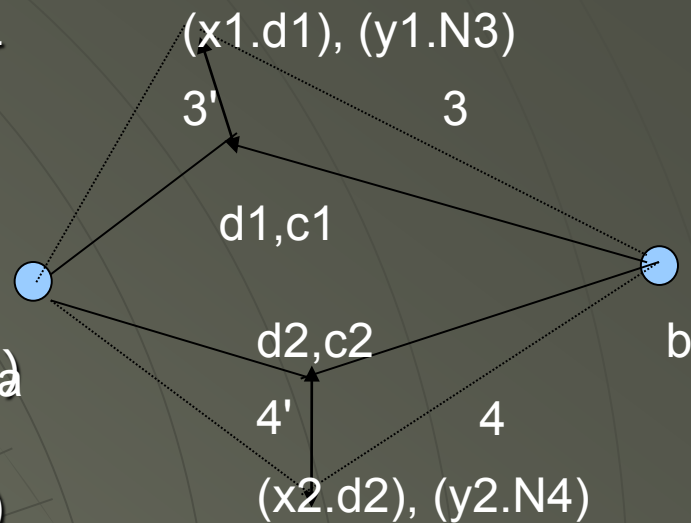
$y2$ = Fraction of users that exit arc (4) and enter the main traffic on arc (2) ($0 < y2 \leq D$)

$N4$ = Number of parking spots occupied on arc (4)



DYNABURBS : Dynamic Assignment for Suburbs

- During each time interval: (Δt) there exist a fraction of users $\{x1(t), t=1, \dots, \Delta t\}$ and a fraction of users $\{x2(t) t=1, \dots, \Delta t\}$ such that ($x1 \neq x2$)
- Similarly there exist a fraction of users (y): $\{y1(t), t=1, \dots, \Delta t\}$ and a fraction of users $\{y2(t) t=1, \dots, \Delta t\}$ such that ($y1 \neq y2$)
- **Dynamic assignment** in this context: to distribute the number of users that go from (a) to (b) during each time interval (Δt) between arcs (1), and (2) in such a way that the network is at equilibrium (costs on arcs (1), and (2) are equal, given the parking option represented by arcs (3), and (4).
- **The idea is:** that (x) et (y) are **random** variables, and as a consequence the number of **vehicles parked** are also variables.
- Users that leave parking during the time interval (Δt) enter the main traffic stream at rate (y).



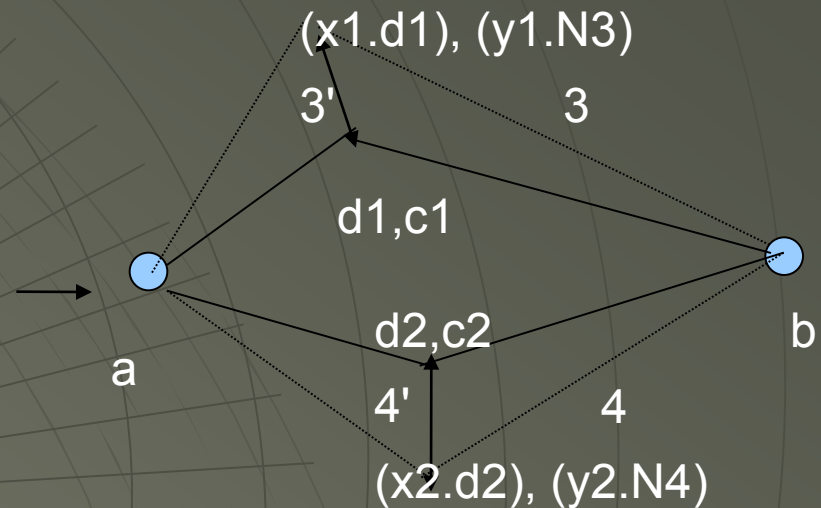
DYNABURBS : Dynamic Assignment for Suburbs

The outcome of a dynamic assignment model gives:

$$c_1(d_1(t), t) = c_2(d_2(t), t)$$

volumes $d_{e1}(t)$, and $d_{e2}(t)$, speeds ($v_{e1}(t)$ et $v_{e2}(t)$) are values at equilibrium and so are $N_{e3}(t)$, $N_{e4}(t)$, the number of cars parked on arcs (3), and (4)

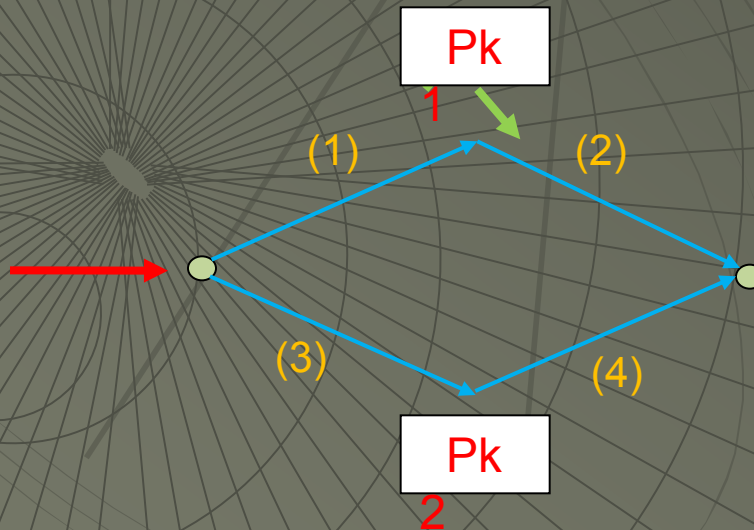
example: $y_1 * N_{e3}(t)$ = the volume of traffic that runs on cold engine and enters arc (1) at network equilibrium



Program DYNABURBS

Application of DYNABURBS: a simple network

- (1) **two types of users**: those who go from an origin to a destination without stopping on the way, those who park in between the origin and the destination
- (2) there are two trip chaining possibilities: either parking at (origin-destination) or parking at parking lot (1) or (2).
- (3) vehicle type: private cars running on gasoline and diesel
- (4) possibility of parking on each arc



DYNABURBS

(5) **Parking rate** is assumed to be fixed at (φ)

(6) The **exit rate** from a parking garage or a side street parking spot is fixed at (v)

