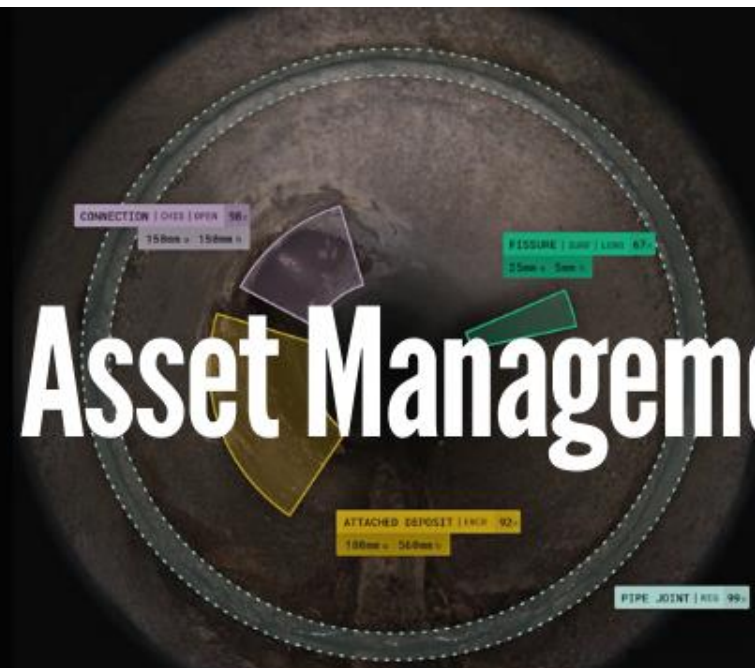


AI-empowered Asset Management



IWA SPECIALIST GROUP ON STRATEGIC ASSET MANAGEMENT



Rita Salgado Brito –
LNEC, Portugal
Chair



Oliver Nachevski – GIZ,
North Macedonia
Vice-chair



Bénédicte Rulleau –
INRAE, France
Secretary and Treasurer



Helena Alegre – LNEC,
Portugal
Member



Rita Amaral – Lis-
Water, Portugal
YWP Member



Cristiano Gonçalves
Nascimento Gouveia –
Caesb. Brazil



Eric Montes –
Maynilad Water,
Philippines



Nicolas Caradot – KWB,
Germany
YWP Member



Elvira Estruch Juan –
ITA, Spain
YWP Member



Boudewijn Neijens –
Copperleaf
Technologies, Canada
Member



Takayuki Sawai – NJS
Co. Ltd, Japan
Member



Sandra Cecilia
Muhirirwe – NWSC,
Uganda
YWP Member



Andres Torres –
Pontificia Universidad
Javeriana, Colombia
Member

Management Committee of IWA SAM SG

- Join over **1,300 water professionals:**
<https://iwa-connect.org/group/strategic-asset-management/timeline>
- The platform for collaboration and knowledge sharing on Asset Management
- Organising the 2022 LESAM conference in Bordeaux



**Alexander
Ringe**

**Berliner
Wasser-
betriebe**
Germany

**Aisha
Mamade**

Baseform
Portugal

**Pascale
Rouault**

KWB
Germany

**Youen
Pericault**

**Luleå
University of
Technology**
Sweden

**Srinath
Kumar**

SewerAI
USA

**Dulcy
Abraham**

**Purdue
University**
USA

**Nico
Caradot**

KWB
Germany

**Elvira
Estruch**

**Universitat
Politècnica de
València**
Spain

AGENDA

- **Introduction**

Nico Caradot | Elvira Estruch

- 1. Improving rehabilitation strategies by using simulation models**

Alexander Ringe

- 2. AI Tools for Asset Management Planning**

Aisha Mamade

- **Q&A Panel Discussion 1**

- 3. Uncertainties in sewer deterioration models: How much can we trust?**

Pascale Rouault

- 4. Data-enabled coordination of urban infrastructures rehabilitation**

Youen Pericault

- 5. AI-Enabled Sewer Defect Coding for Greater Speed and Consistency**

Srinath Kumar | Dulcy Abraham

- **Q&A Panel Discussion 2**

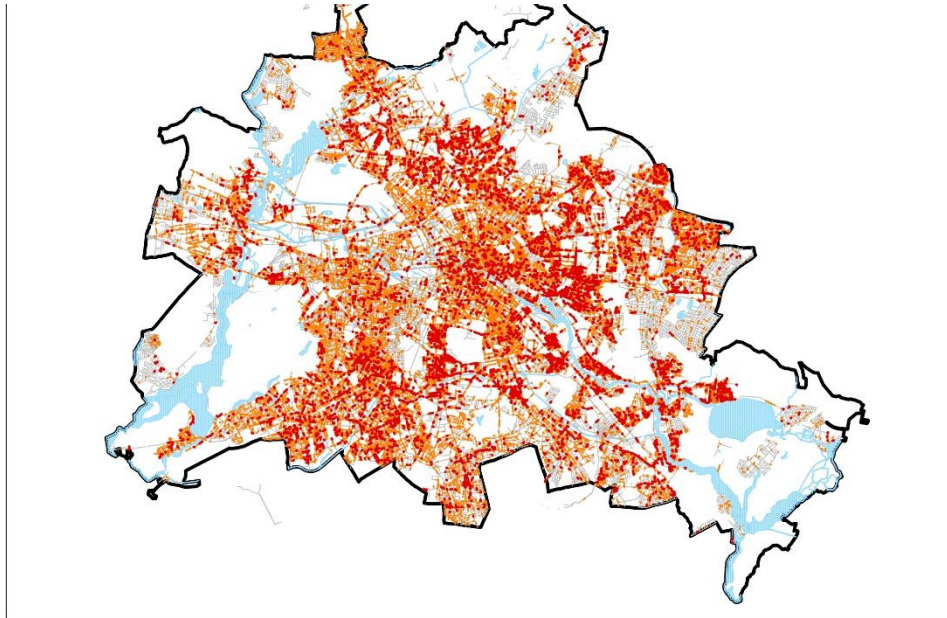
Improving rehabilitation strategies by using simulation models

ALEXANDER RINGE
BERLINER WASSERBETRIEBE,
GERMANY



CHALLENGE SEWER NETWORK BERLIN

Symbolic visualisation - no real rehab need in map!



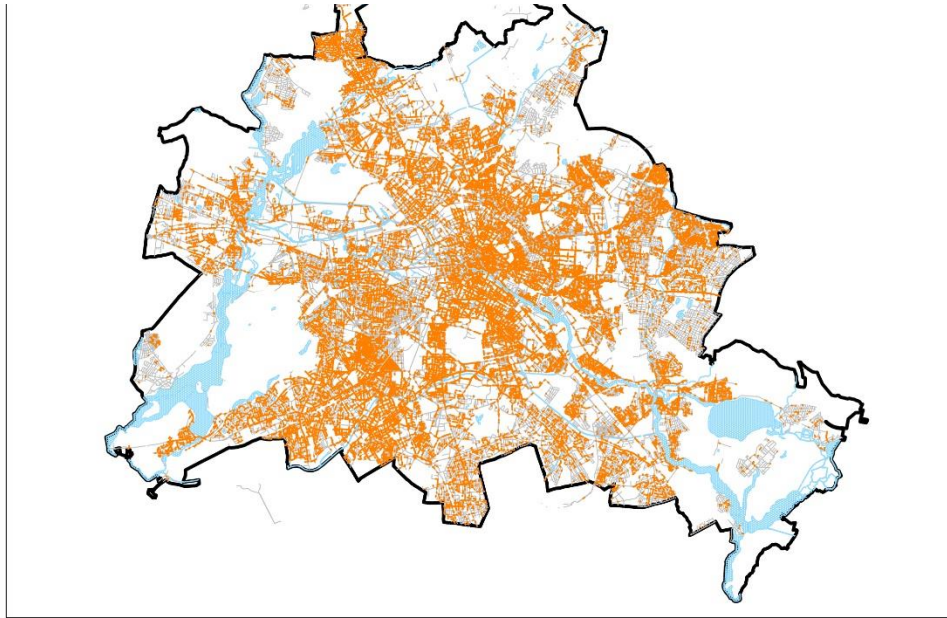
- ~ 10.000 km of sewers
- ~ 750 km/a of CCTV
- € for rehabilitation limited (sewage fee etc.)
- construction constraints in the city
- prevent further deterioration

map legend:

- sewer network Berlin
- symbolic sewer lines with short-term need for rehabilitation
- symbolic sewer lines with mid-term need for rehabilitation

GOAL OF INVESTMENT STRATEGY

Symbolic visualisation - no real rehab need in map!

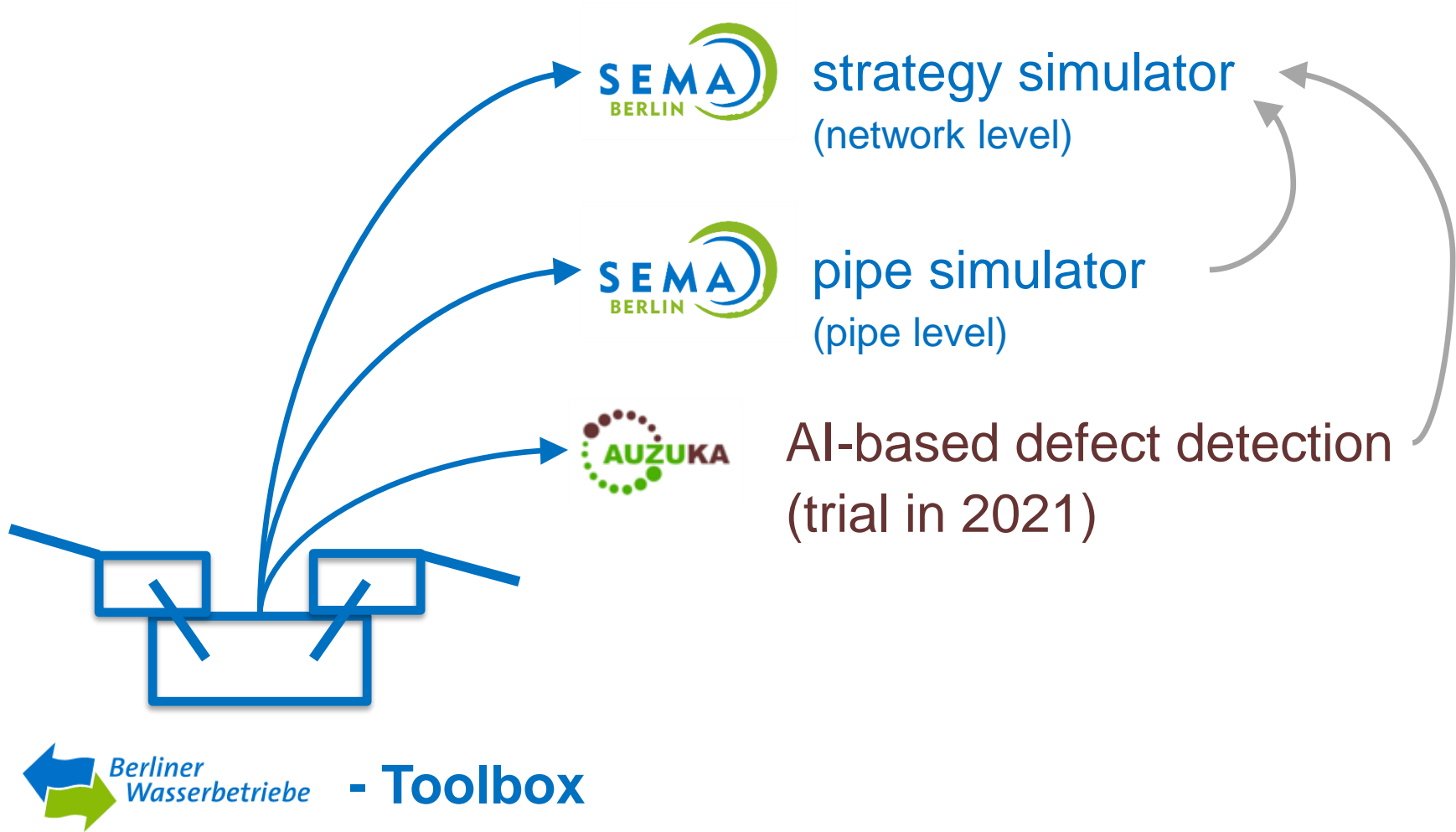


- reduce risk of collapse!
- optimise € for rehabilitation!
- **continuously improve network condition!**

map legend:

- sewer network Berlin
- symbolic sewer lines with short-term need for rehabilitation
- symbolic sewer lines with mid-term need for rehabilitation

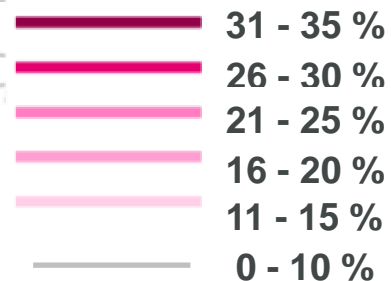
STRATEGY IMPLEMENTATION TOOLS





- Allocation of pipes with short-term rehabilitation need
- Supports inspection strategy (hotspot - search)

Probability of pipe for short-term rehab need:





- Trial in 2021
- Improvement of CCTV evaluation quality (input data of strategy simulator)

Cutout of AI- Viewer:

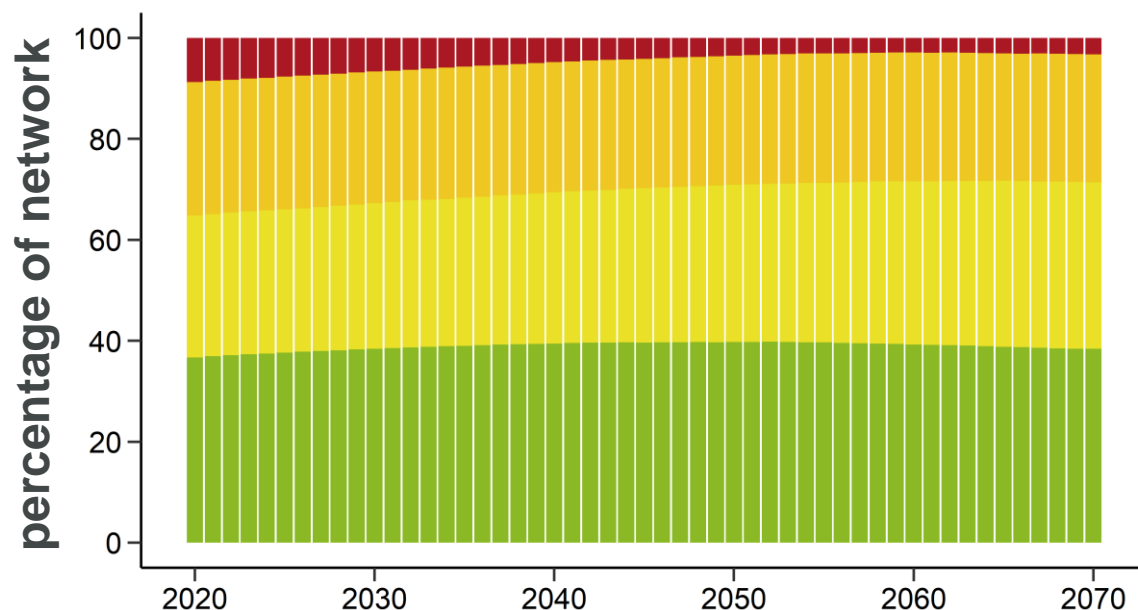


Automatic crack detection



Automatic corrosion detection

Sewer network development



- Simulation of cost effective rehabilitation strategy (renewal & lining & repair)
- Simulation of network development incl. €

Need of rehab:



short-term



mid-term



long-term



none

- Use of simulation models is essential for a cost effective rehabilitation strategy of the sewer network
- Quality of input data determines quality of simulation results - focus on quality of input data!
- The quality of sewer network development modelling also depends strongly on the extent to which it takes into account actual rehabilitation practice and success of inspection strategy!

Thank you very much for your attention!



Alexander Ringe
alexander.ringe@bwb.de



AI Tools for Asset Management Planning

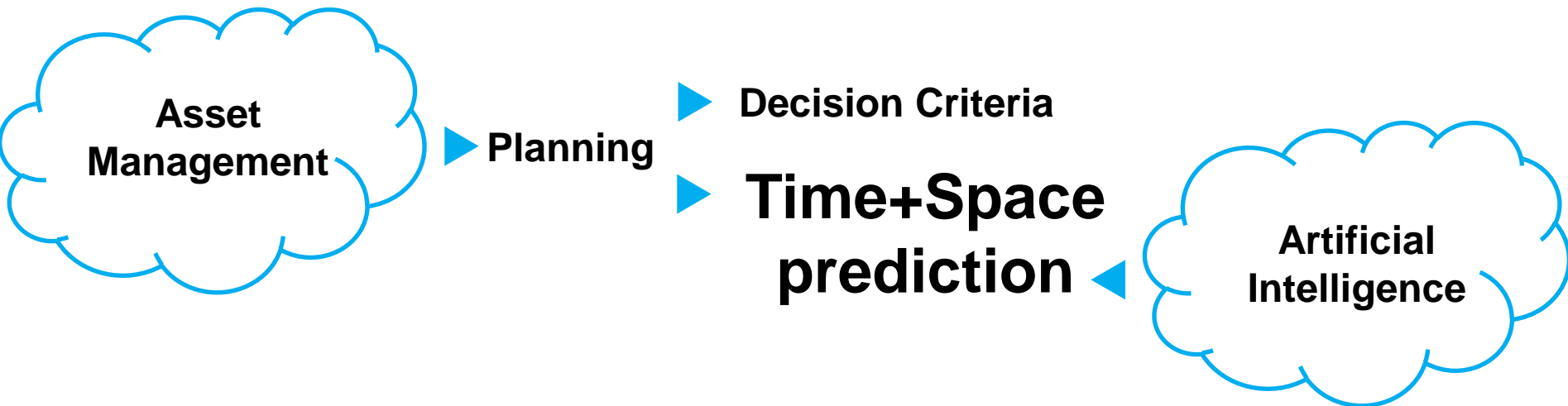
AISHA MAMADE
BASEFORM, PORTUGAL

baseform

inspiring change



SOME DEFINITIONS



BASEFORM PREDICT APPS

Home

baseform
Jane Smart

Recent

Library/systems/Water Supply/FA Lucity 2
18 APRIL 2021

Balance YR-01
14 APRIL 2021

Balance YR-16
14 APRIL 2021

YR-01
05 APRIL 2021

Library/systems/Water Supply/FA Lucity
23 MARCH 2021

Apps

City

People

Network

Zones 3D

Monitor

Maps

Meters

Events

Components

Spectrum

Diagnose

Network model

Indices

Minimum Energy

Energy Balance

Water Balance

Predict

Failure Prediction

Inspection Analysis

Component Importance

Risk of no service

Plan

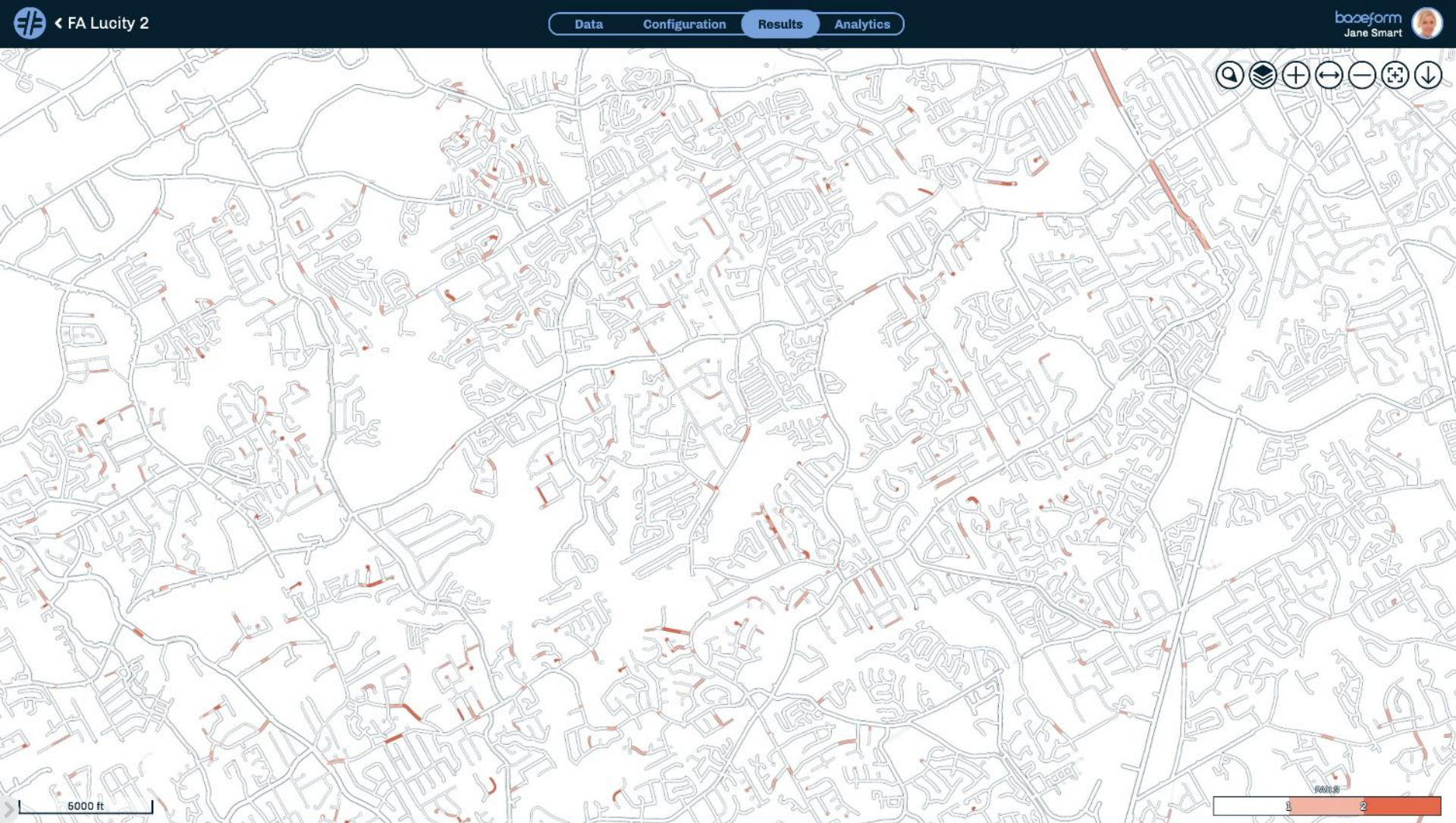
Compare & prioritize

Indicators

Financial Analysis

Infrastructure value index

OBSERVED PIPE BREAKS



PREDICTED BREAK RATE



PREDICTED BREAK RATE



PREDICTED BREAK RATE



HOW GOOD ARE THESE ESTIMATES?



