

Smart Water Technologies: An Overview of Real World Applications

Thursday July 12 2018 1:00 - 3:00 PM ET



Today's Moderator

Elkin Hernandez





Today's Speakers

- Joshua Cantone & Jack Chan
 - MWRDGC's Outcome Driven Approach to Stormwater Planning
- Prateek Joshi
 - Operational Analytics for Water Treatment Plants
- Reese Johnson
 - Smart Sewers: Using Technology to Improve Wet Weather Operations



Our Next Speakers



Joshua Cantone, Ph.D.





Jack Chan, Ph.D., P.E.





MWRDGC's Outcome Driven Approach to Stormwater Planning

Joshua Cantone, Ph.D. Jack T.P. Chan, Ph.D., P.E.





An industry word jumble

- Artificial Intelligence
- Smart Networks
- Intelligent Systems
- Black Box
- Decision Support System
- Big Data
- Data Analytics
- Predictive Analytics
- Optimization
- · Smart Water

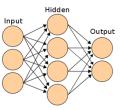


Demystifying Al...

- Artificial intelligence is perhaps the best overarching term - optimization is a part of that
- Two types of AI narrow (weak) and strong
 - <u>Narrow</u>: non sentient machine intelligence used for a narrow task
 - <u>Strong</u>: sentient machine intelligence with consciousness and mind







"STRONG AI"

"NARROW AI"



Real world examples

 Artificial Neural Networks





Fuzzy Logic





• Evolutionary Algorithms

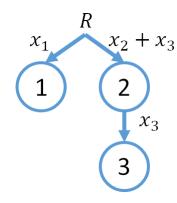






In the water industry...

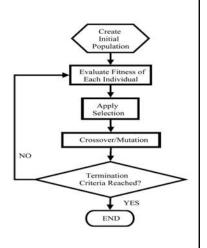
- Partial Enumeration
 - Best engineering judgement
- Linear, Non-linear, or Dynamic Programming
 - Linear equation to be maximized or minimized
 - Linear constraints
 - <u>Examples:</u> Water resources allocation



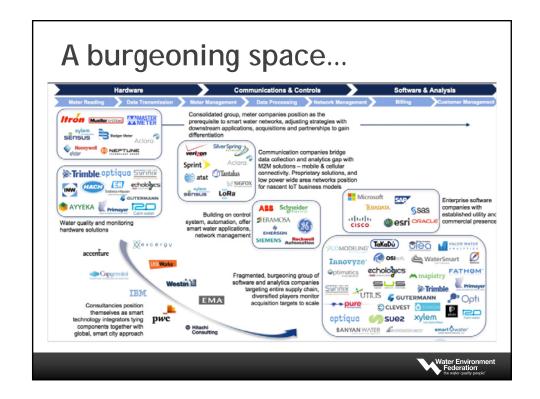


In the water industry...

- Evolutionary Algorithms
 - Based on nature, population based approach
 - <u>Examples:</u> Off-line planning for water distribution systems and wastewater collection systems
- Artificial Neural Networks
 - Learn from big data in order to predict or detect events
 - <u>Examples:</u> pipe breaks, flow prediction, etc.

