

Reducing Unplanned Repair Costs using Predictive Maintenance with MATLAB

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Presenter Biography

- Completed Mechanical Engineering Degree at John Moore's University in Liverpool
- Worked in various sectors including Rail, Water and Sewage Treatment, Construction/Building Services and Nuclear.
- Nuclear Industry for last 20 Years at Capenhurst Site
- Regional Chair of IMechE committee (Merseyside and North Wales)
- Sit on the Industrial Advisory Board For John Moore's University.
- Primarily involved in Storage and Decommissioning including care and maintenance of Storage Containers and the buildings they sit in, with a view to eventual decommissioning.



- The 1st centrifuge plant on the Capenhurst site, E21 was built in the late 1970's and had an output of 200 tSW/a. It took 3 years to decommission.
- URENCO (Capenhurst) Limited became a wholly owned subsidiary of URENCO Limited in 1993 following a restructuring of BNFL and in 2008 it changed it's registered name to URENCO UK Limited (UUK).
- UUK operates world leading centrifuge technology to enrich uranium to enable nuclear power stations around the world to generate electricity.
- The site operates three plants producing a total output of 4600 tSW/a.

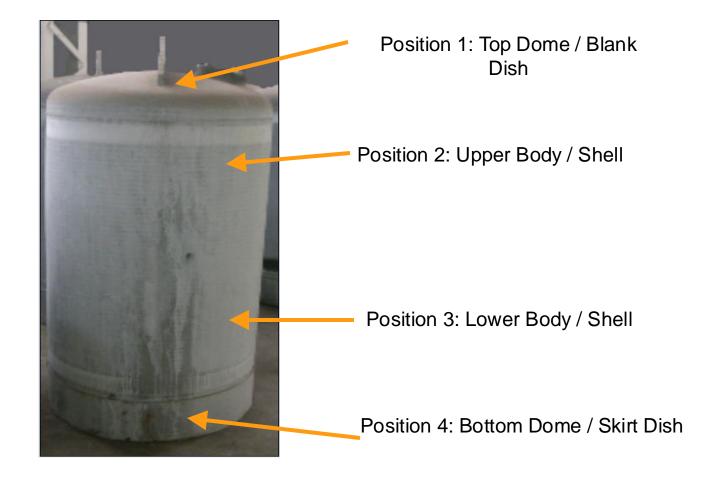
Problem Statement



- Urenco needs to manage an aging inventory of waste containers.
- A scheduled maintenance plan would involve replacing the containers at fixed intervals, regardless of their condition.
- We want to minimise unnecessary worker exposure to a hazardous environment, replacement and inspection costs.
- Gathering data is hazardous, time-consuming, and expensive.
- Can we improve on a fixed maintenance schedule by modelling and simulating the container condition as a function of time, usage, and environmental conditions?

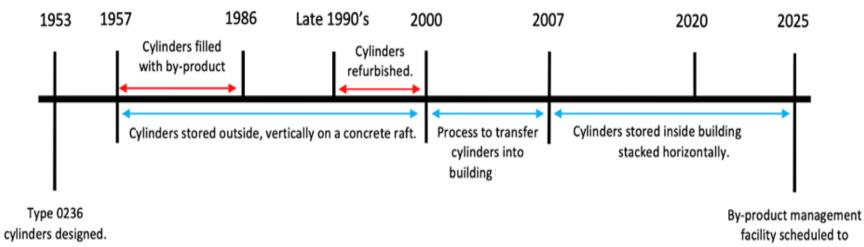
Storage Containers





Time Line

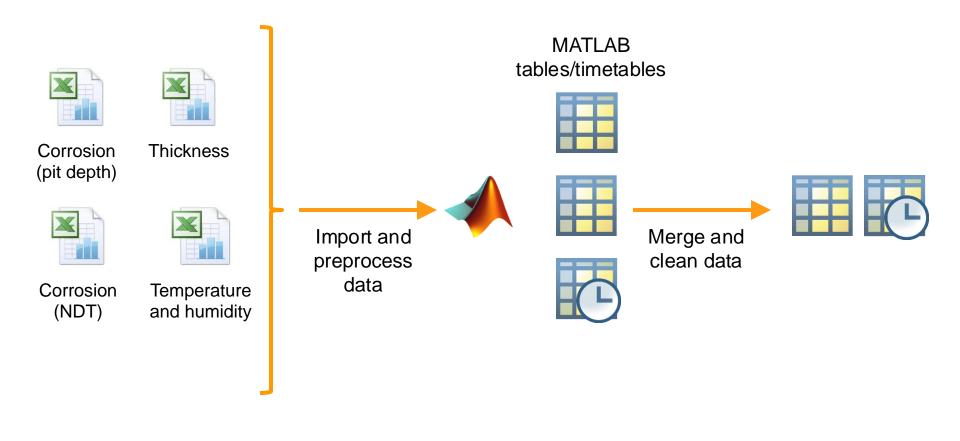




come online.







