

*Transactions of the Korean Nuclear Society Spring Meeting
Jeju, Korea, May 23-24, 2019*

Leak Detection Method for Integrity Monitoring of Spent Nuclear Fuel Dry Storage Casks



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TECHNOLOGY

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1. Introduction (1)

● SNF Dry Storage Casks

- Operated at NPPs over more than 30 years
- Delays in establishing permanent disposal facilities
- Extended storage terms
- Issues of aging management arise
- Integrity monitoring becomes important

● Related Studies

- Canister surface temperature (CST) measurements for detecting helium gas leak from canister
 - CST as a means to detect helium leakage of a welded canister
- ※1. Hirofumi Takeda, et. al, Development of the Detecting Method of Helium Gas Leak from Canister, Nuclear Engineering and Design, Vol. 238, pp.1220-1226, 2008.
- ※2. Jie Li and Yung Y. Liu, Thermal Modeling of a Vertical Dry Storage Cask for Used Nuclear Fuel, Nuclear Engineering and Design, Vol. 301, pp.74-88, 2016.

1. Introduction (2)

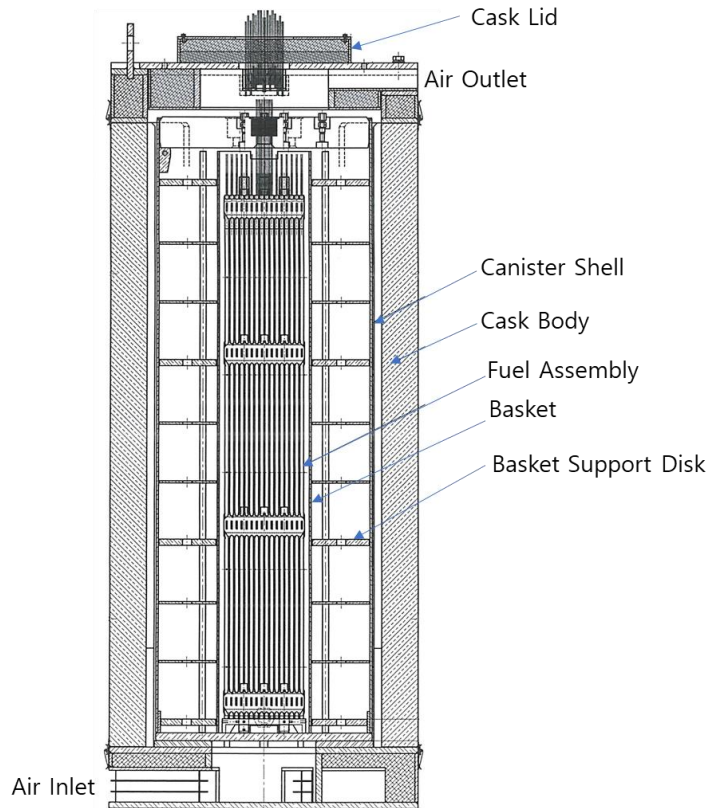
- **Focus of the Present Study**

- **Leak detection method based on CST is proposed for integrity monitoring of SNF dry storage casks**
- **Artificial neural network models are used for predicting :**
 - **canister internal pressure**
 - **peak cladding temperature**
- **Prediction method is validated through a pressure variation test**

2. Pressure Variation Test and CFD Analyses (1)

● Test Rig (1)

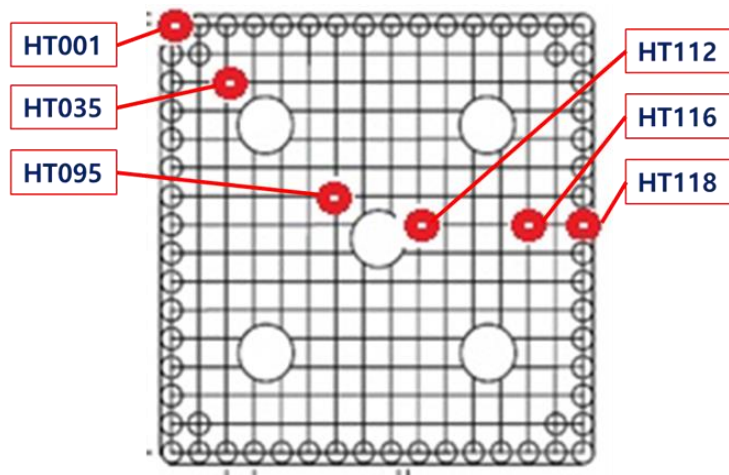
- To investigate thermal behaviors of the vertical dry storage cask
- To analyze relationships among the internal pressure, PCT, and CSTs
- To validate the prediction models