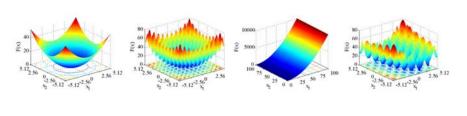




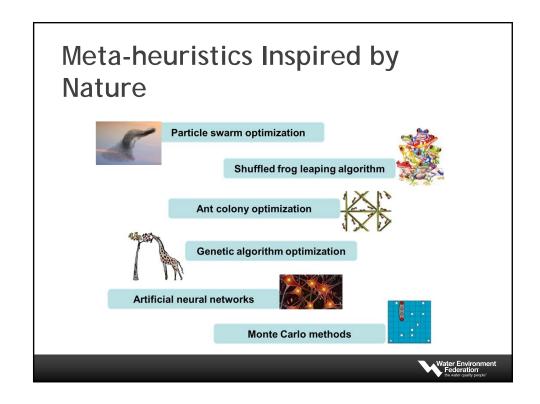


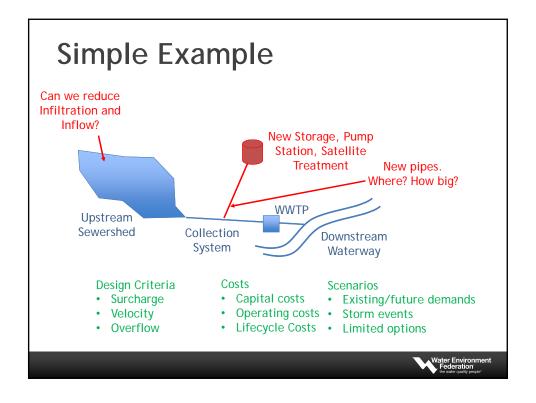
AI (Optimization) for Smart Planning

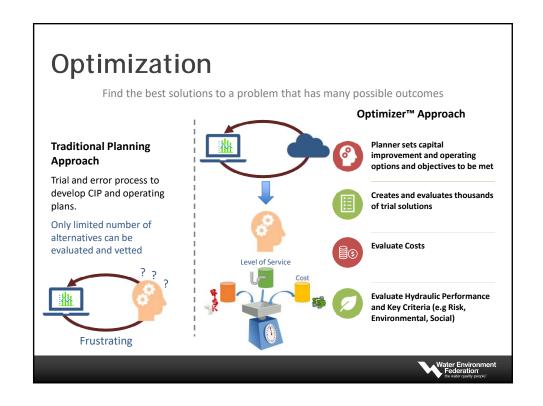
- Optimization is...
 - applying an analytic process to find the best solution to a problem that has many possible solutions
 - transparent, unbiased and adaptive approach
- Optimization is <u>not</u>...
 - black box











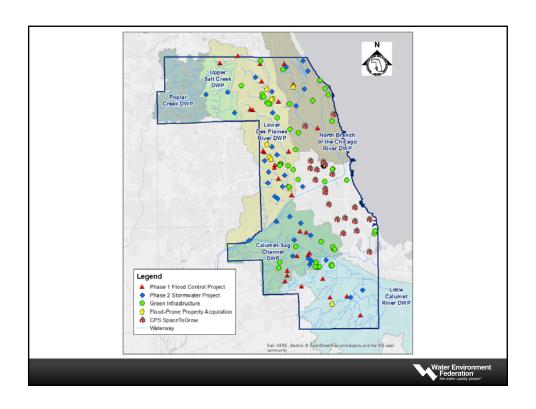
MWRDGC's Outcome-Driven and Adaptive Approach to Stormwater Planning



MWRDGC Background

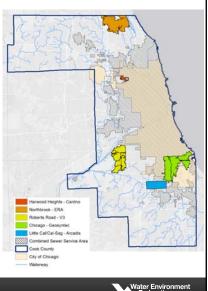
- ➤ 2004 the authority for general supervision of stormwater management in Cook County was conveyed to the District by the Illinois State legislature.
- ➤ 2011 Detail Watershed Plans (DWPs) completed for the 6 major watersheds of Cook County Cal-Sag Channel, Little Calumet River, Lower Des Plaines, North Branch of the Chicago River, Poplar Creek, and Upper Salt Creek.
- ▶ Phase I projects were identified from the DWPs to address overbank flooding "riverine flooding"
- ➤ 2014 the District's authority was amended to allow for flood-prone property acquisition and to plan, implement, finance, and operate local stormwater management projects.
- Phase II projects involves working with local communities and agencies to address local drainage problems.





Stormwater Masterplan Pilots

- ➤ Pilot study areas identified by four Councils of Government and the City of Chicago
- ➤ Study areas comprised of both separate or combined sewer areas
- ➤ Goal was to identify solutions to flooding of structures experienced in storms up to and including a 100-year event



Approach of the Studies

- Analysis of existing overland flooding and basement backup issues, including detailed (H&H) modeling of flooding issues and alternative solutions
- Minimize basement flooding and surface flooding by:
 - Balancing gray and green infrastructure
 - Implementing backflow prevention
 - Outcome driven
- Sought input from local municipalities, other stakeholders, and general public through questionnaires, public workshops, and other outreach tools to get full understanding of flooding impacts, and to identify preferences for green, gray, and/or private property solutions
- Public outreach effectiveness was also measured to evaluate the change in public attitude and willingness to participate in stormwater solutions



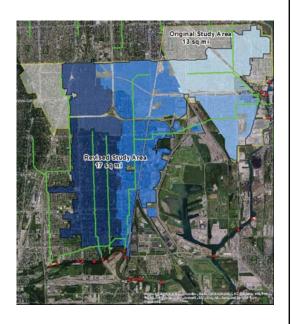


Study Area

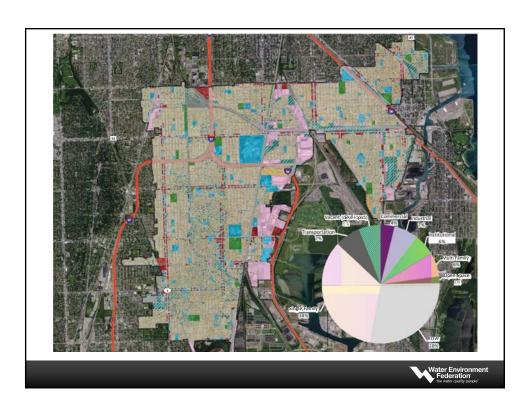
- 17 square miles493 catchments
- 4 major sewersheds
- 44,053 structures (excludes garages)

