

Supporting Information

Cost-effective River Water Quality Management using Integrated Real-Time Control Technology

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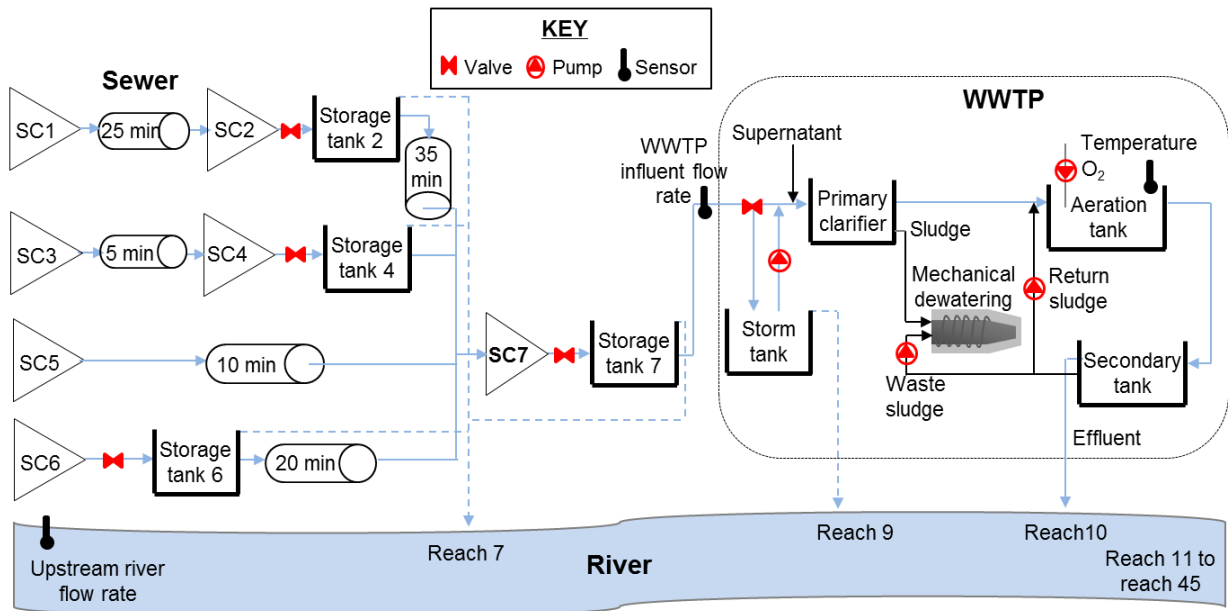
45 **1. Definition of the case study site**

46 **1.1 System configuration and modelling**

47 The dimensions and modelling methods of the catchment, the treatment process units
 48 and the river are provided in Table S1. The layout of the integrated UWWS is shown in
 49 Figure S1.

50 *Table S1 Dimensions of the case study UWWS and the modelling methods*

Process unit	Dimension	Hydraulic/pollutant transport model	Models for sedimentation	Models for biochemical reaction processes
Catchment	Total area of 7 sub-catchments: 7.26 km ²	Nash cascade	Not modelled	
Sewer	--	Translation		
Storage tank	Tank 2: 2800 m ³ ; Tank 4: 1400 m ³ ; Tank 6: 2000 m ³ ; Tank 7: 7000 m ³ ;		Simplified model by a coefficient of settling efficiency	Not modelled
Storm tank	6750 m ³			
Primary clarifier	6785 m ³	Completely mixed reactors	Empirical equation as a function of hydraulic retention time (HRT)	
Aerator	10,400 m ³		--	An extension of Activated Sludge Model No. 1
Secondary clarifier	6600 m ³		3-layer model, using exponential function to simulate settling velocity	Not modelled
Mechanical dewatering	--	--	Idealised solid separation	
River	45 km	SWMM5	Not modelled	Lijklema



51

52 *Figure S1 Schematic representation of the case study site (SC: sub-catchment)*

53 Similar as in previous literature^{1,2}, different levels of simplifications were adopted in
 54 the modelling as it is impractical to simulate in depth all (possibly known) processes in
 55 the context of integrated modelling. Hence, some processes are simulated in a simple
 56 manner (e.g. mixing, sedimentation in storm/storage tanks, sludge dewatering) or not
 57 included (e.g. biochemical reactions in the sewer, sedimentation in the river).
 58 Nevertheless, processes critical for wastewater treatment and its environmental impacts,
 59 namely sedimentation in the secondary clarifier and biochemical reactions in the
 60 aeration tank and the receiving river, are modelled in a relatively detailed manner.