(7) The **number of vehicles parked** in a garage or alongside streets is ($N1$) and ($N2$) vary as a function of vehicles that enter and exit the parking

$$\frac{dN}{dt} = \varphi * d - v * N$$

(8) The « **Wardrop** » equilibrium concept is used which means that if links are used then they have to have the same cost

$$C1(d_1) = C2(d_2)$$

$$D = d_1 + d_2$$

(9) The **Jin** method is used to calculate the Wardrop equilibrium

$$\frac{dd_i}{dt} = -d_i (C_i(d_i) - \bar{C})$$

$$\bar{C} = \frac{\sum d_i * (C_i(d_i))}{D}$$

DYNABURBS

The procedure applied is as follows:

- **pathFlows** at equilibrium are calculated
- The number of **cars parked** at equilibrium are calculated
- The number of **vehicles running on cold** engine at equilibrium: these are vehicles that leave the parking after starting their engines, are calculated
- At network equilibrium cold emissions and hot emissions of pollutants are calculated

At **dynamic** equilibrium:

- the number of vehicles **parked** vary in time
- The number of **vehicles parked** affects the **equilibrium** which means that during each time interval a new network equilibrium is calculated as a function of the number of cars parked in the previous interval.
- since **cold emissions** are calculated as functions of **vehicles parked**, then cold emissions **change** during each time interval

An exemple of DYNABURBS

- the simple network with two links is revisited:
- Running DYNABURBS (in Scilab) gives the following results:
given $D = 5400$ cars

- Φ_i = fraction of vehicles that park at equilibrium

$$\text{arc}(1) = 0.3406603 \quad \text{arc}(3) = 0.1163102$$

- Nu = fraction of vehicles that enter the traffic stream after being parked

$$\text{arc}(2) = 0.4663210 \quad \text{arc}(4) = 0.3439935$$

PathFlows = traffic volume at equilibrium

$$(1) + (2) = 3425.2019 \quad (3) + (4) = 1974.7981$$

An example of DYNABURBS

NbVhPk = number of vehicles parked at equilibrium

1492.7815 865.49751

Nu(2)*NbVhPk (2) = number of vehicles running on cold engine on arc (2)

608.43883

Emission_VI_gas_H_CO = hot emission of CO – gasoline (grams)

1141.5196 2671.2764 1905.7792 1055.2549

Emission_VI_gas_C_CO = cold emission of CO – gasoline (grams)

3454.2036 1750.3366 12803.709 816.55459

Emission_VI_dis_C_CO = hot emission of CO – diesel (grams)

108.57572 24.647047 50.337814 9.6717734

Emission_VI_dis_H_CO = cold mission of CO – diesel (grams)

410.53761 498.07075 288.21765 178.08296

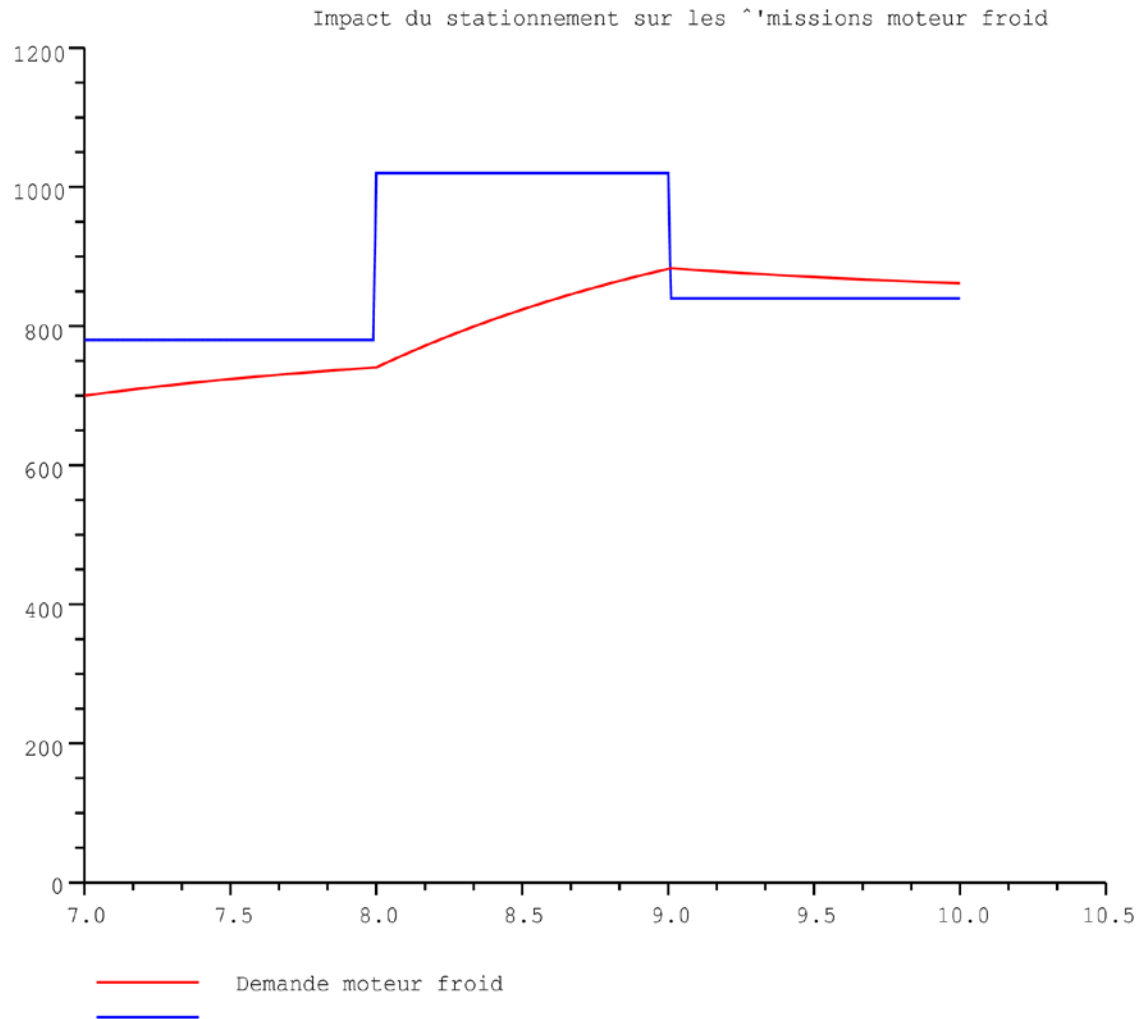
DYNABURBS

observations:

- as the **volume of traffic varies** in time, the number of vehicles parked (**N**) **vary** accordingly.
- the number of vehicles running on **cold engine** is equal to the number of vehicles that have left **parking garages** and side streets parking places
- if the number of vehicles running on cold engine were estimated as ($\phi \times \text{demand}$), this estimation would have given **systematic errors**
- Thus, the number of vehicles parked (**N**) should be calculated first and the number of vehicles running on cold engine should be calculated as a function of (**N**)

DYNABURBS

The impact of time varying parking pattern on cold emission estimation:

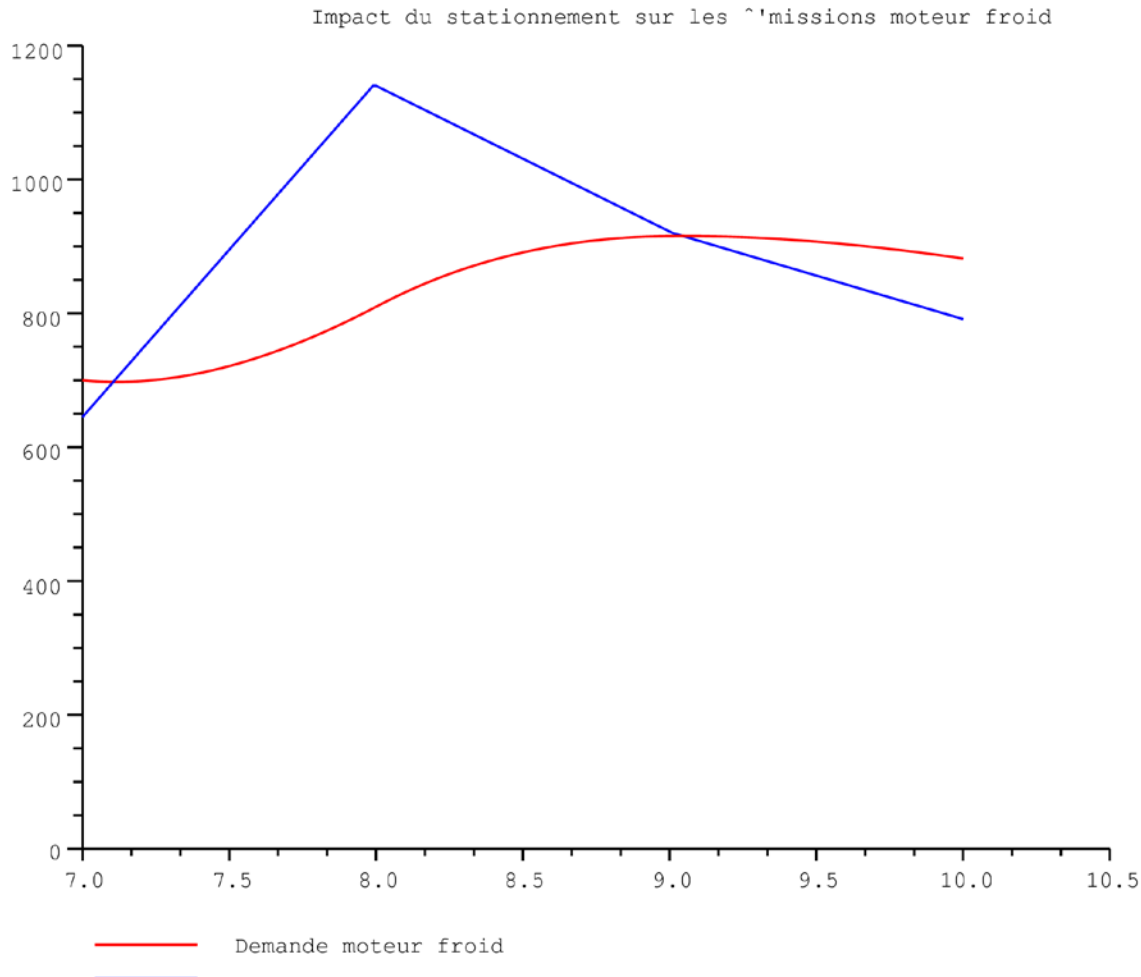


x-axis is time and
y-axis is cold
emissions.

Blue line represents
cold emissions based
on link flows which
are piece wise
constant functions of
time . **Red line**
represents cold
emissions based on
link flows and parking

DYNABURBS

The impact of time varying parking pattern on cold emission estimation:



x-axis is time and **y-axis** is cold emission. **Blue line** represents cold emissions based on link flows which are piece wise linear functions of time . **Red line** represents cold emissions based on link flows and parking

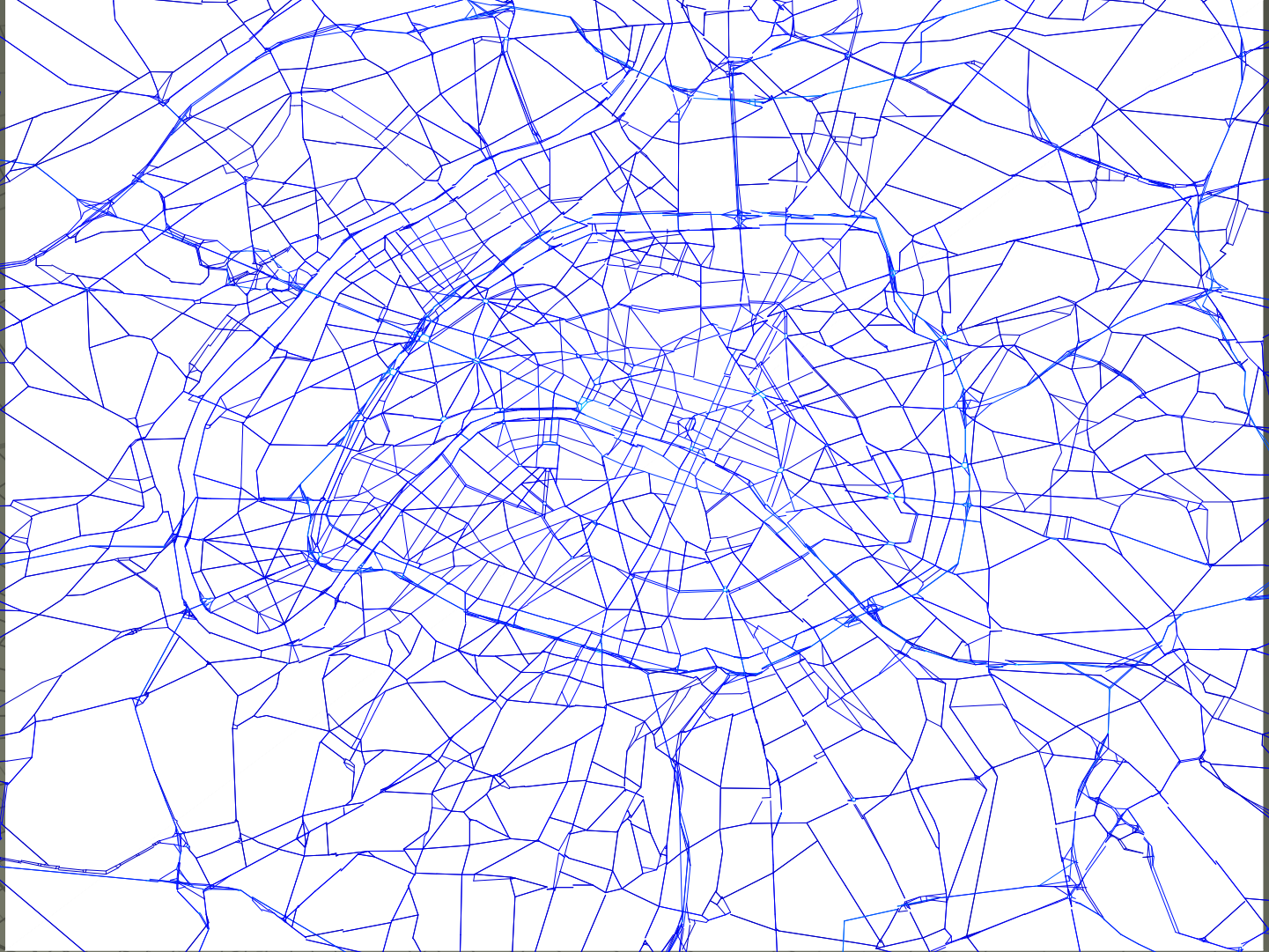
Pollutant emission.