Structures flooded:

- 5 yr: 25,466 (58%)
- 25 yr: 32,610 (74%)100 yr: 41,188 (93%)









Conceptual Tunnel Alternative

- 5 year level of service (only)
- Northern portion of Area 4 (only)
- Structures removed: 27,131
- Preliminary

estimate: \$255M +

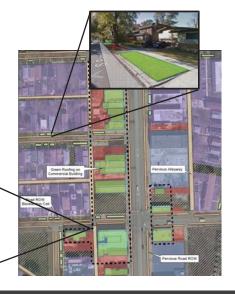






- Performed intense screening of GI applicability within the study area
 - Identify viable GI practice alternatives for urban landscape of Chicago
 - Determine maximum extent of GI implementation
- Associated GI practices with each land use category (defined in model)







Summary of Unit Cost Estimates and Model Input Unit Costs

Gl Practice		Units	Low (-30%) Cost Estimate	Model Input Cost Estimate	Actual Cost Estimate	High (+50%) Cost Estimate
P1	Pervious Pavement (Commercial, Industrial, and Institutional)	SF	\$21	\$25	\$34	\$46
P2fP3	Pervious Pavement (Roadway ROW and Residential Alleys)	SF	\$20	\$24	\$32	\$45
B1	Bioretention (ROW)	SF	\$29	\$35	\$45	\$62
B2	Bioretention (Commercial, Industrial, and Institutional)	SF	\$18	\$22	\$27	\$38
B3	Bioretention (Residential)	SF	\$19	\$23	\$31	\$43
C1	Aboveground Cisterns on Residential Properties	CF	\$55	\$66	\$85	\$117
C1	Below-Ground Cisterns on Residential Properties	CF	\$75	\$90	\$100	\$150
C1	Alleyway Concept #1—Below-Ground Aluminized CMP with Asphalt Pavement	CF	\$62	\$74	\$97	\$132
C1	Alleyway Concept #2—Below-Ground ChamberMaxx Storm Arch with Asphalt Pavement	CF	\$30	\$36	\$47	\$65
C1	Alleyway Concept #3—Below-Ground ChamberMaxx Storm Arch with Pervious Pavement	CF	\$35	\$42	\$55	\$75
C1	Alleyway Concept #4—Below-Ground StormTrap with Asphalt Pavement	CF	\$44	\$53	\$69	\$94
C2	Cisterns (Commercial)	CF	\$13	\$16	\$20	\$28
G1	Green Roof (Commercial, Industrial, and Institutional)	SF	\$8	\$20	\$10	\$16

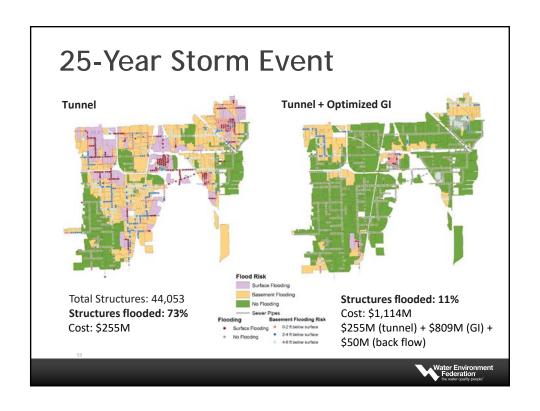
**Unit cost of \$20/square foot was selected for green roofs based on information provided by local green roof installation companies— Omni Ecosystems and LiveRoof.