61 pH and variable temperature are not included in the river water quality model. As a
 62 result, the biochemical reactions (e.g. nitrification, BOD deoxygenation) influenced by
 63 temperature change are modelled based on constant temperature (17 °C) over the
 64 simulated year. However, the effect was found to be of minor significance by changing
 65 the temperature setting from 5 °C to 30 °C. Also, the generation of un-ionized ammonia,

66 which is toxic to fish and controlled by the UK regulation (99%ile: 0.04 NH₃-N mg/L)³,
 67 cannot be simulated because pH and temperature are the key factors influencing the
 68 equilibrium between un-ionized ammonia and ionized ammonia. Still, its risk can be
 69 estimated from the simulation results on total ammonia given the river pH and
 70 temperature⁴. For the studied river, the risk is considered to be low as the un-ionized
 71 ammonia limit is automatically complied with if the total ammonia limit is met under
 72 conditions where the river pH is lower than 8.0 and the temperature below 25 °C or at a
 73 higher pH below 8.5 and the temperature lower than 10 °C.

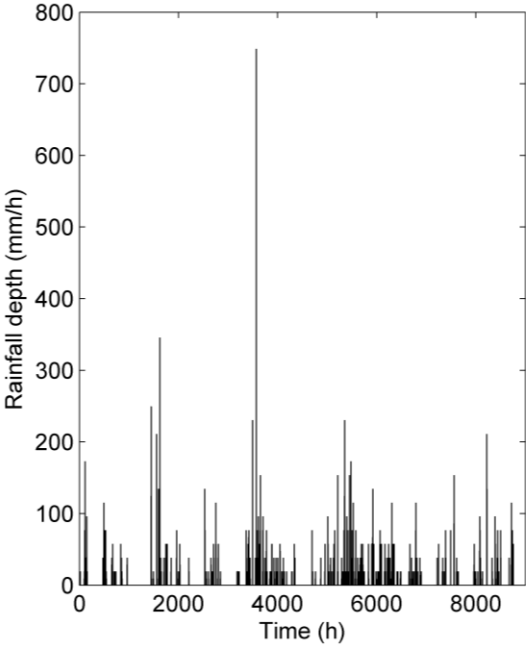
74 1.2 Flow and water quality input data

75 The flow and water quality data of the DWF in the sewer system ², rainfall runoff ² and
 76 supernatant flow from the sludge dewatering unit in the WWTP⁵ are presented in Table
 77 S2. The values for the runoff and supernatant are assumed to be constant in the
 78 simulation, while that for the DWF are average values and are used by multiplying pre-
 79 defined diurnal patterns².

80 *Table S2 Flow and water quality data for dry weather flow, rainfall runoff and*
 81 *supernatant flow*

	Flow rate (L/s)	Water quality (mg/L)					
		COD	COD _{soluble}	SS	VSS	NH ₄ +NH ₃	NO ₃
Dry weather flow	318.3	606	281	335	245	27.7	0
Rainfall runoff	--	100	46	190	139	2	0
Supernatant	20	8,221	84	7,595	6,155	12	0

82 A one-year simulation was set up so that long-term performance of the system can be
83 evaluated. In the original model established for this case study site², the evaluation of
84 system performance was rather short-term (e.g. one week) so wastewater temperature
85 and upstream river flow rate and water quality were assumed to be constant. To
86 accommodate long-term simulations, a pattern of seasonal wastewater temperature
87 was defined and one-year input data sets (rainfall and corresponding river data) were
88 incorporated into the model. As no monitoring data on temperature of the Norwich
89 WWTP were available, a seasonal pattern (18 °C, 23 °C, 19 °C and 15 °C from spring
90 to winter) was assumed by adjusting a WWTP wastewater temperature pattern reported
91 in the literature⁶ to data on the local climate of Norwich⁷. The rainfall time series is
92 shown in Figure S2.