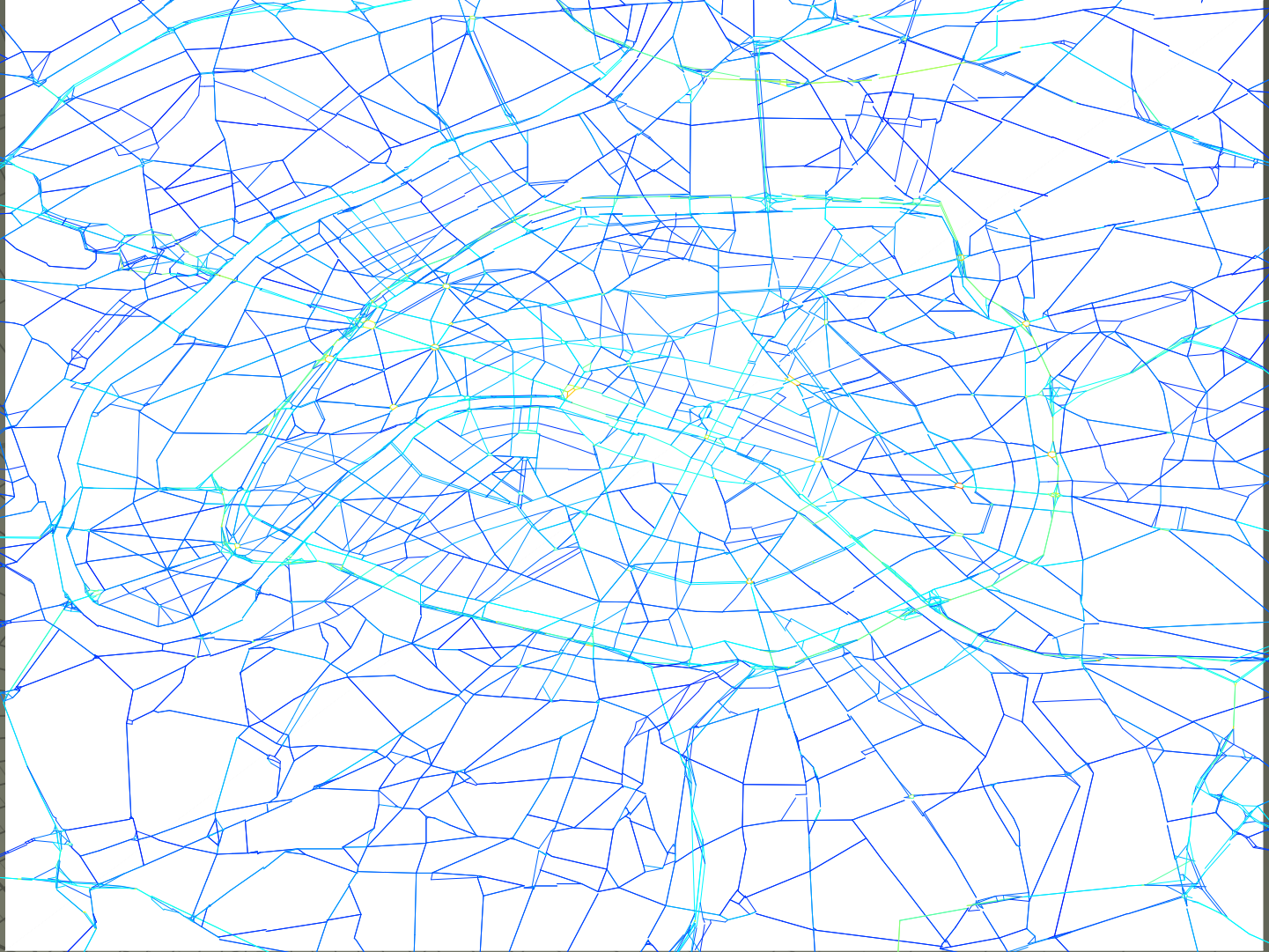
Paris area, morning (6 to 9 pm)

