

# Selection of Severe Transport Situations in the High-Pressure Gas Network of GTS

## Introduction of a Similarity Measure based on Optimal Flow Patterns

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- 1 Introduction
- 2 Process of checking scenarios
- 3 Current representation
- 4 Flow representation
- 5 Analysis and result
- 6 Conclusions

# Gasunie Transport Services

- Gas Transmission System Operator of the Netherlands
- Regulated part of the holding Gasunie
- Ensure safety of supply of natural gas
- No control on entry and exit capacities



# Gasunie Transport Services

## Complex network

- Entry and exit points (end points)
- Pipelines: length, diameter
- Loops
- Storages
- Compressor and blending stations

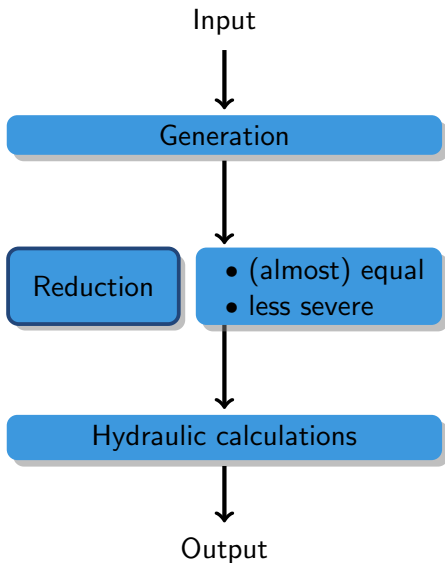


## Goal of the research

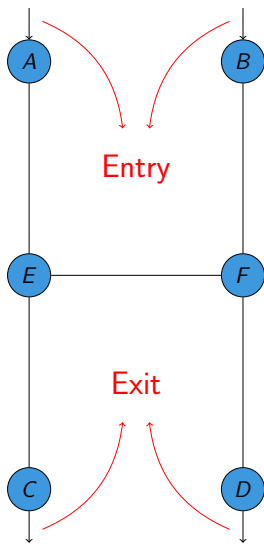
- Extremely many gas transportation situations possible
- Computations are time consuming
- Stress tests: severe and realistic transport scenarios
- Still a large set of scenarios

*Find a small representative set of stress tests*

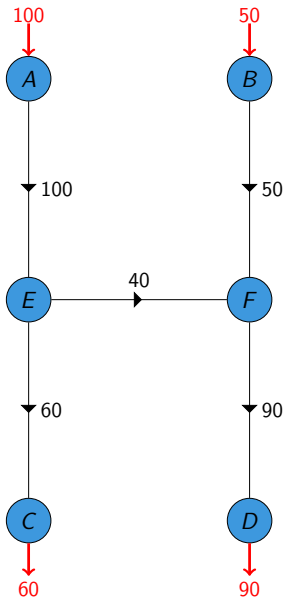
# Process of checking scenarios



## Example: H-network



## End point representation

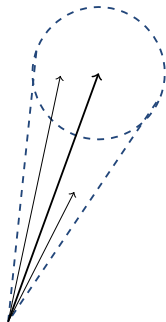


$$\begin{matrix} A \\ B \\ C \\ D \end{matrix} \begin{pmatrix} 100 \\ 50 \\ -60 \\ -90 \end{pmatrix}$$

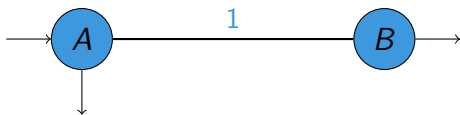


# End point representation

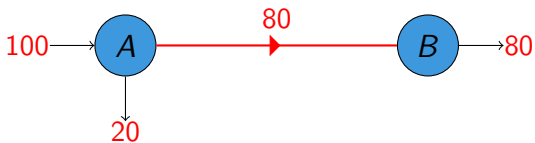
- Generation of stress tests
  - Taking all directions into account
  - Many scenarios
- Reduction
  - Equal
  - Similar
  - Less severe
- Find distance function