HOW GOOD ARE THESE ESTIMATES?



CAN I CREATE A RENEWAL PLAN WITH THESE RESULTS?



< FA Lucity 2

Data

Configuration

Results

Analytics

baseform
Jane Smart



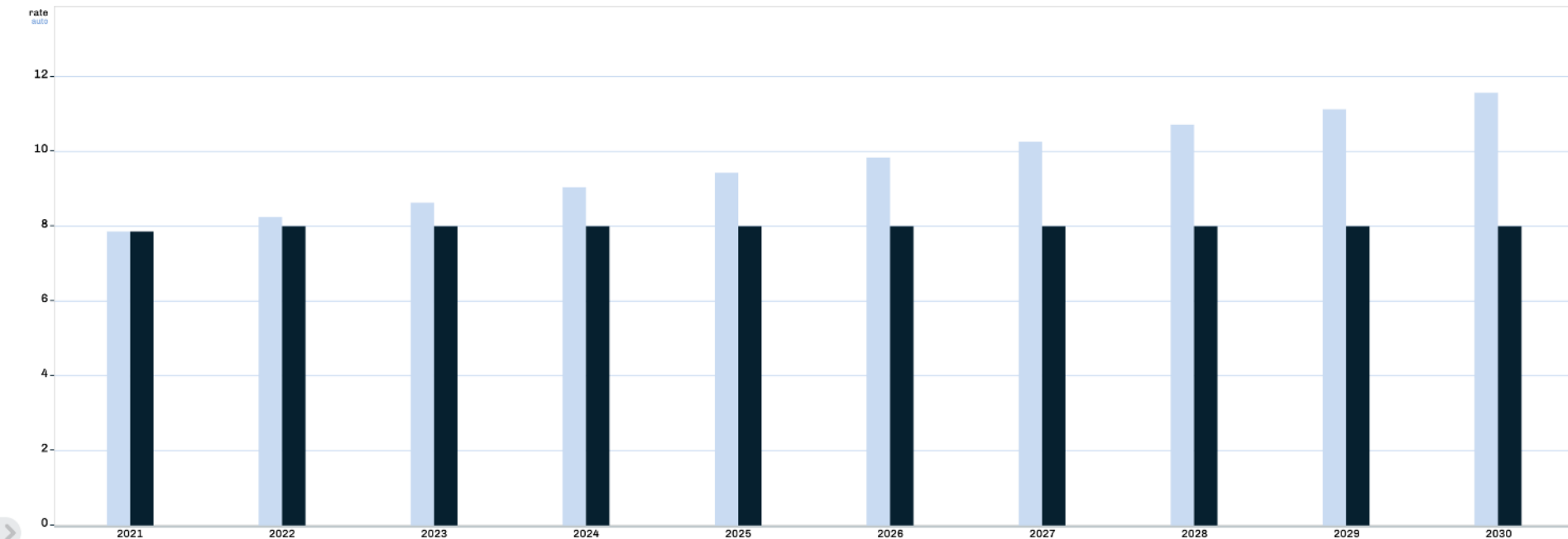
REPLACEMENT PLANNING

View Fail ra... ▾

INFO

Planning horizon (years)	10
Prioritization Criteria	Reduce failures, minimum replacement length ▾
Yearly Target	Maximum failure rate ▾ 8
Replace with	Set: DIP ▾
Group by	-- ▾

Replacement Planning



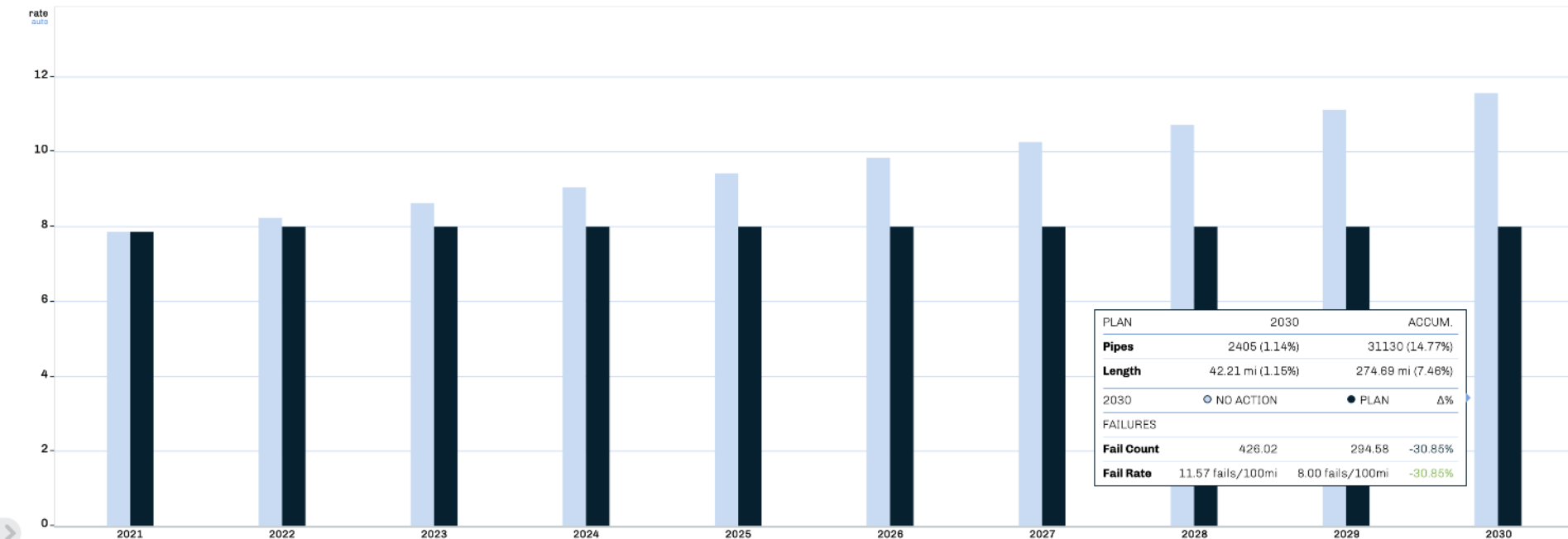
HOW MUCH BETTER IS THIS PLAN?

REPLACEMENT PLANNING View Fail ra...

INFO

Planning horizon (years)	10
Prioritization Criteria	Reduce failures, minimum replacement length
Yearly Target	Maximum failure rate 8
Replace with	Set: DIP
Group by	--