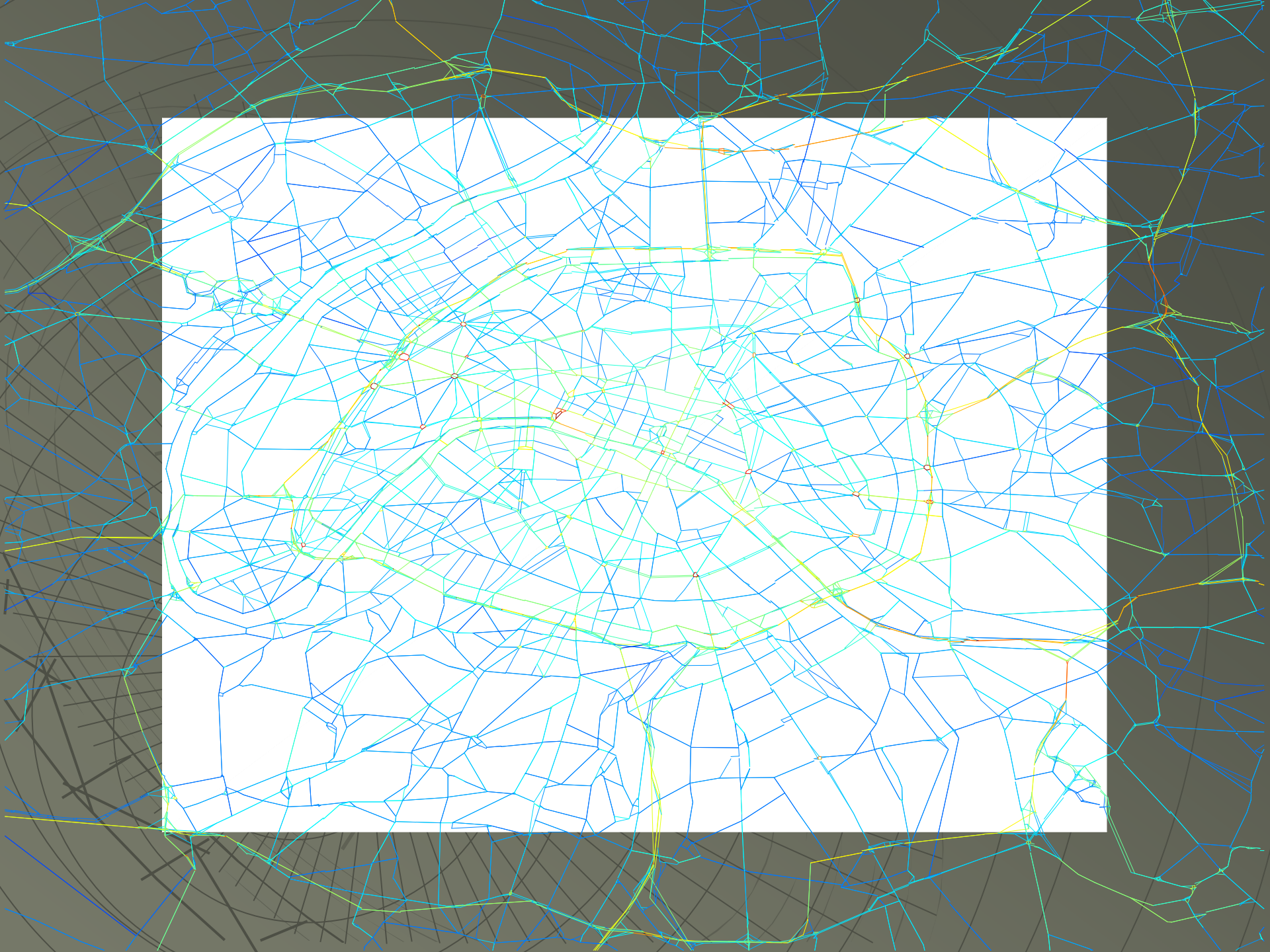
CO emissions (hot and cold emissions)

Dynamic traffic volume, speeds, parking

Cars running on gasoline







Correction factors in REALITY

There are two variables in the model that can be corrected using correction factors:

- Basic emission rates (**BER**) are functions of **speed**.

The **question** is which **speed equation to use**, and whether the speed function chosen is representative of **what goes on the roads ?**

- how to apply **correction factors** to speed measurements?

- let's denote the speed correction factor by (**SCF**). what method should be used in determining correction factors ?

- **extreme Temperatures** impact speeds. Thus accurate temperature measurement and correction factors are needed. Let's denote the temperature correction factor by (**TCF**)

Correction factors in REALITY

- which speed equations to choose for BER calculation?

The approach used is « bootstrapping & confidence interval method »

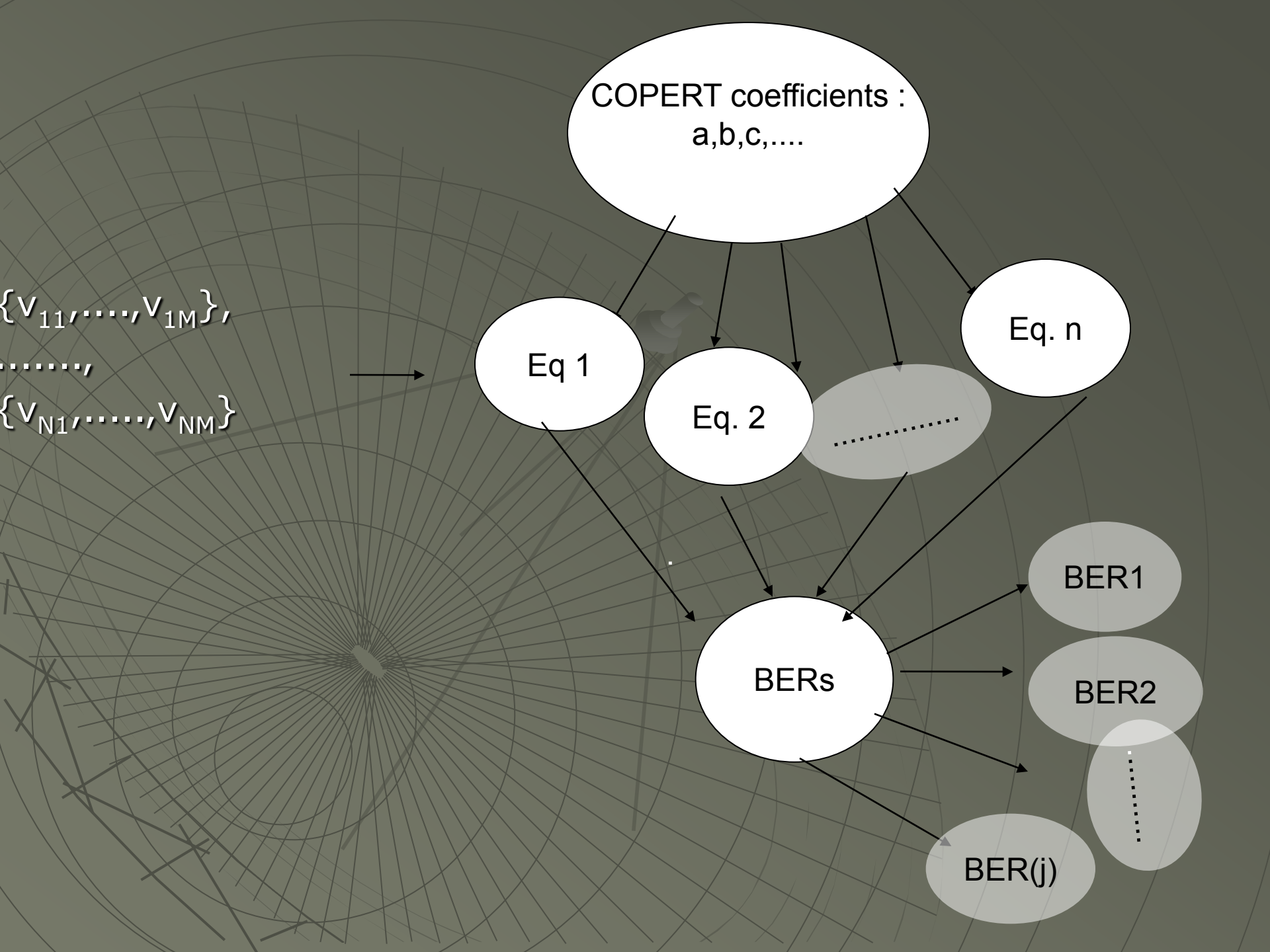
i = number of samples ; $i = \{1, \dots, N\}$ = (example: different places in a road network)