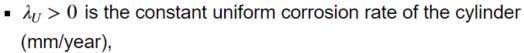


Model Equation

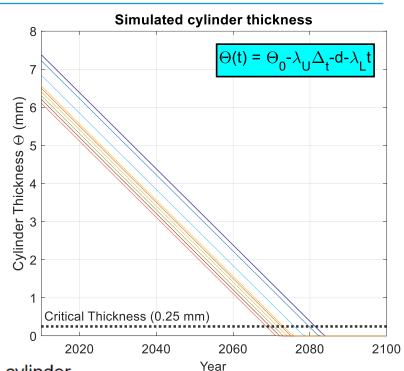
$$\Theta(t) = \Theta_0 - \lambda_U \Delta_t - d - \lambda_L t$$

where:

- $\Theta(t)$ is the cylinder thickness at time *t* (mm),
- Θ_0 is the current cylinder thickness (mm),
- *d* is the corrosion depth (mm),
- *t* is the time (years),
- $\lambda_L > 0$ is the local corrosion rate (mm/year),

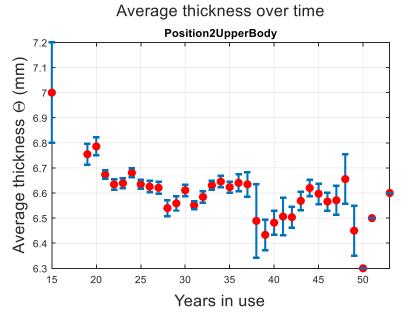


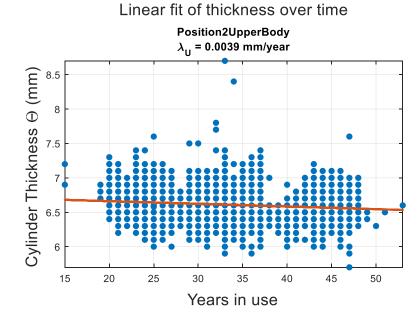
 Δ_t is the time difference (in years) between the measurement of the corrosion depth and the cylinder thickness.





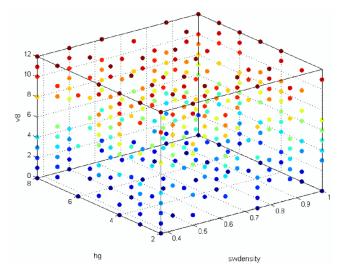
The corrosion rate parameters λ_U , λ_L are estimated using the historical data and the environmental variables (temperature and humidity).





Monte Carlo Simulation

- Some model parameters have a higher level of uncertainty.
- We need to ensure that the model is robust across the parameter space.
- The marginal distribution of each parameter is modelled using an appropriate probability distribution.
- We then sample from these distributions to produce a range of possible failure times.

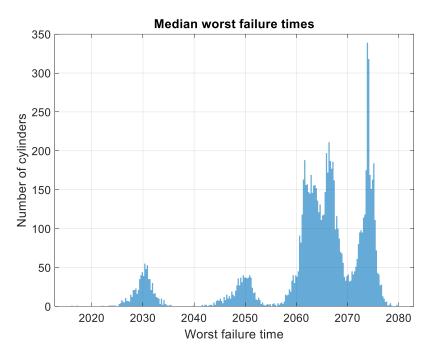




Simulation Process

- In the model equation, solve

 Θ(t) = Θ_{crit}, where Θ_{crit} is
 some critical thickness (e.g.,
 0.25mm). This gives an
 estimated failure time t.
- Repeat the simulation for each cylinder position.
- Take the minimum failure time over all the positions to obtain the overall cylinder failure time.









- Detailed prediction of containers to be examined for possible failures in spreadsheet format to ensure data can be interpreted easily.
- The model is extensible when new data is available from regular inspections.
- Graphs of predictions and timescales to be aware of and act as benchmarks for future.
- Report format enabling clear information to regulators and senior staff.





- Ability to build upon base model to add any future failure modes.
- Annual update to base information from ongoing inspections.
- Sharing the model with colleagues and regulators via a user interface.