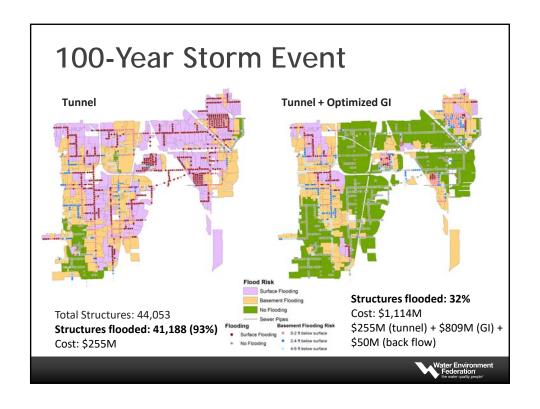


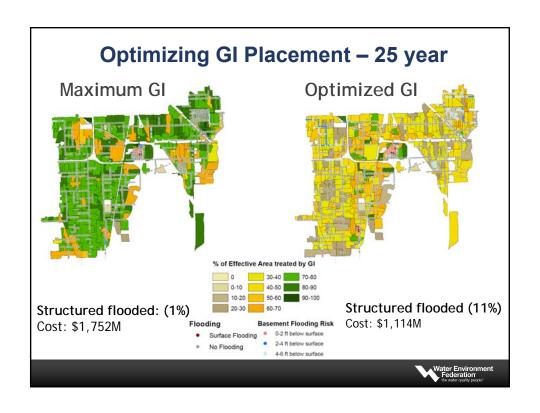
Modeling Approach

- Direct representation of GI in combined sewer model
 - Converted City's trunk sewer model from InfoWorks to SWMM and incorporated GI into the SWMM model
 - Allowed direct comparison of green vs gray performance
- Optimization of GI types, coverage, and placement locations
 - 70,000+ combinations (comparing performance & cost)
 - Evaluated targeted scenarios (implementation strategies)









Findings

- Traditional solutions can lack resiliency
- · GI is effective & placement can be optimized



Next Steps -Stormwater Master Planning

- Evaluate Master Planning needs throughout Cook County
- Develop outcome-driven but adaptive approach, centered on managing local stormwater issues with multi-disciplined teams
- Establish logical watershed/sewershed based study areas and prioritization
- Create a guidance document to guide master planning for each study area



Our Next Speaker



Prateek Joshi





Operational Analytics for Water Treatment Plants

Prateek Joshi CEO, PLUTO AI prateek@plutoai.com



Who am I?

Founder of PLUTO AI. Launched in Mar 2016.

Author of 8 books on Artificial Intelligence

Affiliations: WEF, WE&RF, AWWA, LIFT, TAG











Problem

Energy is the largest expense of operating a treatment plant. No access to data intelligence to reduce the energy consumption.





Why is it a problem?

Operating reactively as opposed to proactively

No info on the **future operating conditions** of the plant

Compliance requirements force treatment plants to make **conservative decisions**, leading to higher operating costs.



Solution

Real-time operational analytics that can provide operational parameters to maximize throughput and minimize compliance risk, thereby increasing energy efficiency.





What can AI do?

Extract wisdom from treatment plant data in real time

Perform scenario planning based on future conditions

Provide **step-by-step operational insights**



How does AI do that?

It learns the behavior of the plant using **historical data**

It predicts the **future conditions**

It **forecasts the values** of key operational parameters

It **performs scenario planning** to choose the best path



Terms (mis)used in the field

Artificial Intelligence is the goal

Machine Learning is a vehicle to get there

Deep Learning is a type of vehicle that's fast, winged, and self-driving.



Data requirements for Al

Al needs <u>historical data</u> to learn about the plant's behavior and build a model

Al needs <u>real-time data</u> to predict the future behavior based on the above model



Deploying AI at a plant

Operators need information that's actionable

The information needs to be **accurate**

Always need to be **compliant**





Data-centric simulations

Simulations built on static equations only work under ideal conditions. What about real world scenarios?

Need to use data-centric simulations to understand all the future possibilities and pick the best one.





Digital Twin

<u>Digital equivalent</u> of a physical asset or a process

Digital Twin can be used for **scenario planning**

It enables us to visualize multiple future scenarios by varying key input variables



More water with less energy

Key metric: Energy spent per gallon produced

Need to correlate data to this metric and build models

Computational modeling of energy systems is critical



Reducing energy footprint

Finds key performance indicators by sifting through data

Can find the right values for the parameters to reduce energy footprint for that process





Membrane analytics

Membrane systems like RO Membranes and MBR are **energy intensive**

Al can find the <u>optimal values</u> for operational parameters like driving pressure, permeability, cleaning schedule, lifecycle impact, and more.

Al can provide <u>operational oversight</u> to help the operators ensure that there are no shocks in the system



Al vs. Excel

Excel can't predict future conditions

Excel can't build simulations

Excel can't learn and adapt



Predict. Forecast. Simulate.

Predict the occurrence of an event

Forecast the future values of parameters

Simulate various events for scenario planning



Generic Al vs. water Al

One-size-fits-all approach doesn't work in water. We need Al models that are specifically trained to understand the behavior of water treatment systems





Key takeaways

Al takes real-time plant data and **converts it to wisdom**

Al finds **optimal values** for key operational parameters

Al does **scenario planning** using data-centric simulations



THANK YOU

Prateek Joshi CEO, PLUTO Al prateek@plutoai.com



Our Next Speaker



Reese Johnson, PE, PMP