j = the number of arcs in each sample (i); $j=M$, (all samples have a fix number of arcs)

v_{ij} = a matrix of average speeds

example: $\{v_{11}, \dots, v_{1M}\}$, v_{1M} = average speed in sample (1), on arc (M);

In general: v_{ij} = average speed in sample (i), arc (j);
 $i=1, \dots, N$; $j= 1, \dots, M$



COPERT coefficients :
a,b,c,....

Eq 1

Eq. 2

.....

Eq. n

BERs

BER1

BER2

.....

BER(j)

$\{V_{11}, \dots, V_{1M}\}$,
.....,
 $\{V_{N1}, \dots, V_{NM}\}$

Correction factors in REALITY

- the **BERs** are calculated for each of the equations (Eq.1,...Eq.n) and for each arc.
- the **estimated** BERs are then compared:
 - If there are **no variations** among these estimated BERs, and among the (N) different samples, then any of the (n) equations can be used for BER estimation.

If on the other hand, there **are variations** among these estimated BERs, and among the (N) different samples, then:

if the BER are **under estimated** in comparison with other sources of BER estimation, then let's denote these BERs by (e^i_j)

- **apply** SCF to average speeds
- **modify** coefficients (a,b,c,...) by adding **white noise** (normally distributed $N(0, \sigma^2)$)
- **recalculate** BERs , if **no variations**, then can choose among any of the modified equations, otherwise, repeat the process until convergence obtained.

Correction factors in REALITY

- if the BERs are **over estimated** in comparison with other sources of BER estimation, then let's denote these BERs by (e^h_i)
 - **apply** SCF to average speeds
 - **modify** coefficients (a,b,c,...) by adding **white noise** (normally distributed $N(0, \sigma^2)$)
 - **recalculate** BERs , if **no variations**, then can choose among any of the modified equations, otherwise, repeat the process until convergence.
- if the BERs are either **under estimated** and **over estimated** in comparison with other sources of BER estimation, then let's denote these BERs by (e^v_i)
 - **apply** SCF to average speeds
 - **modify** coefficients (a,b,c,...) by adding **white noise** (normally distributed $N(0, \sigma^2)$)
 - **recalculate** BERs , if **no variations**, then can choose among any of the modified equations, otherwise, repeat the process until convergence.

Correction factors in REALITY

- Calculation of mean BERs for each sample:

$$\bar{t}_i = \frac{\sum_{j=1}^M (e_j^v)}{M} \quad i=1, \dots, N$$

\bar{t}_i = mean BER per sample

M = number of arcs in each sample

e_j^v = BER values: v = signifies either **over** or **under** estimation

- Calculation of variance and standard deviation for each sample:

$$S_i^2 = \frac{1}{(M-1)} \sum_{j=1}^M (e_j^v - \bar{t}_i)^2 \quad i=1, \dots, N$$

S^2 = variance

s = standard deviation

$$s = \sqrt{S_i^2}$$

Correction factors in REALITY

\bar{t} = total mean BER

\bar{t}_i = mean BERs per sample

$$\bar{t} = \sum_{i=1}^N \bar{t}_i$$

- Confidence interval

$$\left(\bar{t} \mp 0.90 \left(\frac{s}{\sqrt{N}} \right) \right)$$

- The **BERs** are recalculated applying **SCF** to adjust speeds:
- if the recalculated BERs fall in the confidence interval, then any of the speed equations used to calculate BERs are acceptable .

Speed correction factor: (SCF)

- « **bootstrapping** » and « **confidence interval** » methods are used in finding the speed correction factor.
- Several **data samples** are taken:
 - for example **average speeds** resulting from several runs of a dynamic assignment model.
 - let $i = 1, \dots, K$; be the number of data sets
 $K =$ the maximum number of samples
 - each data set contains **a fixed** number of arcs
 - Though the number of links are fixed, their **types** vary from highways, and expressways to urban streets.

let x_{11}, \dots, x_{M1} ; x_{Mi} be average speeds on each link (i)

Correction factors: speed (SCF)

- **Average speed** is calculated as the mean of all link average speeds

$$\bar{x}_i = \frac{(x_{1,i} + \dots + x_{M,i})}{M}$$

$$i = 1, \dots, K$$

- Let $x_{M,i} \sim N(0, \sigma^2)$ be normally distributed
- calculation of residuals:

$$e_i = (x_i - \bar{x}_i)^2$$

- let $e_i \sim \sigma^2 \chi^2(M-1)$ have a chi distribution with (M-1) degrees of freedom
- If the population standard deviation is known, (σ), the confidence interval can then be calculated:
- **Total average speed** for all samples

$$\bar{x} = \frac{\sum_{i=1}^K \bar{x}_i}{K}$$

Correction factors: speed (SCF)