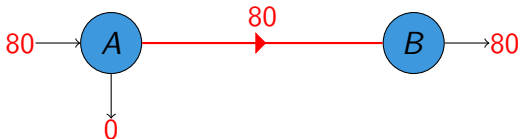
## Example: One Pipeline Network



## Example: One Pipeline Network



$$x = \begin{pmatrix} 100 \\ -20 \\ -80 \end{pmatrix}$$



$$y = \begin{pmatrix} 80 \\ 0 \\ -80 \end{pmatrix}$$

$x, y$ : same transport load

## Example: One Pipeline Network

$$x - y = (20, -20, 0)^T$$

$$\begin{aligned}\mathcal{L}_1(x - y) &= \sum_i |x_i - y_i| \\ &= 20 + 20 \\ &= 40 \neq 0\end{aligned}$$

$$\begin{aligned}\mathcal{L}_2(x - y) &= \sqrt{\sum_i (x_i - y_i)^2} \\ &= \sqrt{400 + 400} \\ &= 20\sqrt{2} \neq 0\end{aligned}$$

## Quadratic Form Distance

$$QFD_{\mathbf{A}}(x, y) = \sqrt{(x - y)^T \mathbf{A} (x - y)}$$

with  $a_{ij} = 1 - d_{ij}/d_{\max}$

$x, y$  : vector of capacities

$d_{ij}$  : transport distance from  $i$  to  $j$

$d_{\max} : \max_{ij} d_{ij}$

## Example: one Pipeline Network

$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad x - y = \begin{pmatrix} 20 \\ -20 \\ 0 \end{pmatrix}$$

$$QFD_{\mathbf{A}}(x, y) = \sqrt{(20 \quad -20 \quad 0) \begin{pmatrix} 1 & 1 & 0 \\ 1 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 20 \\ -20 \\ 0 \end{pmatrix}} \\ = 0$$

## Quadratic Form Distance

- Works for complex networks
- High-pressure network:  $400 \times 400$  matrix
- $QFD_{\mathbf{A}}$  is not a metric distance function

$$QFD_{\mathbf{A}}(x, y) = \sqrt{(x - y)^T \mathbf{A} (x - y)}$$

# Transport moment

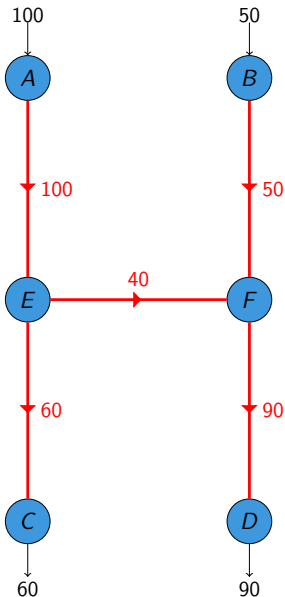
Transport moment ( $T$ ): linear measure of transport load

$T = \text{flow} \times \text{transport distance}$

$$Q \cdot \Delta p \sim Q \cdot L$$



# Flow representation



End point:

$$\begin{matrix} A \\ B \\ C \\ D \end{matrix} \begin{pmatrix} 100 \\ 50 \\ -60 \\ -90 \end{pmatrix}$$

Flow:

$$\begin{matrix} AE \\ BF \\ EC \\ FD \\ EF \end{matrix} \begin{pmatrix} 100 \\ 50 \\ 60 \\ 90 \\ 40 \end{pmatrix}$$

## Weighted norm

$$\mathcal{L}_1^L(x, y) = \sum_i L_i |x_i - y_i| \quad (\text{Weighted } \mathcal{L}_1\text{-norm})$$

$$\mathcal{L}_2^L(x, y) = \sqrt{\sum_i L_i^2 (x_i - y_i)^2} \quad (\text{Weighted } \mathcal{L}_2\text{-norm})$$