93

94

Figure S2 Rainfall data (Oct 2012 to Oct 2013) for the case study

95 **2. Formulation of operational cost**

96 Energy cost refers to the expenditure incurred in pumping, aeration and sludge
97 treatment as calculated using Equations (S1)-(S4):

98
$$\text{Operational cost} = C_{\text{pump}} + C_{\text{aeration}} + C_{\text{sludge}} \quad (\text{S1})$$

99
$$C_{\text{pump}} = 0.16 \times E_{\text{pump}} \quad (\text{S2})$$

100
$$C_{\text{aeration}} = 0.16 \times E_{\text{aeration}} \quad (\text{S3})$$

101
$$C_{\text{sludge}} = 1.24 \times 10^{-4} \times V_{\text{ts}} \times C_{\text{ts}} \quad (\text{S4})$$

102 where C_{pump} (\$) is the cost for pumping, E_{pump} (kWh) is the total electricity
103 consumption from pumping within the simulation period, C_{aeration} (\$) is the cost for
104 aeration, E_{aeration} (kWh) is the total electricity consumption from aeration, C_{sludge} (\$) is the
105 cost for sludge treatment, V_{ts} (m³) is the total volume of thickened waste sludge, and C_{ts}
106 (mg/L) is the concentration of the thickened waste sludge. The constant 0.16 is the
107 electricity tariff rate (\$/kWh) defined for pumping and aeration in this study. The
108 constant 1.24×10^{-4} is the mechanical dewatering cost (\$) per gram of dry waste
109 sludge⁸.

110 **3. Value ranges for operational variables**

111 *Table S3 Baseline values and ranges of the operational variables (unit: m³/d)*

Operational variable	Baseline value	Lower bound value	Higher bound value
Tank 2 overflow threshold	24,900	15,000	40,000

Tank 4 overflow threshold	11,300	6,800	18,000
Tank 6 overflow threshold	23,000	14,000	37,000
CSO (tank 7)	137,500 (i.e. 5DWF)	82,500 (i.e. 3DWF)	220,000 (i.e. 8DWF)
Storm tank overflow threshold	82,500 (i.e. 3DWF)	55,000 (i.e. 2DWF)	137,500 (i.e. 5DWF)
Storm tank emptying threshold	24,000	16,800	31,200
Storm tank emptying rate	12,000	7,200	24,000
Return sludge pumping rate	14,400	7,200	24,000
Waste sludge pumping rate	660	240	960
Aeration rate	720,000	240,000	1,200,000

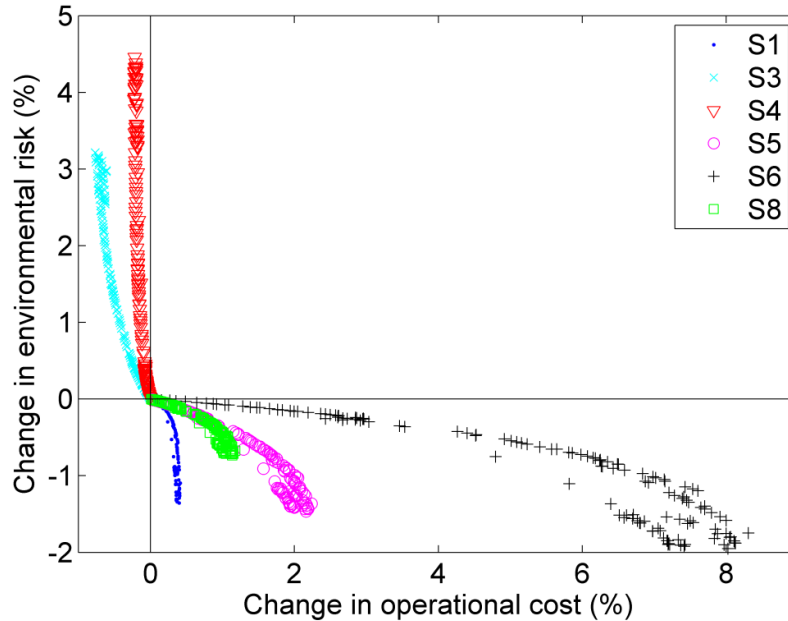
112 **4. If-Then control rules for the case study**

113 *Table S4 RTC rules for aeration rate control in accordance to wastewater inflow rate,*
114 *temperature and upstream river flow rate*