- The statistic (Z):

$$Z = \frac{(\bar{x} - \mu)}{\left(\frac{\sigma}{\sqrt{K}}\right)}$$

- if $1 - \alpha = 0.95$ (95% confidence interval)

$$P(-z \leq Z \leq z) 1 - \alpha = 1 - 0.95$$

$$\varphi(z) P(Z \leq z) 1 - \frac{\alpha}{2} = 0.975$$

$$z = \varphi^{-1}(\varphi(z)) = \varphi^{-1}(0.975) = 1.96$$

$$0.95 = 1 - \alpha = P \left(-1.96 \leq \frac{(\bar{x} - \mu)}{\left(\frac{\sigma}{\sqrt{K}}\right)} \leq 1.96 \right)$$

$$P \left(\bar{x} - 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \leq \mu \leq \bar{x} + 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)$$

- μ is the population mean

Correction factors: speed (SCF)

-the confidence interval is

$$\left(\bar{x} - 1.96 \left(\frac{\sigma}{\sqrt{K}} \right), \bar{x} + 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)$$

- (SCF) in the case of over estimation of speed is given as:

$$SCF = \frac{1}{\left(\bar{x} - 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)}$$

$$\tilde{v} = v^h * SCF = \frac{v}{\left(\bar{x} - 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)}$$

$v \sim$ = adjusted (corrected) speed

v^h = over-estimated speed

Correction factors: speed (SCF)

- The speed correction factor (SCF) in the case of **under** estimation is given

$$SCF = \left(\bar{x} + 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)$$

If the **average speed** is **under** estimated, then the estimated speed \tilde{v} is given as:

$$\tilde{v} = v^l * SCF = v * \left(\bar{x} + 1.96 \left(\frac{\sigma}{\sqrt{K}} \right) \right)$$

\tilde{v} = corrected speed

v^l = under estimated speed

Correction factors : temperature (TCF)

- The following method is used to estimate **temperature** correction factor:
- for each day of the month, maximum and minimum temperatures are recorded and denoted by (T_{max} , T_{min})
- daily averages are calculated as

$$\hat{T} = \frac{\sum (T_{max} + T_{min})}{2}$$

- Monthly average is calculated as:

$$\bar{T} = \frac{\sum (\hat{T})}{31}$$

- standard deviation is calculated as:

$$S^2 = \frac{\sum_{i=1}^{31} (T_i - \bar{T})^2}{30}$$

$(n-1) = 30$

Correction factors : temperature (TCF)

$$s = \overline{S^2}$$

- The temperature correction factor is calculated as:

$$TCF = \frac{(T_{obs} - \bar{T})}{\left(\frac{s}{\sqrt{31}}\right)}$$

- If the observed or forecast temperature is greater than the monthly average, then it is corrected in the following manner:

$$BER_{mod} = BER_{old} * (1 + TCF) \quad (T_{obs} > \bar{T}_{month}) \quad (mod = modifier)$$

$$T_{obs} < \bar{T}_{month}$$

$$BER_{mod} = BER_{old} * (1 + TCF)$$

Pollutant concentration for a road network

- the conventional method :
 - each of the links of a road network is considered a linear source of pollution – (**line source modeling**)

- Only those links are considered that have the « **urban street canyons** » effect

- **mathematical models** used in calculating linear micro scale pollution concentration:

Street canyon models

Building wake models are two examples of this type of modeling

(source: Sprin, A. Air Quality at street level: Strategies for Urban Design. Cambridge: Harvard Graduate School of design (1986))

- **significant factors**:

- Pollutant **emission** levels

- **Air circulation**, depends on the wind, its speed and temperature

- **Street isolation level**: is defined by the shape and the height of buildings that surround the street

- **Absorption rate of** pollutant particles by different **materials** in the environment such as building materials, materials used to pave roads, etc.

Road level pollutant concentration: **APOLARIS**

APOLARIS : Atmospheric Pollution Activity-Road
Initiated Source

The **objective** of this model is to calculate pollutant (CO, VOC, NO_x, CO₂, SO_x) concentration from traffic emission

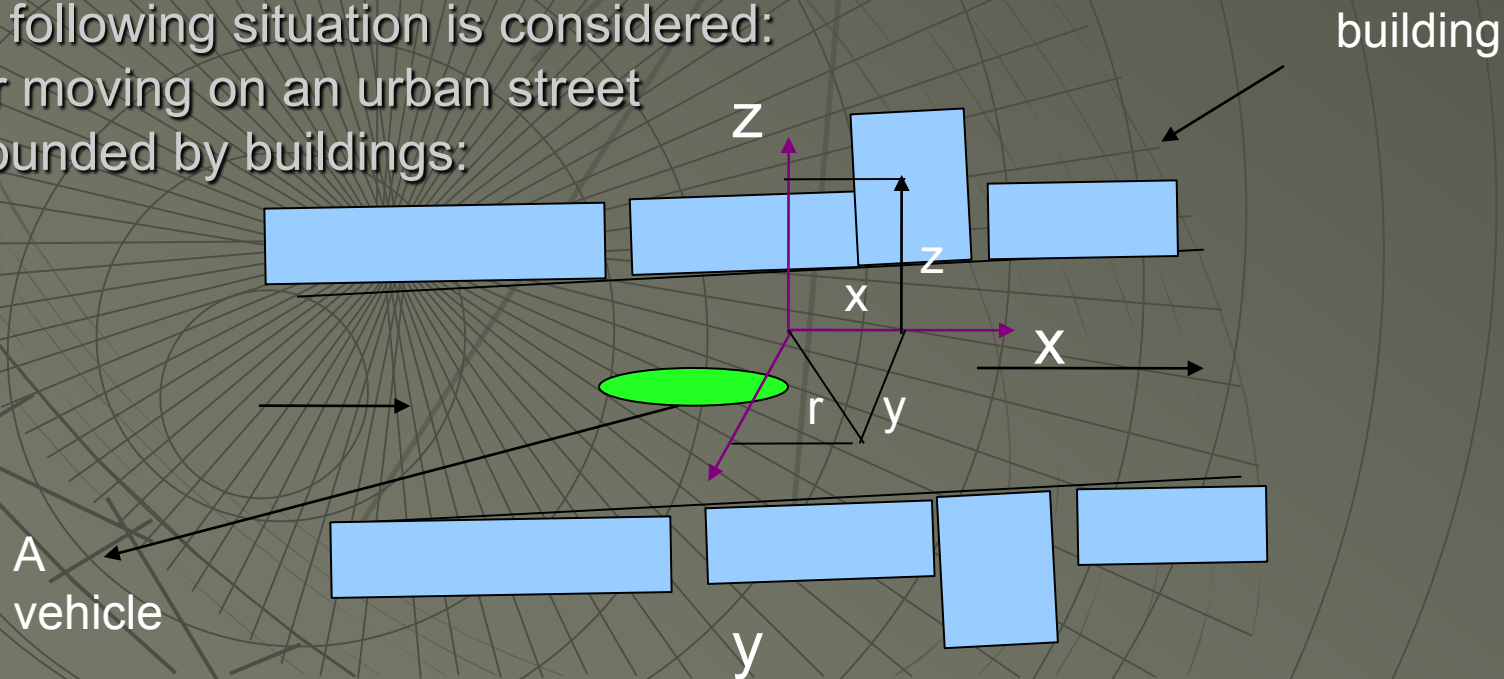
A **car** is considered to be a **linear source** of pollution emission