$x_i, y_i$  : flow through pipeline  $i$  of scenario  $x, y$

$L_i$  : length of pipeline  $i$

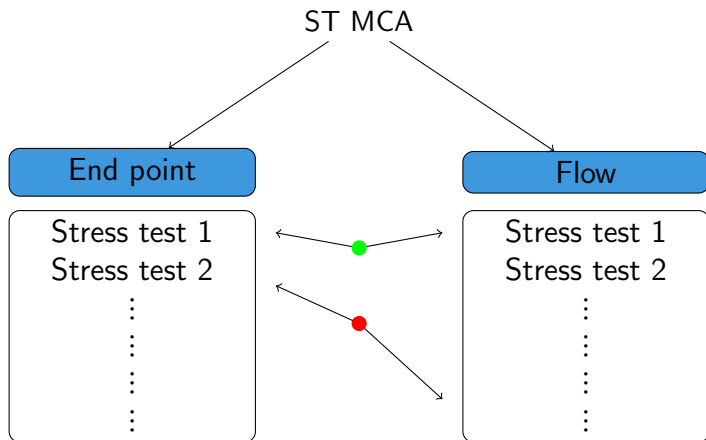
- $\mathcal{L}_1^L$ : close to transport moment

## Comparison of representations

<b>Representation</b>	<b>End point</b>	<b>Flow</b>
Pre-calculations	Less	More
Close to network physics	Less	More
Simple distance function	Less	More
Metric distance function	No	Yes
Adaptable distance function	Difficult	Easily

Possible adaption: weighted  $\mathcal{L}$ -norm:  $L \rightarrow L/D^5$

# Analysis of clustering



## Analysis

- Measure leads to clustering of stress tests
- Compare clustering of both representations

## Result

The more complex the network, the more difference in clustering in both representations

# Conclusions

- Representation of transport scenarios in physical terms is possible:

*flow representation*

- Simple distance function, with physical transport properties
- Clustering similar for small, different for complex networks
- From experiments unclear which clustering is the best

# Recommendations

- Hydraulic calculations for comparison (simple network)
- Analyse the effect of loops
- Add more network information
  - Analyse  $L/D^5$  instead of  $L$
  - Addition of pressure drop
- Direct generation of flow pattern

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## Distance functions for stress tests

$$QFD_{\mathbf{A}}(x, y) = \sqrt{(x - y)^T \mathbf{A} (x - y)} \quad (\text{Quadratic Form Distance})$$

with  $a_{ij} = 1 - d_{ij}/d_{\max}$

$$\mathcal{L}_1^L(x, y) = \sum_i L_i |x_i - y_i| \quad (\text{Weighted } \mathcal{L}_1\text{-norm})$$

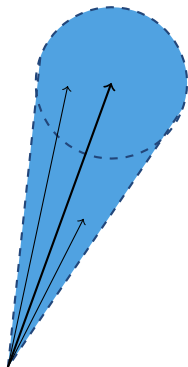
$$\mathcal{L}_2^L(x, y) = \sqrt{\sum_i L_i^2 (x_i - y_i)^2} \quad (\text{Weighted } \mathcal{L}_2\text{-norm})$$

## Pressure Drop Equation

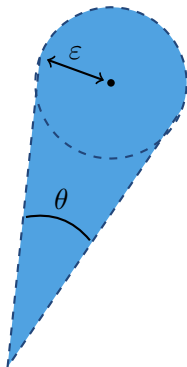
$$p_{in}^2 - p_{out}^2 = k \cdot \frac{L}{D^5} \cdot Q \cdot |Q|$$

- $p_{in}$ : pressure at begin of pipeline
- $p_{out}$ : pressure at end of pipeline
- $k$ : constant
- $L$ : pipeline length
- $D$ : pipeline diameter
- $Q$ : flow through pipeline

# Cone of Clustering



Clustered scenarios



Definition cone

# Transport moment

## End point representation

$$T(\tau) = \sum_j c(X_j)d(\tau, X_j) - \sum_i c(N_i)d(\tau, N_i)$$

## Flow representation

$$T = \sum_i L_i Q_i$$

## QFD in flow representation

$$\begin{aligned} QFD(x) &= \sqrt{x^T \begin{pmatrix} w_1^2 & & \\ & \ddots & \\ & & w_n^2 \end{pmatrix} x} \\ &= \sqrt{\sum_i (w_i x_i)^2} \\ &= \mathcal{L}_2^w(x) \end{aligned}$$