Scenario	Wastewater inflow rate to the WWTP (m ³ /d)	Temperature (°C)	Upstream (reach 2) river flow rate (m ³ /d)	Aeration rate tier (m ³ /h)
S1	> 41,250	> 15	> 300,000	Y ₁
S2	<= 41,250	> 15	> 300,000	Y ₁
S3	> 41,250	<= 15	> 300,000	Y ₂
S4	> 41,250	> 15	<= 300,000	Y ₂

S5	$\leq 41,250$	≤ 15	$> 300,000$	Y_1
S6	$\leq 41,250$	> 15	$\leq 300,000$	Y_1
S7	$> 41,250$	≤ 15	$\leq 300,000$	Y_2
S8	$\leq 41,250$	≤ 15	$\leq 300,000$	Y_1

115 The suitability of the assignment of the aeration tier for each scenario is tested. As it is
116 certain to assign Y_2 to the 'worst' environmental condition and to assign air flow Y_1 to
117 the 'best', S2 and S7 need not to be examined. For the rest of the scenarios, the
118 assignment of aeration tier is tested by changing it to the alternative option (i.e. from Y_1
119 to Y_2 , or Y_2 to Y_1) and checking if great improvement in system performance can be
120 achieved. The changes in the two objectives are presented in Figure S3. By altering the
121 aeration tier from Y_2 to Y_1 for S3 and S4, cost reduction can be achieved but with a
122 disproportionate increase in risk. Similarly, disproportional cost is increased if the
123 aeration tier Y_1 is changed to Y_2 for S5, S6 and S8. It is uncertain however of whether
124 the aeration tier for S1 needs to be changed from the produced results. The slope of the
125 curve suggests more percentage of risk can be reduced by a lower percentage of cost
126 increase. Nevertheless, the rule is not changed because a) the amount of change is
127 marginal and b) the reduction in operational cost is harder to achieve for this case
128 compared to the environmental risk. Note that if the aeration tier of S1 is changed, the
129 framework of the RTC strategies will be altered, as the condition of river flow rate will be
130 redundant for the "If-Then" rules. Therefore, the suitability of parameters selected for the
131 RTC rule conditions can be checked through the optimization of the controlled variable
132 values.



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134 *Figure S3 Changes in operational cost and environmental risk by varying aeration tiers*
 135 *from Y1 to Y2 or Y2 to Y1 of S1, S2-6 and S8*

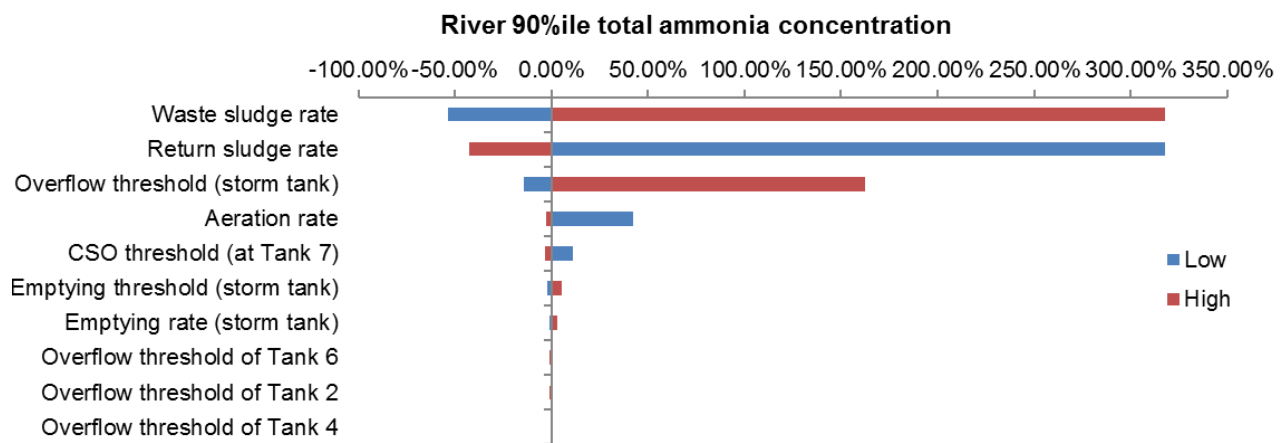
136 **5. OAT analysis results**

137 In the OAT analysis, the setting of one operational variable is changed at a time (to
 138 the lower or higher bound value), while keeping others at their baseline values. Then
 139 the variable is returned to its baseline value, and the process is repeated for each of the
 140 other variables in the same way. The baseline and lower and higher bound values of the
 141 operational variables are listed in Table S3. Sensitivity is measured by running a one
 142 year wet weather simulation and recording the value changes in the output parameters
 143 (i.e. cost and environmental objectives as defined in Equations (1) to (3)).