- To calculate **total concentration**, pollutant concentration produced on a road should be added to the pollutant concentration from **fixed source** emissions on that road. Fixed source emission in this context is pollution emitted from the **surrounding buildings** and **human activities** other than traffic.

Micro level pollution concentration: road pollutant Concentration estimation : **APOLARIS**

APOLARIS : Atmospheric Pollution Activity-Road Initiated Source

- meteorological considerations: wind intensity is assumed to be lower on urban streets. Wind intensity is assumed to be fixed during the concentration calculation.
- it is assumed that temperature is higher around urban streets .
- the following situation is considered:
a car moving on an urban street
surrounded by buildings:

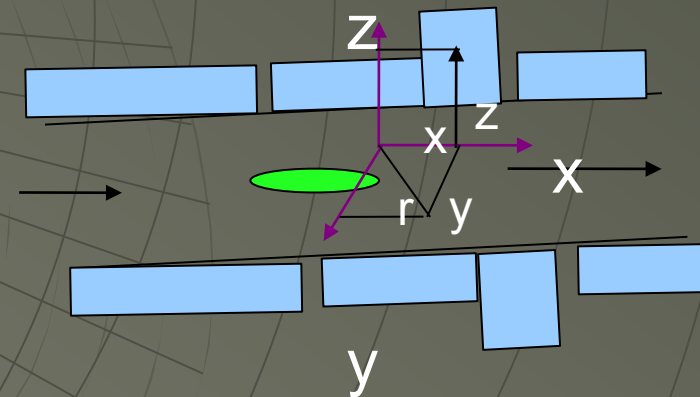


APOLARIS

-the CTM (Chemistry Transport Model) is used to calculate road pollution concentration, but with some modifications :

$$\begin{aligned} & \frac{dC}{dt} + U \left(\frac{dC}{dx} \right) + V \left(\frac{dC}{dy} \right) + W \left(\frac{dC}{dz} \right) \\ & \frac{d}{dx} \left(k_n \frac{dC}{dx} \right) + \frac{d}{dy} \left(k_h \frac{dC}{dy} \right) + \frac{d}{dz} \left(k_z \frac{dC}{dz} \right) \\ & + R + Q + \Psi + (\alpha \times \Phi) \end{aligned}$$

- **1st step**: the concentration of pollutants in the (X) direction is calculated , the solution is denoted by $C_t^{i*}(x)$



APOLARIS

- **2nd step**: the concentration of pollutants in the (Y) direction is calculated the solution is denoted by $C_t^{i*}(y)$
- **3rd step**: the concentration of pollutants in the (Z) direction is calculated the solution is denoted by $C_t^{i*}(z)$

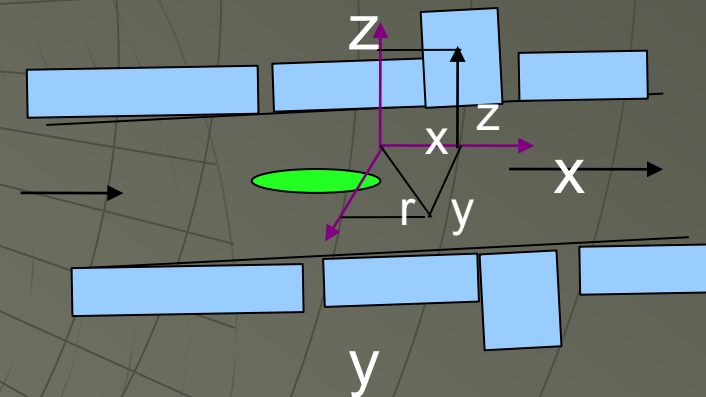
- total pollution concentration on arc (i) is calculated as:

Eigen-functions of the static part of the CTM are calculated . These eigen-functions are decomposed as products of concentration functions of x, y and z

$$C_t^{\text{total}} = C_t^{i*}(x) C_t^{i*}(y) C_t^{i*}(z)$$

So first partial one-directional problems are solved

Then the solution to the CTM is obtained as a weighted sum of the eigen-functions



APOLARIS

C = pollution concentration

U, V, W = wind components

U = wind in the east-west direction

V = wind in the north-south direction

W = wind in the vertical direction

K_h = horizontal turbulent diffusion

K_z = vertical turbulent diffusion

APOLARIS

E = represents particle movement (used in plume modeling)

R = speciation

Q = pollutant emissions from traffic = **pollutant emissions** on the links of a network

D = quantity of pollutants absorbed by a dry surface

W = quantity of pollutants absorbed by a wet surface

Φ = quantity of pollutants absorbed **Street isolation level**

α = absorption rate

Ψ = quantity of pollutants absorbed by humans: function of the density of activities

- to adapt **Chemical Transport** (TC) to road level :

The streets are considered as « canyon streets » and it is considered that vehicles are moving objects that have a linear trajectory

APOLARIS

- The following **hypothesis** are made:
 - during any interval (t), wind has the components **U,V,W** on any arc (i)
 - given that an arc is considered as « **canyon street** », and that there is wind, then **turbulence** exists and the following turbulence coefficients are considered for each arc (i) during interval (t): (K_h) and (K_z)
 - the variables **E=D = W = 0** for the following reasons:
 - the surface of an arc is considered to be **laminated**, $D=W=0$
 - it is considered that **pollutant particles** move solely due to **wind intensity** and no other cause; thus $E=0$
 - the variables **Q** = pollutant emissions from traffic and **R** = speciation, are kept