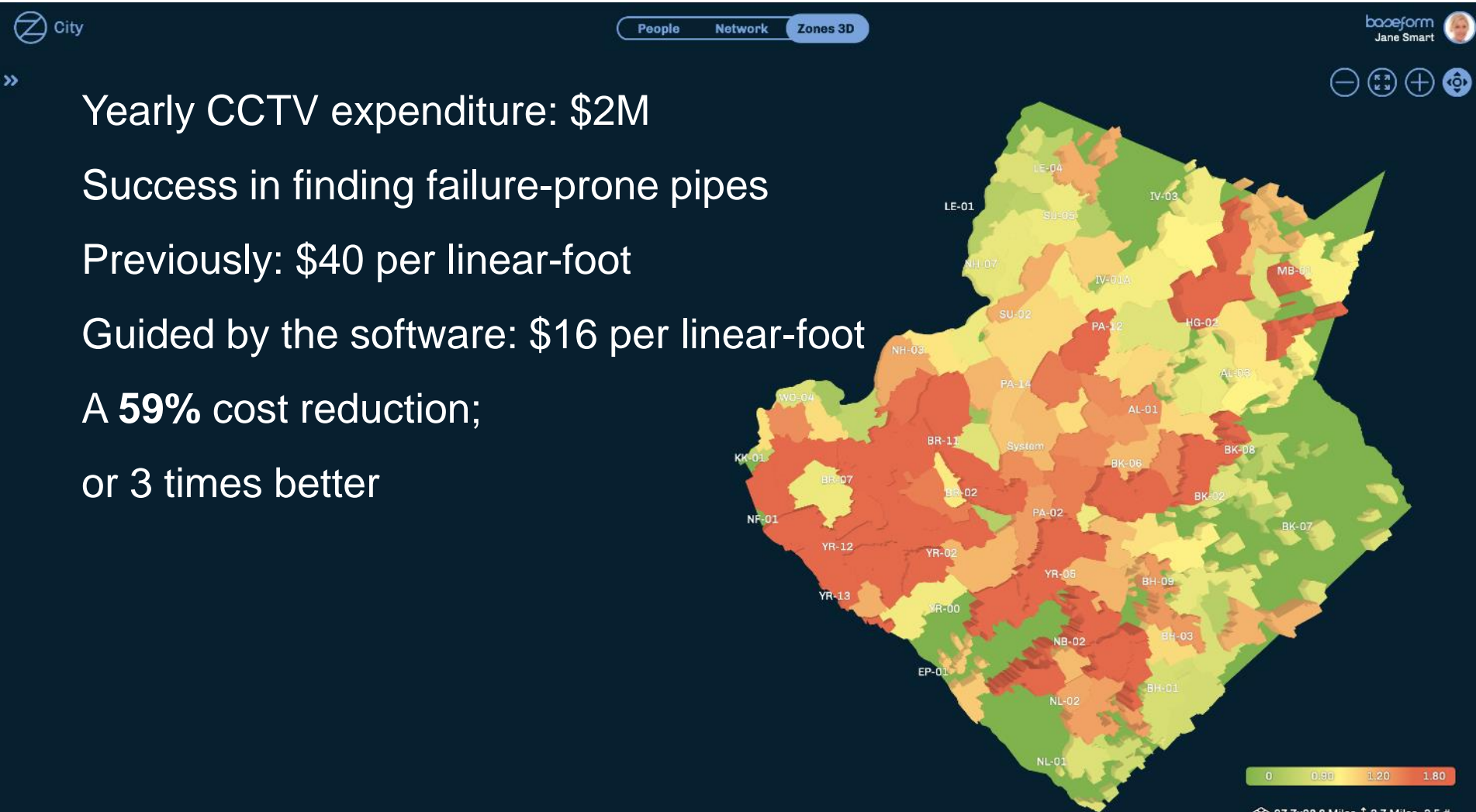
Replacement Planning



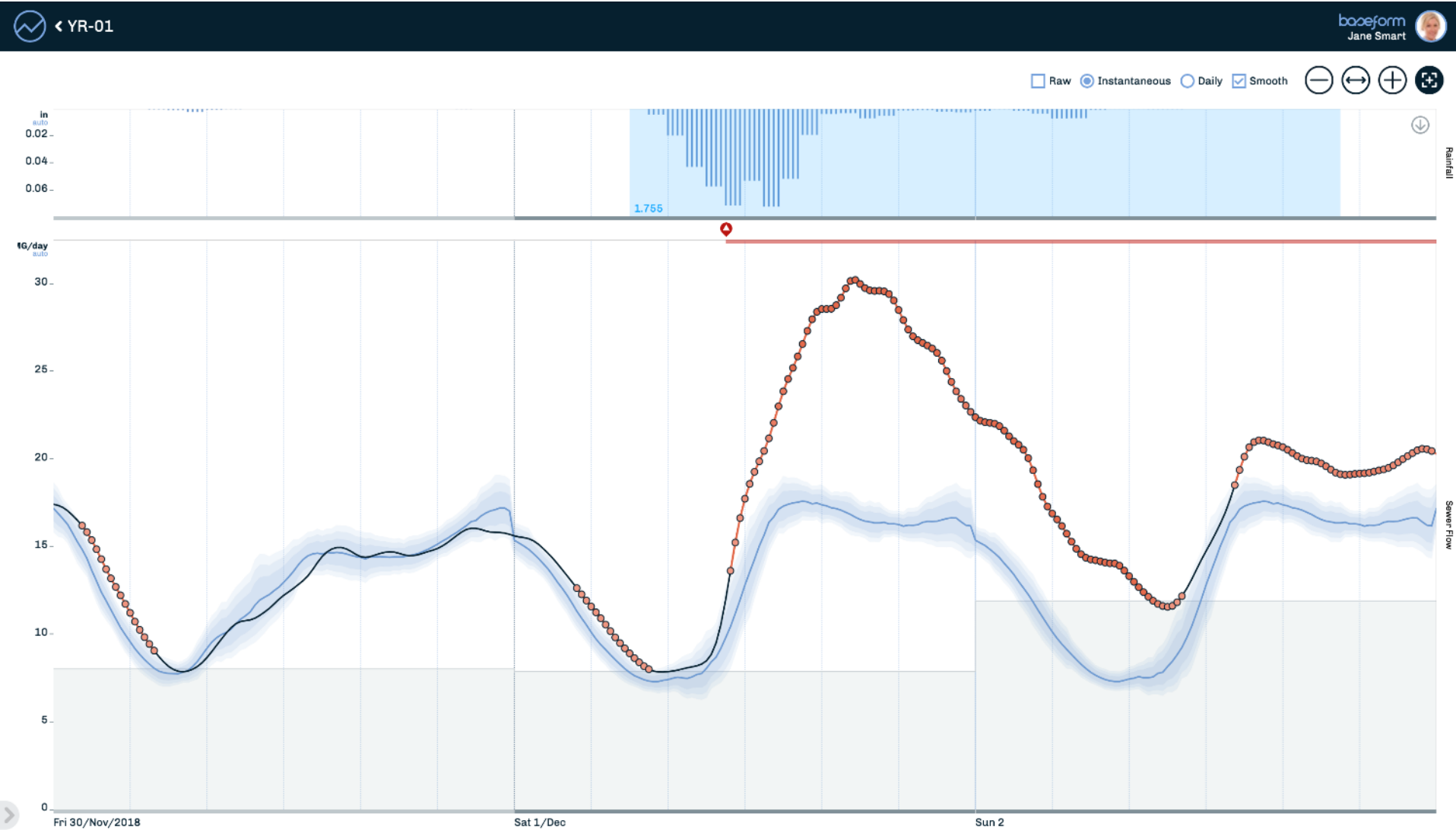
HOW MUCH BETTER IS THIS PLAN?

PLAN	2030	ACCUM.	
Pipes	2405 (1.14%)	31130 (14.77%)	
Length	42.21 mi (1.15%)	274.69 mi (7.46%)	
2030	○ NO ACTION	● PLAN	Δ%
FAILURES			
Fail Count	426.02	294.58	-30.85%
Fail Rate	11.57 fails/100mi	8.00 fails/100mi	-30.85%

PREDICTING STRUCTURAL CONDITION



PREDICTING HUMAN CONSUMPTION



THANKS!



**the fundamental purpose
of the water service has not changed**

**what is changing
is our ability to make
informed decisions.**

aisha.mamade@baseform.com

Q&A Discussion 1

ALEXANDER RINGE, AISHA MAMADE
(MODERATED BY NICO CARADOT & ELVIRA ESTRUCH)

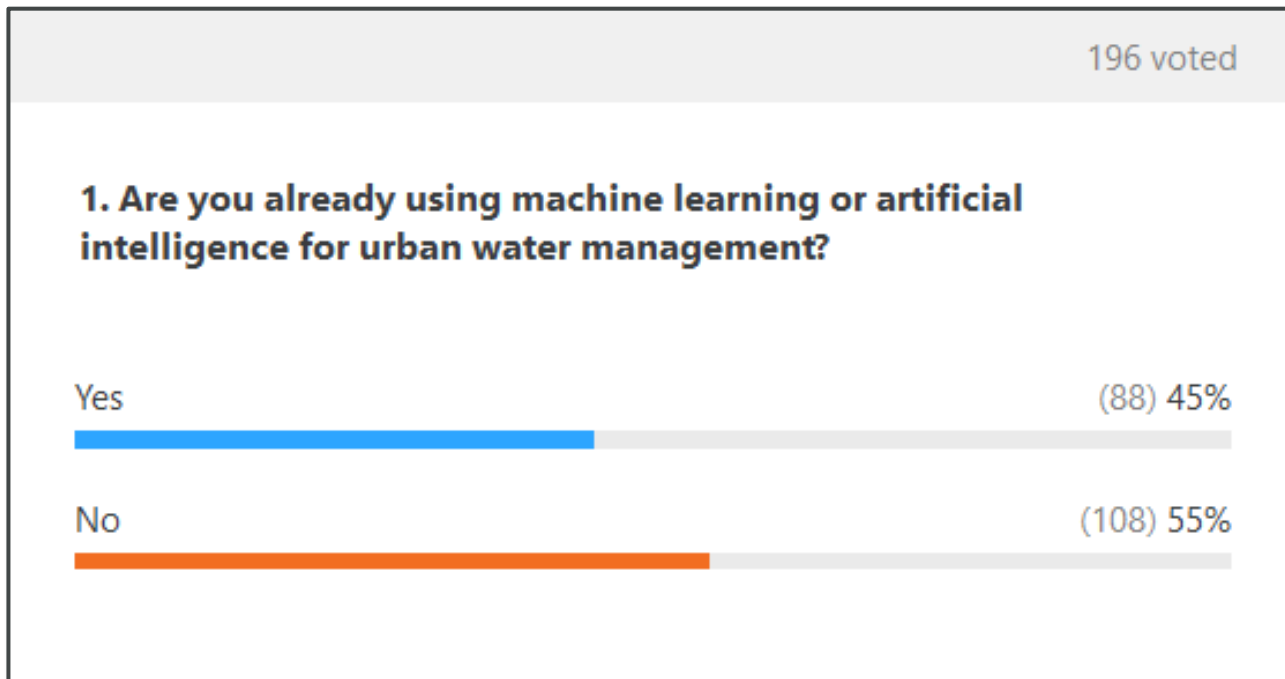
POLL 1: USE OF ARTIFICIAL INTELLIGENCE

Single choice

1. Are you already using machine learning or artificial intelligence for urban water management?

- Yes
- No

POLL 1: USE OF ARTIFICIAL INTELLIGENCE



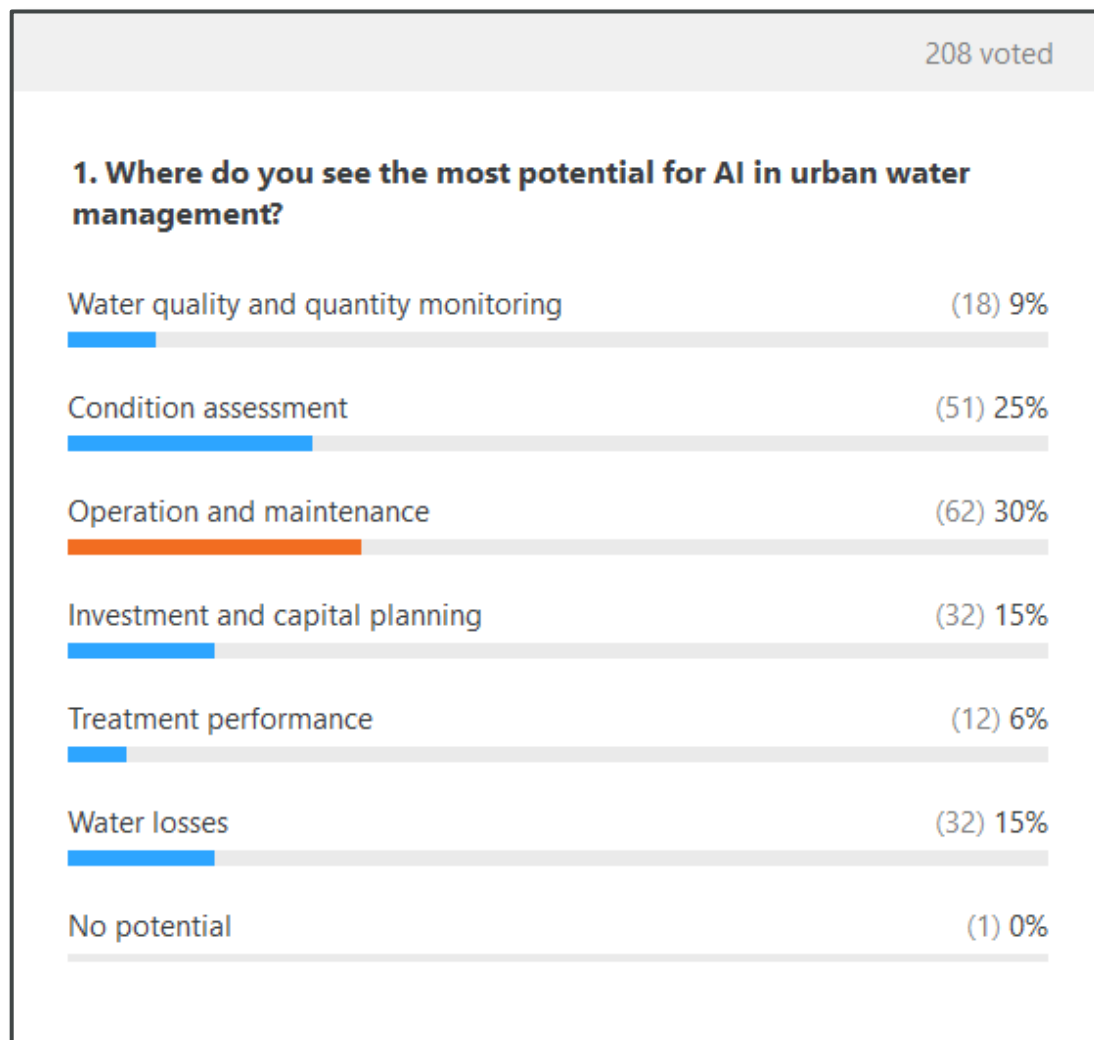
POLL 2: POTENTIAL FOR AI

Single choice

1. Where do you see the most potential for AI in urban water management?

- Water quality and quantity monitoring
- Condition assessment
- Operation and maintenance
- Investment and capital planning
- Treatment performance
- Water losses
- No potential

POLL 2: POTENTIAL FOR AI



Uncertainties in sewer deterioration models: How much can we trust?