144 Results of the OAT analysis are represented in tornado graphs from Figure S4 to
 145 Figure S6, where the operational variables are ranked by the greatest range of
 146 percentage change for any model output. For example, as shown in Figures S4, the
 147 waste sludge rate produces a 318% increase in the 90%ile river total ammonia

148 concentration compared to the base scenario when the waste sludge rate is at the high
 149 bound (960m³/d), and a 54% decrease at the high bound (240 m³/d). The difference
 150 between 318% and -54% is the largest among all operational variables. Results suggest
 151 that waste sludge pumping rate, return sludge pumping rate, overflow threshold of the
 152 storm tank in WWTP and aeration rate are the most essential factors influencing
 153 environmental total ammonia concentration; aeration is the major source of operational
 154 cost, followed by waste sludge pumping rate and return sludge pumping rate.

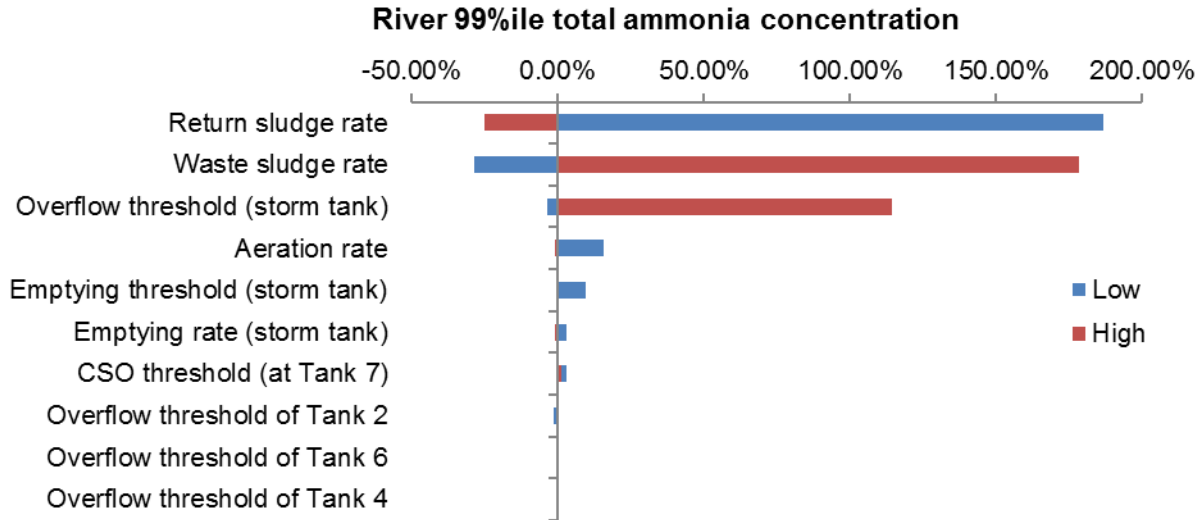
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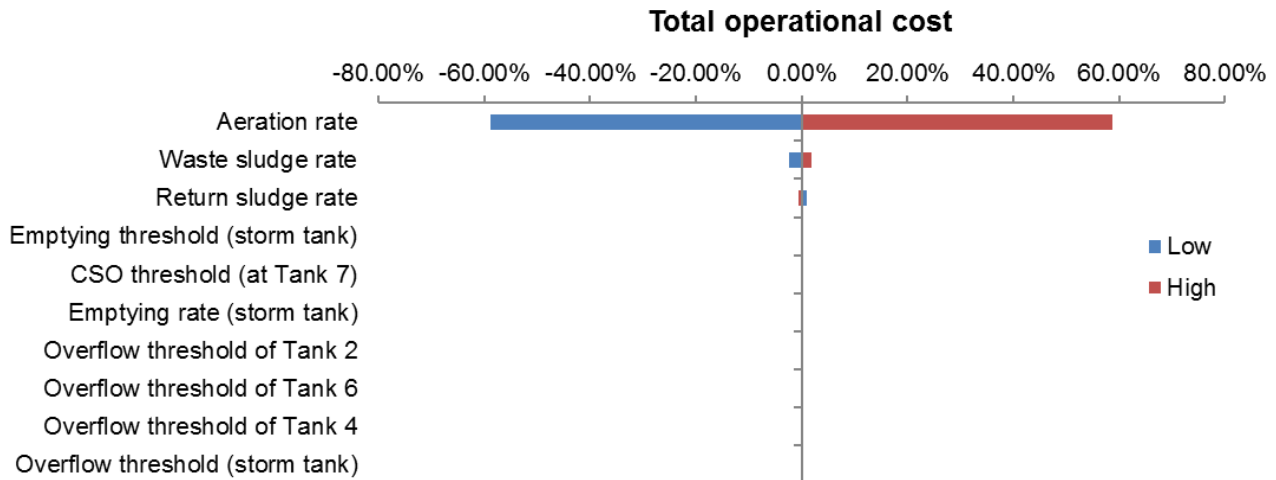
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157 *Figure S4 Percentage changes in river ammonium 90% values in reach 11 from the*
 158 *base case when operational variable values are varied to low bound and high bound*
 159 *values*

