

COLLEGE OF ENGINEERING, MATHEMATICS AND PHYSICAL SCIENCES

A Risk-Based Decision Support System for Failure Management in Water Distribution Networks

Submitted by Josef Bicik to the University of Exeter
as a thesis for the degree of
Doctor of Philosophy in Engineering
in June 2010

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ABSTRACT

The operational management of Water Distribution Systems (WDS), particularly under failure conditions when the behaviour of a WDS is not well understood, is a challenging problem. The research presented in this thesis describes the development of a methodology for risk-based diagnostics of failures in WDS and its application in a near real-time Decision Support System (DSS) for WDS' operation.

In this thesis, the use of evidential reasoning to estimate the likely location of a burst pipe within a WDS by combining outputs of several models is investigated. A novel Dempster-Shafer model is developed, which fuses evidence provided by a pipe burst prediction model, a customer contact model and a hydraulic model to increase confidence in correctly locating a burst pipe.

A new impact model, based on a pressure driven hydraulic solver coupled with a Geographic Information System (GIS) to capture the adverse effects of failures from an operational perspective, is created. A set of Key Performance Indicators used to quantify impact, are aggregated according to the preferences of a Decision Maker (DM) using the Multi-Attribute Value Theory. The potential of distributed computing to deliver a near real-time performance of computationally expensive impact assessment is explored.

A novel methodology to prioritise alarms (i.e., detected abnormal flow events) in a WDS is proposed. The relative significance of an alarm is expressed using a measure of an overall risk represented by a set of all potential incidents (e.g., pipe bursts), which might have caused it. The DM's attitude towards risk is taken into account during the aggregation process.

The implementation of the main constituents of the proposed risk-based pipe burst diagnostics methodology, which forms a key component of the aforementioned DSS prototype, are tested on a number of real life and semi-real case studies. The methodology has the potential to enable more informed decisions to be made in the near real-time failure management in WDS.

ACKNOWLEDGEMENTS

Firstly, I would like to express the deepest gratitude to my first supervisor Professor Dragan Savić, who gave me this life-changing opportunity, to do my PhD in the Centre for Water Systems. I am very grateful for all the support, guidance and encouragement I received from him as well as from my second supervisor Professor Zoran Kapelan throughout the period of my studies.

My further thanks go to Dr. Christos Makropoulos for his supervision during the first year of my PhD. The stimulating discussions we held helped shape this PhD into its current form.

I would like to acknowledge the financial support received through the NEPTUNE project (grant EP/E003192/1) funded by the U.K. Science and Engineering Research Council, and Industrial Collaborators. The additional funding provided by Yorkshire Water after the end of the NEPTUNE project, which contributed towards successful completion of this PhD thesis, is also gratefully acknowledged.

The kind assistance and support of academic as well as industrial partners of the NEPTUNE project is much appreciated. In particular I would like to thank Mr Ridwan Patel from Yorkshire Water Services, Mr Derek Clucas from United Utilities, and Drs. Steve Mounce and John Machel from the Pennine Water Group.

I would like to thank all the current and past members of the Centre for Water Systems, with whom I had the pleasure of working, for creating an inspiring research environment. I am particularly thankful for the help and support received from Drs. Mark Morley, Gianluca Dorini, Darko Joksimović and Haytham Awad. Grateful acknowledgements for proofreading this thesis and correcting the English also go to Mrs. Alexandra Slater.

Finally, I would like to thank my family and my fiancé Kristyna for their love, support, patience and understanding during these years. This thesis is dedicated to them.

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LIST OF ABBREVIATIONS

AHP Analytic Hierarchy Process

AIJ Aggregation of Individual Judgements

AIP Aggregation of Individual Priorities

ANN Artificial Neural Network

API Application Programming Interface

AVG Average

BFOD Binary Frame Of Discernment

BPA Basic Probability Assignment

BPBM Pipe Burst Prediction Model

BST British Summer Time

CC Customer Contact

CCM Customer Contacts Model

CPU Central Processing Unit

CR Consistency Ratio

DB Database

DBMS Database Management System

DM Decision Maker

DMA District Metered Area

DRM Discolouration Risk Model

D-S Dempster-Shafer

DSS Decision Support System

EE Engineered Event

EM Eigenvector Method

EPS Extended Period Simulation

ES Expert System

FDD Fraction of Delivered Demand

FIS Fuzzy Inference System

FMEA Failure Mode and Effects Analysis

FMECA Failure Mode, Effects, and Criticality Analysis

FTP File Transfer Protocol

GA Genetic Algorithm

GIS Geographic Information System

GM Geometric Mean

GMT Greenwich Mean Time

HACCP Hazard Analysis and Critical Control Point

HDA Head Driven Analysis

HM Hydraulic Model

HTTP HyperText Transfer Protocol

KPI Key Performance Indicator

MAUT Multi-Attribute Utility Theory

MAVT Multi-Attribute Value Theory

MCDA Multi-Criteria Decision Analysis

NP Non-deterministic Polynomial time

NSGA Non-dominated Sorting Genetic Algorithm

ODBC Open Database Connectivity

OFWAT The Office of Water Services

OGC Open Geospatial Consortium, Inc.®

OODBMS Object-Oriented Database Management System

ORDBMS Object-Relational Database Management System

OWA Ordered Weighted Averaging

PCR Proportional Conflict Redistribution

PDD Pressure Dependent Demand

PHP Personal Home Page

PNG Portable Network Graphics

PRV Pressure Reducing Valve

PVC Polyvinyl Chloride

RDBMS Relational Database Management System

RI Random Index

R-T Real-Time

SBX Simulated Binary Crossover

SCADA Supervisory Control And Data Acquisition

SCEM-UA Shuffled Complex Evolution Metropolis

SSE Sum of Squared Errors

TBL True Burst Location

TBM Transferable Belief Model

TPD Third Party Damage

UI User Interface

UML Unified Modelling Language

WA Weighted Average

WDS Water Distribution System

WFS Web Feature Service

WMS Web Map Service

WMSY Work Management System

WSS Water Supply System