PASCALE ROUAULT
KWB, GERMANY

KOMPETENZZENTRUM
WasserBerlin



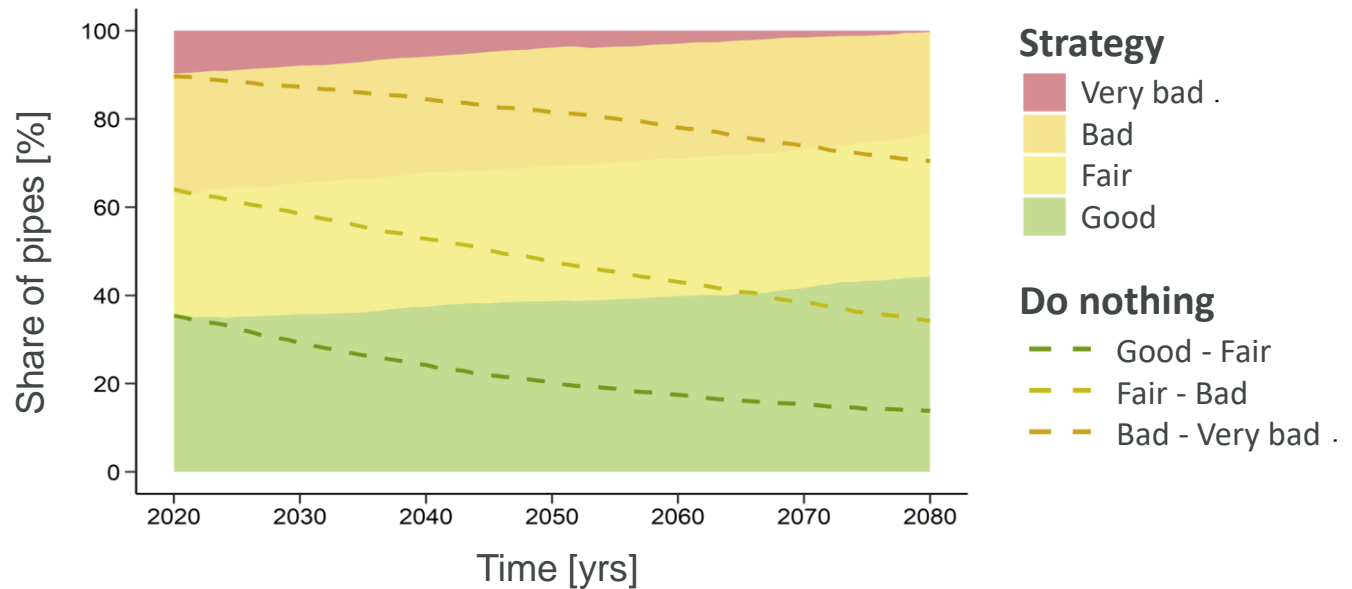
MODEL TOOLS ARE POWERFUL

See Presentation Alexander Ringe (BWB) for the city of Berlin

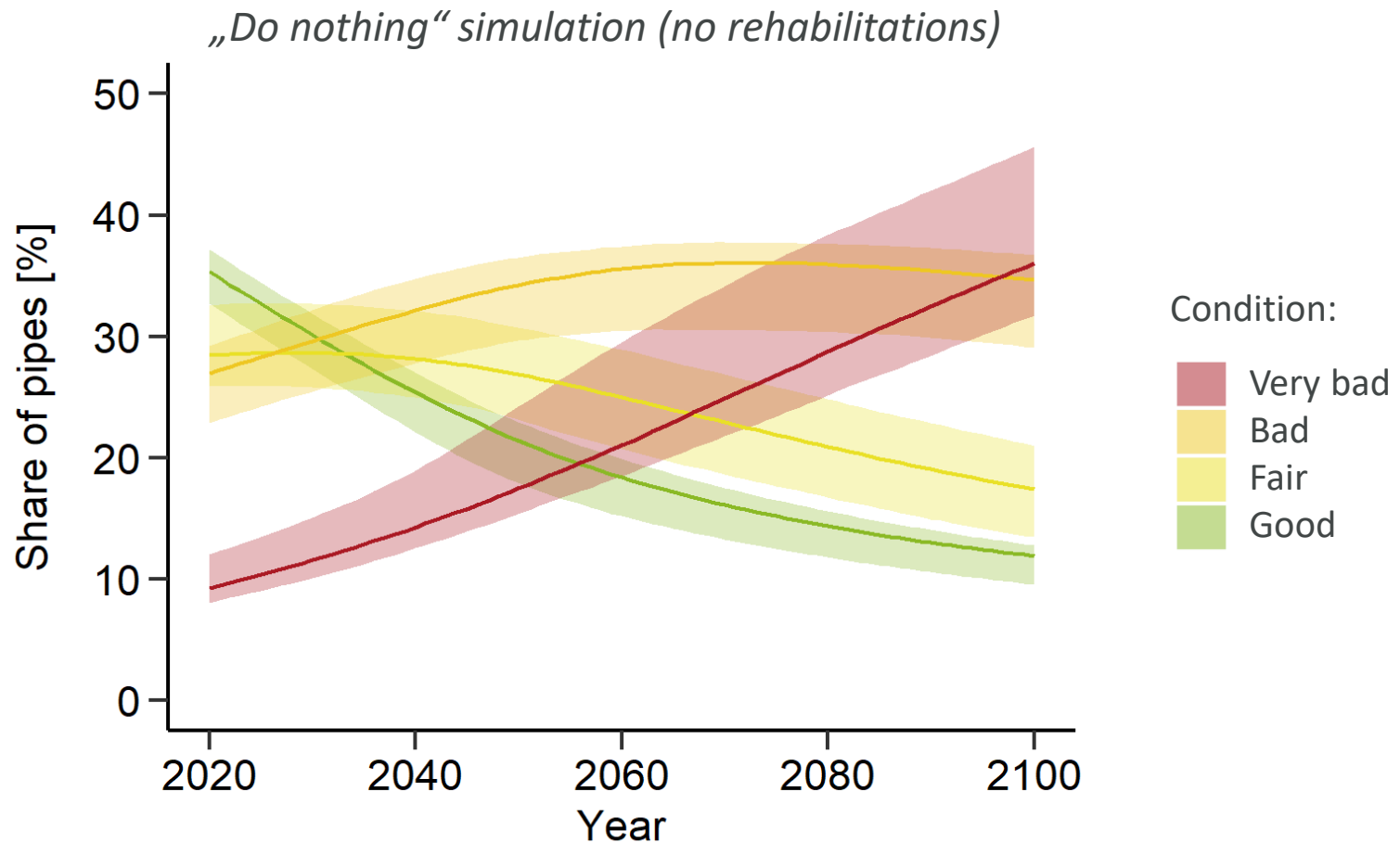
Strategic network simulator

Long-term predictions on the sewer **network** condition under consideration of different investment strategies

Statistical
model
„GompitZ“



AS EVERY MODEL THEY ARE UNCERTAIN BUT HOW UNCERTAIN?

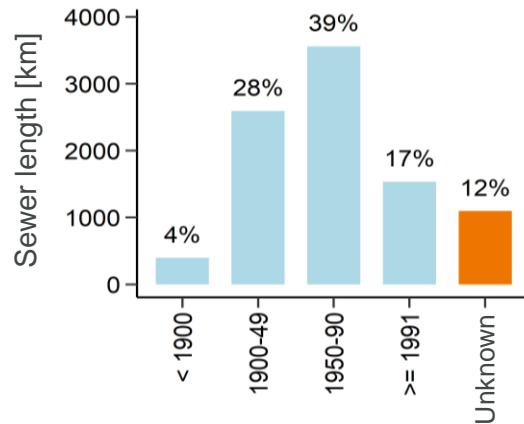


SOURCES OF UNCERTAINTIES

Group	Uncertainty source
Network data (input data uncertainties)	<ul style="list-style-type: none">• Missing data (e.g. construction year)
CCTV data (calibration uncertainties)	<ul style="list-style-type: none">• Subjectivity of condition assessment• Survival bias in condition data• Data availability (data quantity and representativity)
Deterioration + rehabilitation models (model structure uncertainties)	<ul style="list-style-type: none">• Statistical representation of condition• Representation of repairs and linings

1. MISSING DATA

Construction years in Berlin:



Two approaches tested for filling gaps:

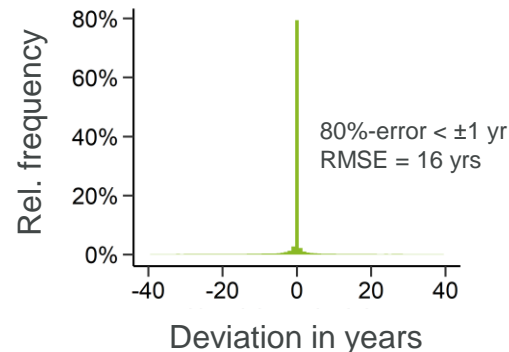
- **Nearest neighbour** model based on construction year of neighbouring pipes
- **Random Forest** (input: sewerage type, shape, diameter, soil type, buildings, ...)

Findings:

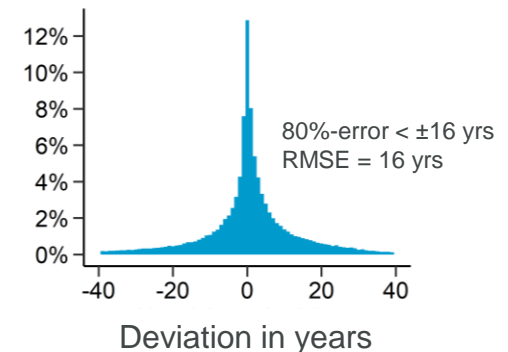
- All data gaps can be closed
- Accurate prediction for majority of pipes
- Symmetric error distribution
- Marginal effect on long-term sewer condition prediction

Uncertainty range: \updownarrow 0.5%

Nearest neighbour



Random forest



2. SUBJECTIVE ASSESSMENT

Uncertainty matrix: Probabilities of being in real condition i when inspected in condition j

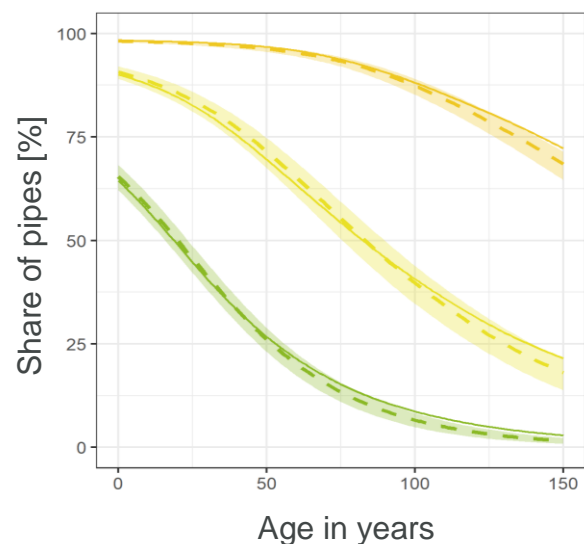
		real			
		1	2	3	4
Inspected	1	82,2	7,7	4,3	3,5
	2	13,5	72,2	16,4	6,4
	3	3,6	18,2	72,9	18,0
	4	0,7	1,9	6,4	72,1
Σ		100	100	100	100

Uncertainties originate from (i) **subjective assessment** and (ii) **delayed documentation of rehabilitations** → **70% for Berlin (assumption!)**



Propagation of uncertainties from **subjective assessment** in survival curves (clay pipes)

(solid line = default survival curve)



quantified on basis of double inspections of the same pipe within max. 5 yrs without reported rehab ($n = 4685$)