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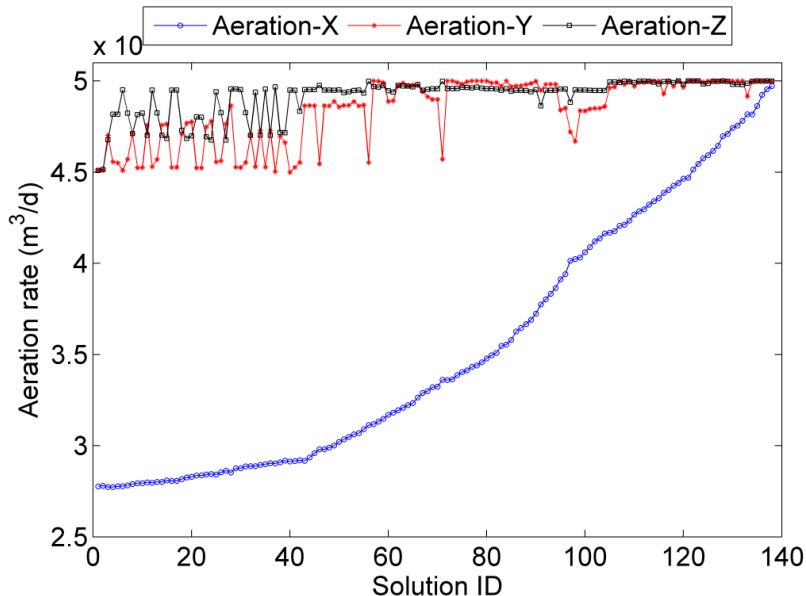
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 162 *Figure S5 Percentage changes in river ammonium 99% values in reach 11 from the*
 163 *base case when operational variable values are varied to low bound and high bound*
 164 *values*



165
 166 *Figure S6 Percentage changes in total operational cost from the base case when*
 167 *operational variable values are varied to low bound and high bound values*

168 **6. Optimization results with three aeration tiers**

169 Figure S7 shows the variable values of the optimal solutions when three aeration tiers
 170 are used and optimised by NSGA-II. Each solution corresponds with one set of X, Y and
 171 Z values. As shown in the figure, the values of Y and Z for most solutions are close,
 172 suggesting only two aeration tiers could be sufficiently enough.