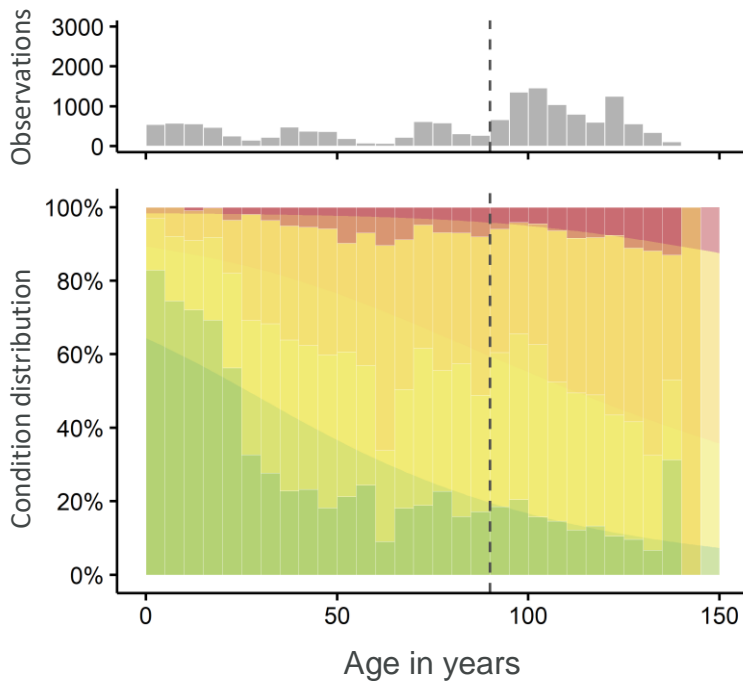
- Relevant uncertainties and systematic error in default survival curves
- Delayed documentation of rehabilitations is a major factor

↑ Uncertainty
↓ range: **3.8%**

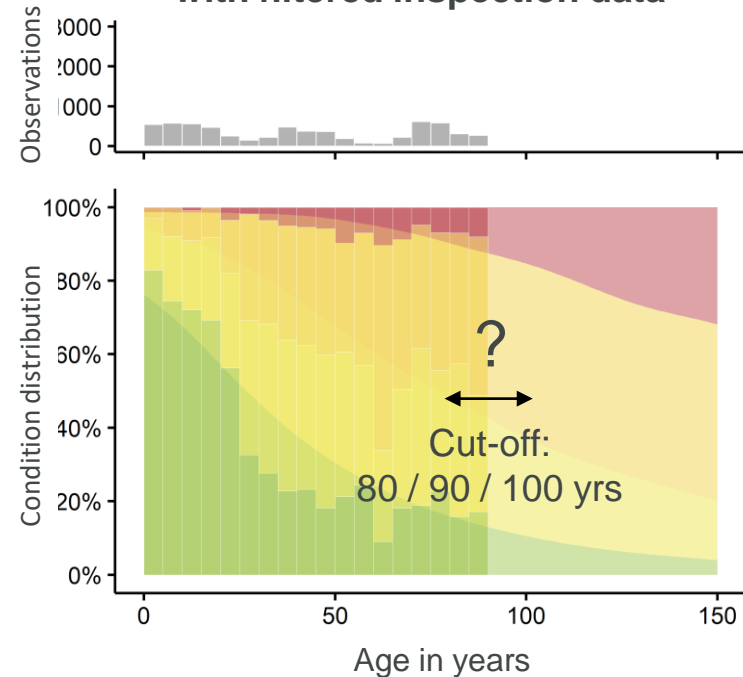
3. SURVIVAL BIAS

Example: clay pipes (combined sewage)

Inspection data with survival bias



Calibration of survival curves
with filtered inspection data



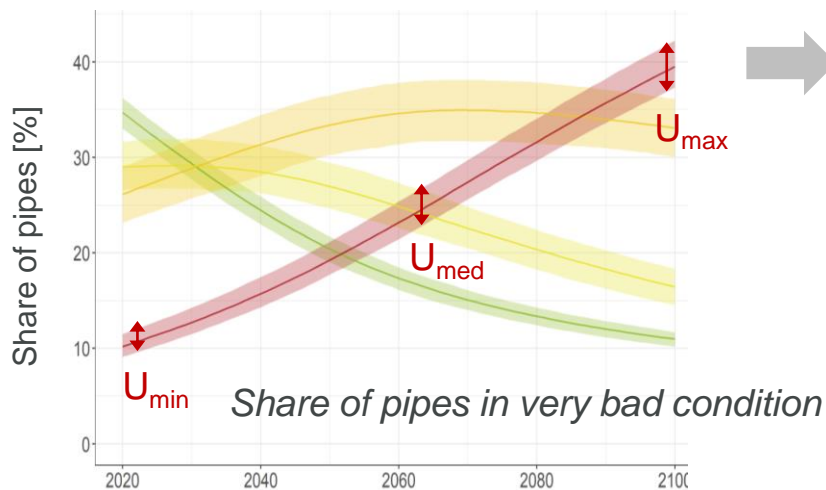
- Inspection data are biased: too few inspections of old pipes in very bad condition → too optimistic survival curves
- However, survival bias and ageing patterns are hard to distinguish

↑ Uncertainty
↓ range: 3.9%

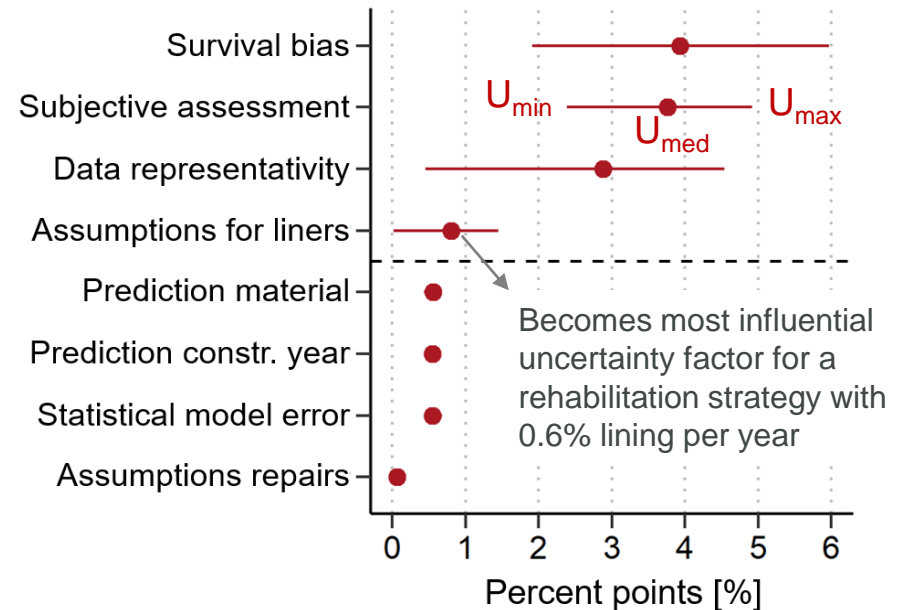
RANKING OF UNCERTAINTY FACTORS

Uncertainty propagation for do-nothing simulation

(Example: subjective assessment)



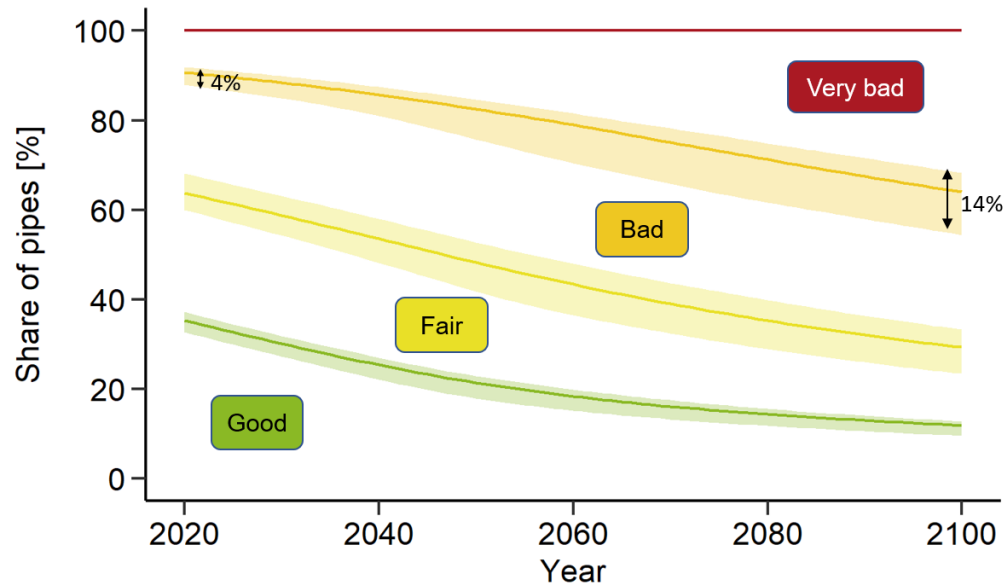
Ranking of uncertainty factors



- Four relevant uncertainty sources have been identified
- Assumptions for liners can become the most influential factors for rehab strategies with relevant lining shares, e.g. > 0.5% per year

OVERALL UNCERTAINTIES

Overall uncertainties (“Do-nothing” strategy)



- Predictions for the condition of the sewer network are subject to relevant uncertainties
- Overall uncertainties regarding the proportion of pipes in very bad condition for a 80- yrs forecast are 14 percentage points for a do-nothing-simulation
- For a selected rehabilitation strategy with a rehab rate of 1% per year * uncertainties increase to approx. 20 percentage points

* 20% renewal, 60% lining, 20% repair

- The results can help utilities in...
 - Developing more reliable rehabilitation strategies
 - Reducing the identified uncertainties in the future

- Potential for reduction of uncertainties is seen in...
 - Support of condition assessment with AI-based image recognition
 - An improved and prompt documentation of rehabilitations and the condition of rehabilitated sewers
 - Experiments and long-term observations on the deterioration of liners, plastic pipes, reinforced concrete and other marginal material

Contact: pascale.rouault@kompetenz-wasser.de

Data-enabled coordination of urban infrastructures rehabilitation

YOUEN PERICAULT
LULEÅ UNIVERSITY OF TECHNOLOGY
& LULEÅ MUNICIPALITY, SWEDEN



LULEÅ KOMMUN



COORDINATED REHABILITATION

- The fact of rehabilitating *adjacent* infrastructure assets (i.e. pipes, pavement) earlier or later than the end of their *functional lives* in order to perform the civil works in a single project.
- If done right, major opportunity to meet rehabilitation needs of multiple infrastructures at a lower economical, social and environmental cost.



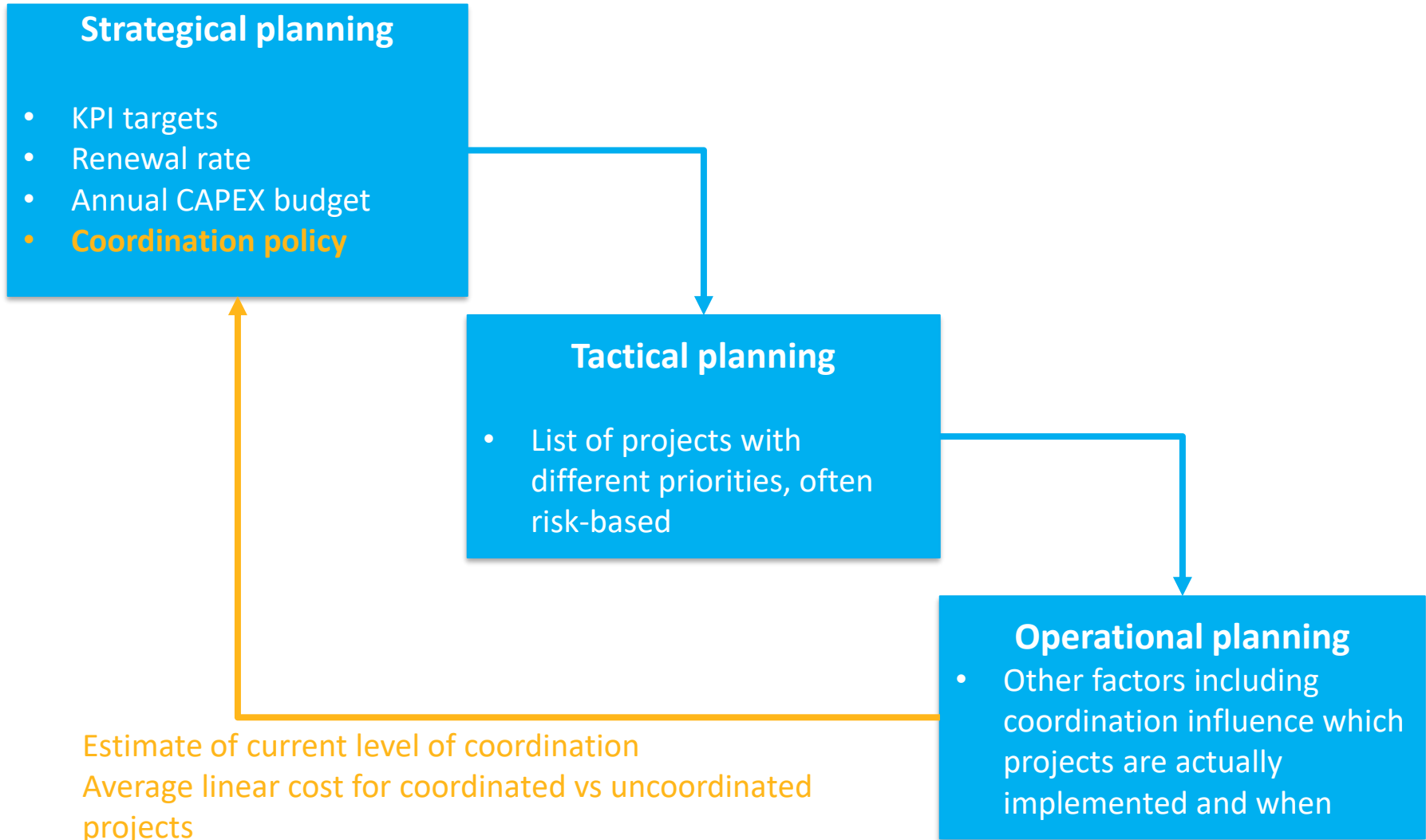
Probably lower linear costs
per rehabilitated asset

Lower risk of damaging
adjacent infrastructures

Under-utilized functional life
for assets replaced earlier

Lower performance and higher
risks for assets replaced late

COORDINATED REHABILITATION IN THE IAM PROCESS



INTERESTING QUESTIONS FOR A WATER UTILITY

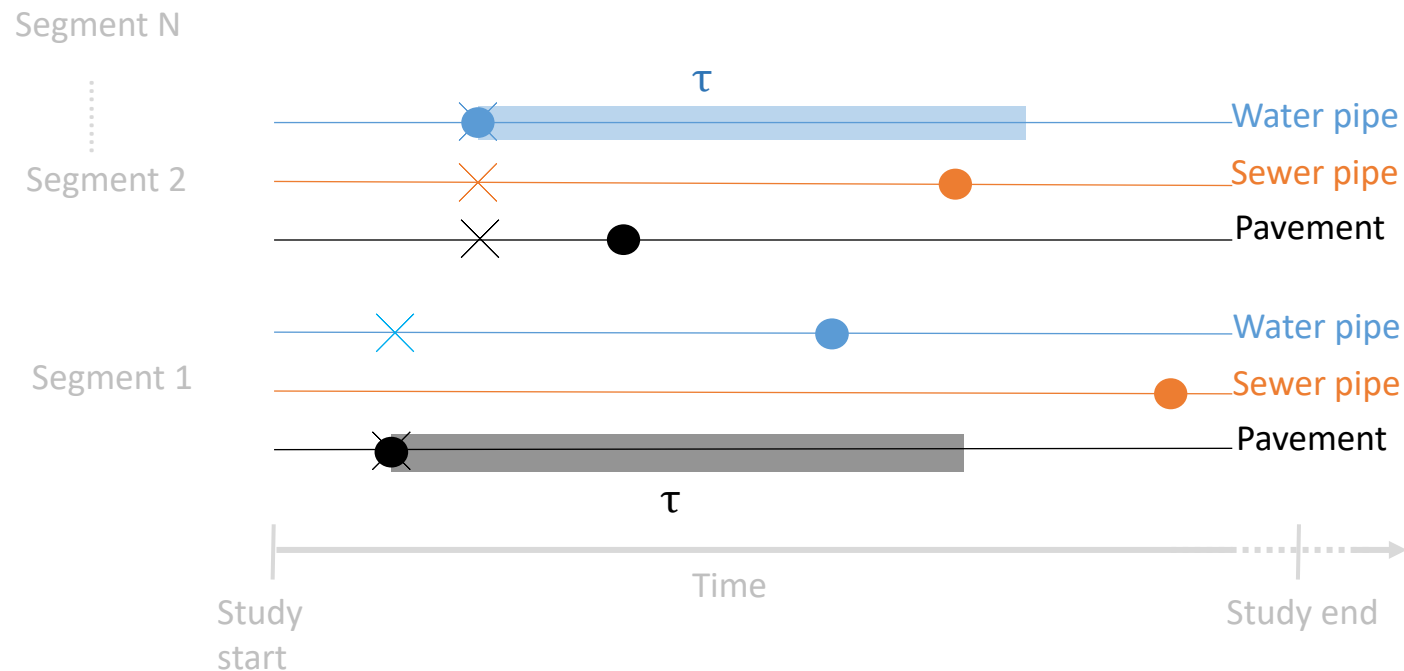
- How much coordinated rehabilitation do I perform today?
Should I perform more? Less?
- Could I make a better use of coordination by steering its use from the strategical level?
- What are the long term consequences of coordination on my costs and level of risk/performances?

There is today a lack of method and modelling tools to address these questions at the strategical decision level.

Develop a better understanding of coordination at the strategical level allow for a more pro-active use of it to reach given goals.

QUANTIFYING COORDINATION

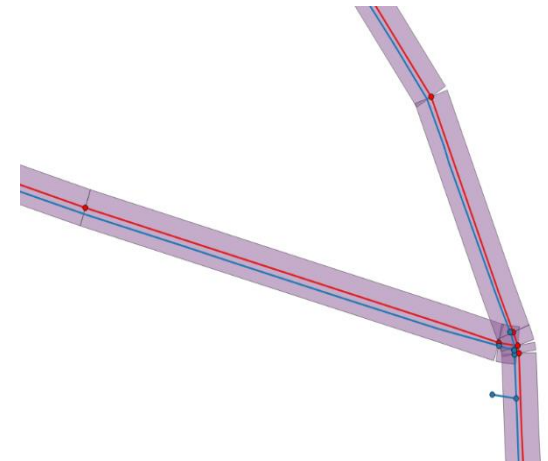
- Research at LTU: concept of coordination window
- $\tau=0 \rightarrow$ opportunistic coordination; $\tau=\infty \rightarrow$ systematic coordination
- Concept can be used to define coordination policies or to analyse previous coordination projects



MODELLING COORDINATION

- MURM prototype: compute long term rehab rate and costs for a given coordination policy (tau value)
- Main input data: linear costs of different rehab actions, cohort survival functions, **integrated asset inventory**:

	Cohort water	Cohort sewer	Inst. Year water	Inst. year sewer	Length (m)
Segment 1	Grey cast iron	Clay	1925	1925	45
Segment 2	Ductile cast iron	Concrete	1975	1990	32
.....
Segment N	PE	PVC	2001	2001	52



- Functional lifetimes sampled according to survival functions, inter-infra correlation can be chosen.
- Simulation is repeated and average cost values computed (monte-carlo approach)

EXAMPLE OF RESULTS: RESIDENTIAL STREETS OF LULEÅ

Figure showing cumulated capital cost for the period 2020-2120
as a function of coordination window size (τ)

Contact panelist if you wish to receive a copy of the figure:
youen.pericault@ltu.se

- 3500 segments,
176km
- Search for optimal
levels
- Multi-criteria
decision problem
- Total cost for
rehab of water,
sewer and road
network.

EXAMPLE OF RESULTS: RESIDENTIAL STREETS OF LULEÅ

Figure showing the evolution of yearly capital costs, global warming potential and renewal rates over the period 2020-2120 for a tau value of 0, 20 and 100 years.

Contact panelist if you wish to receive a copy of the figure:
youen.pericault@ltu.se

- Visualisation of yearly economical and environmental costs
- Renewal rates to support each utility in securing adequate future budget to follow the agreed tau value.
- Good base to start discussing costs sharing principles.

THANK YOU FOR YOUR ATTENTION!

AI-Enabled Sewer Defect Coding for Greater Speed and Consistency

DULCY M. ABRAHAM
PURDUE UNIVERSITY, USA

PURDUE
UNIVERSITY

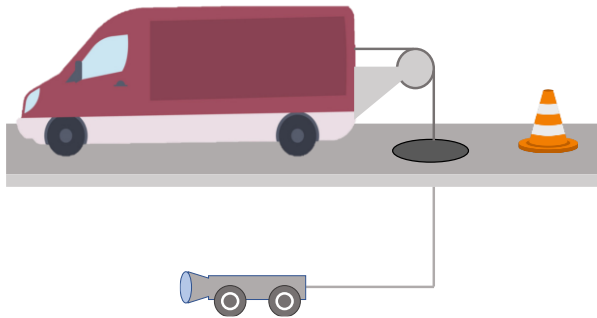
SRINATH SHIV KUMAR
SEWERAI, USA



SEWER INSPECTION PROCESS

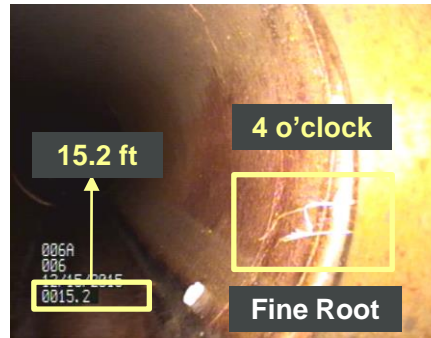
Water Research Foundation (WRF) # 4902

CCTV Inspections



Costly and Slow

Manually Identify Defects



Prone to Human Error

Inspection Report



Photo	Distance	Description	Time
	10.1	Root	00:00:00
	10.2	Root	00:00:00
	10.3	Root	00:00:00
	10.4	Root	00:00:00
	10.5	Root	00:00:00

Inconsistent Reports

How can AI Address these Issues?