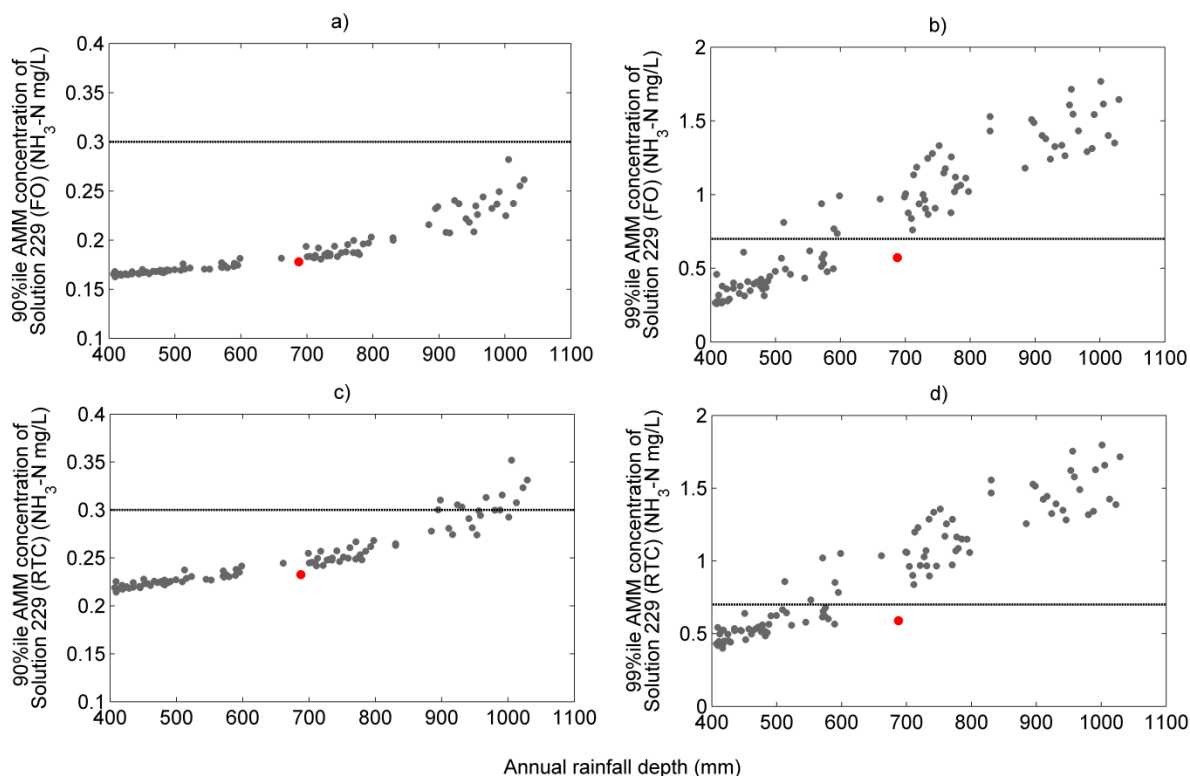
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175 *Figure S7 Operational variable values of the optimal RTC solutions with three aeration*
 176 *tiers*

177 7. Uncertainty analysis against rainfall input

178 The uncertainty of system performance against rainfall changes are presented by
 179 results of two typical solutions, i.e. *GoodSol (OO)* and *GoodSol (RTC)* as shown in
 180 Figure S8. Dynamic simulation is run for each rainfall input data series and the resulting
 181 environmental quality values are shown in the figure as compared with those by the
 182 original rainfall data (shown in red dots). It can be seen from Figures S8b and S8d that
 183 the 99%ile limits can be easily violated even under less intensive (measured in total
 184 depth) rainfall inputs. This is expected as the 99%ile total ammonia concentration is
 185 highly influenced by sewer overflows thus cannot be effectively addressed by control
 186 measures in the WWTP. The effluent water quality is less affected by rainfall variations,
 187 which in turn results in satisfactory 90%ile total ammonia concentration in the river.
 188 Though the control solution has 19% less headroom to the 90%ile standard limit than

189 the fixed operational solution, it is shown to withstand 30% more intensive rainfall
190 without violating the standard limit. This suggests a high robustness of the strategy to
191 precipitation changes, as a 10% rainfall increase (in total depth) until 2050 is what used
192 by regulators in the UK for the preparation of climate change⁹.



193
194 *Figure S8 Comparison of performance of GoodSol (OO) (a and b) and GoodSol (RTC)*
195 *(c and d) under 100 one-year rainfall input data series*

196 As the headroom decreases such as stressed by a higher environmental target or in
197 pursuit of a lower cost solution, the robustness of the strategy reduces. For example,
198 the RTC solution in Figure 8 could only cope with 7% more intensive rainfall if the
199 90%ile limit is changed to 0.25 NH₃-N mg/L (i.e. headroom diminishing from 22% to 6%).
200 As such, the trade-offs between cost savings and confidence level of regulatory
201 compliance should be appraised and understood to choose a balanced solution.

202

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