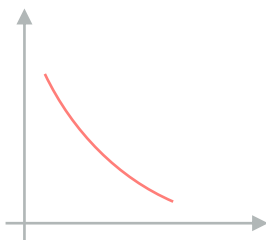
TERMINOLOGY

Artificial Intelligence

Deep Learning

Numerical

*Predicting Sewer
Deterioration Rates*



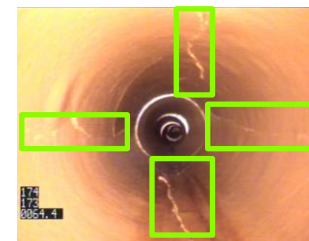
Audio

*Acoustic Leak
Detection*



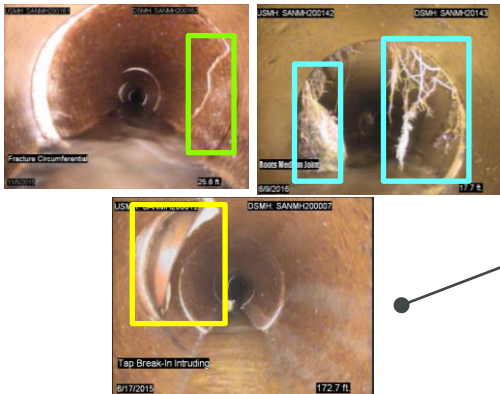
Vision

*Automated Sewer
Defect Detection*



SEWER DEFECT DETECTION

Training

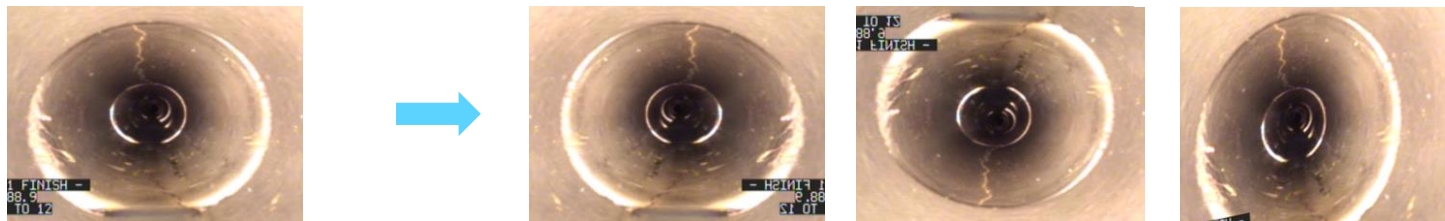


Deep Neural
Network

Inference



12,000 original images each for training



500,000 images after augmentation

FINDINGS OF WRF4902 STUDY



Better Recall than Humans

AI misses < 10% of defects, humans miss 25% of defects (Dirksen et al. 2013)

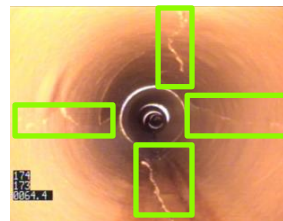
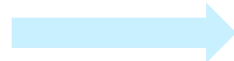
Worse Precision than Humans

AI creates 50% False Positives (humans < 5%)

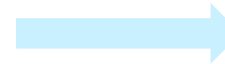
Human experts are needed to review AI's outputs



AI Defect Detection



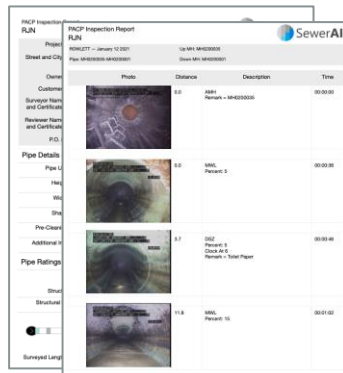
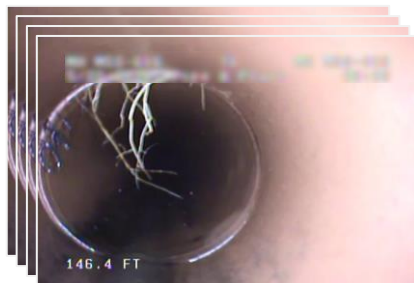
Human Review



- ***Add Images Into Training Set***
- ***Customer Deliverable***

SEWERAI PIONEER™ PLATFORM

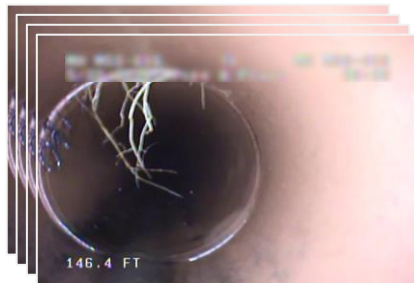
www.sewerai.com/pioneer



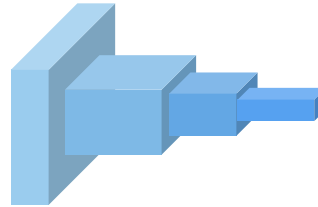
SEWERAI PIONEER™ PLATFORM

www.sewerai.com/pioneer

Under the Hood



*Deep Neural Networks
Trained on 500,000
images*

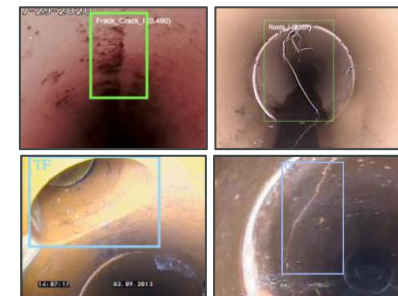


*Post-Processing
Algorithms*

**Proprietary
Algorithms**



Defects Identified by AI



*Humans Review AI's
Outputs*

- Code Defects
- Correct AI Errors
- Re-Train AI



Photo	Distance	Description	Time
	0.0	CRACK Manhole 1 - 10' (0.000000)	00:00:00
	0.0	ROOT Manhole 1 - 10' (0.000000)	00:00:00
	0.7	BLOCK Manhole 1 - 10' (0.000000)	00:00:00
	11.8	MANHOLE Manhole 1 - 10' (0.000000)	00:00:00

LARGE UTILITY IN CALIFORNIA: EAST BAY MUNICIPAL UTILITIES DISTRICT

I&I Investigation → 65 miles of Digital Side Scan data in 18 days

Video Draw Mode 360 Video X

Crack/Fracture (0.35)

0:00 0:05

0:05.885 2.90 ft 1x

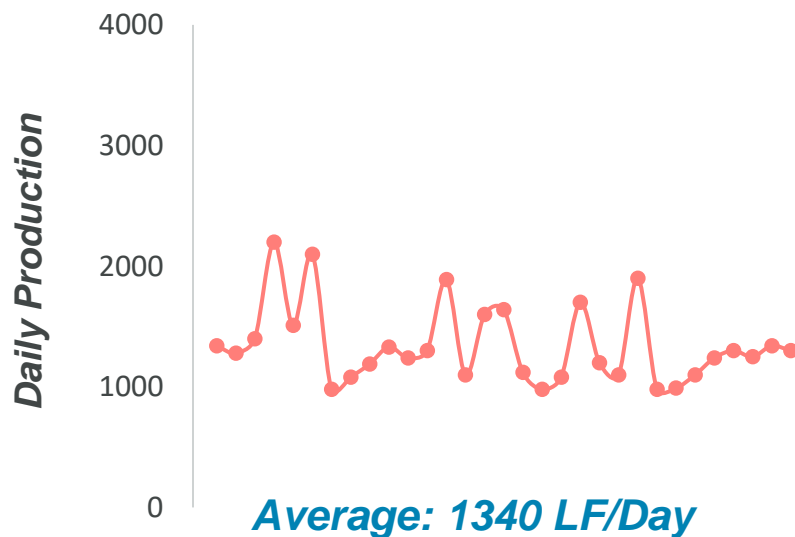
Time	Distance	Code	Continuous	Dim. 1	Dim. 2	%	Joint	Clock From	Clock To	Remarks
0:00	0	AMH	X							Added by SewerAI 1
0:00	0	MWL	X			1 X				Added by SewerAI 1
0:06	3.27	FM	X				Y...	11 X	1 X	
0:07	3.76	FM	X				Y...	12 X	1 X	
0:27	13.27	FM	X				Y...	7 X	1 X	
1:09	34.06	FM	X				Y...	9 X	1 X	
1:28	43.6	FM	X				Y...	11 X	1 X	
1:34	46.28	FM	X				Y...	11 X	1 X	
1:51	54.99	FM	X				Y...	8 X	1 X	
1:54	56.08	FM	X				Y...	9 X	1 X	
1:54	56.08	FM	X				Y...	1 X	1 X	
1:57	57.5	BSV	X				Y...	3 X	1 X	
1:57	57.5	BVV	X				Y...	8 X	1 X	
1:58	57.99	HSV	X				Y...	4 X	1 X	

Full Pipe

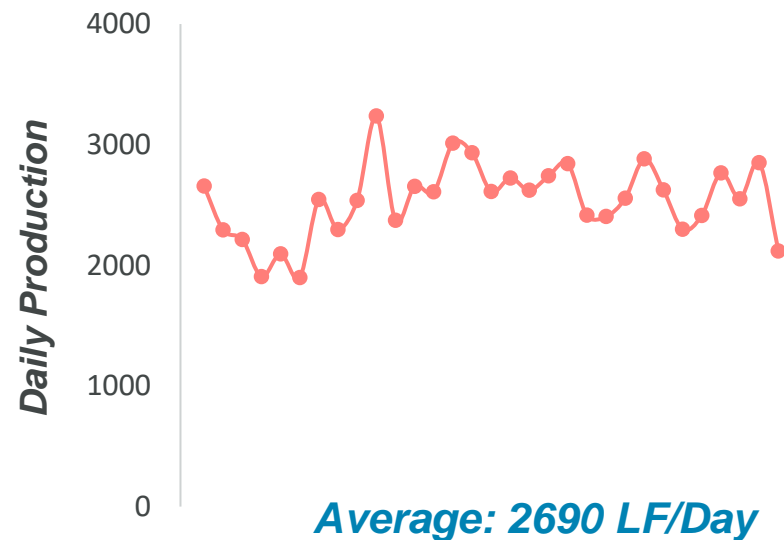
0:00.000 0:10.277 0:20.555 0:30.832 0:41.110

CONTRACTOR IN US WEST COAST PIPE AND PLANT SOLUTIONS INC.

Traditional Workflow



AI-Enabled Workflow



LESSONS LEARNED

AI can be more accurate and consistent than humans at identifying defects

Important to have human in-the-loop for reviewing AI outputs

AI-Enabled workflow → Significant improvement in productivity, accuracy, and consistency

Drone Inspections



Jetter Cameras



www.sewerai.com/pioneer

Q&A Discussion 2

**PASCALE ROUAULT, YOUEN PERICAULT,
DULCY ABRAHAM, SRINATH KUMAR**
(MODERATED BY NICO CARADOT & ELVIRA ESTRUCH)

EURO-SAM

SEWER ASSET MANAGEMENT WORKSHOP

- Held **online** on **16-17 June 2021**
- Free of charge
- 20 speakers
- Latest R&D outcomes on sewer asset management
- **Register now at:**
 - <https://iwasamsg.wordpress.com/>
 - <https://udam.home.blog/>



Join our network of water professionals!



IWA brings professionals from many disciplines together to accelerate the science, innovation and practice that can make a difference in addressing water challenges.

Use code **WEB21RECRUIT**

for a **20% discount off** new membership.

Join before 31 December 2021 at:

www.iwa-connect.org

inspiring change

Register at: <https://digital.worldwatercongress.org/>