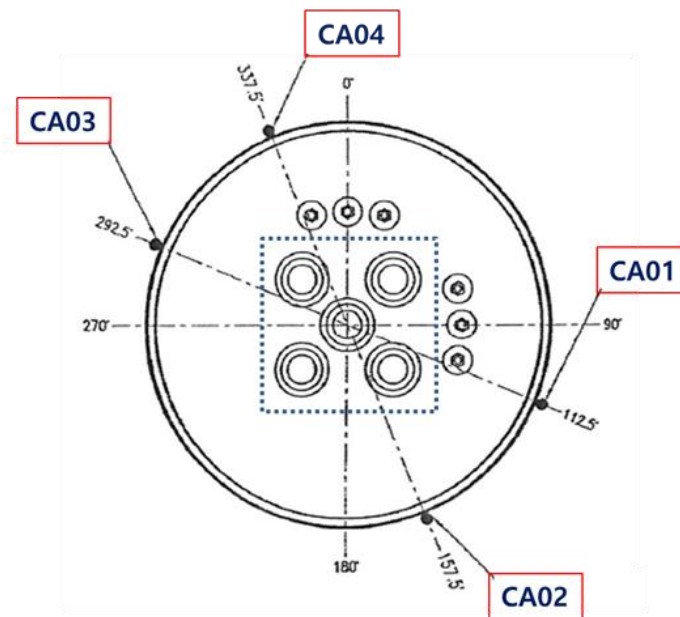
2. Pressure Variation Test and CFD Analyses (2)

● Test Rig (2)

- Scaled-down with 1/3 height with a single CE Type fuel assembly
- Fuel temperature sensors are attached at five axial levels of the active fuel length (10%, 30%, 50%, 70%, and 90%)



<Sensor-attached rods in the fuel assembly>

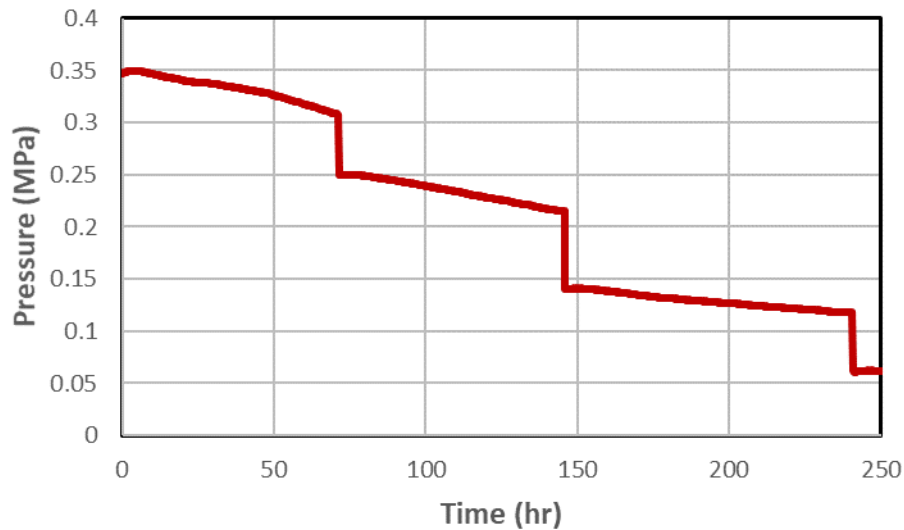


<CST measurement positions on the canister>

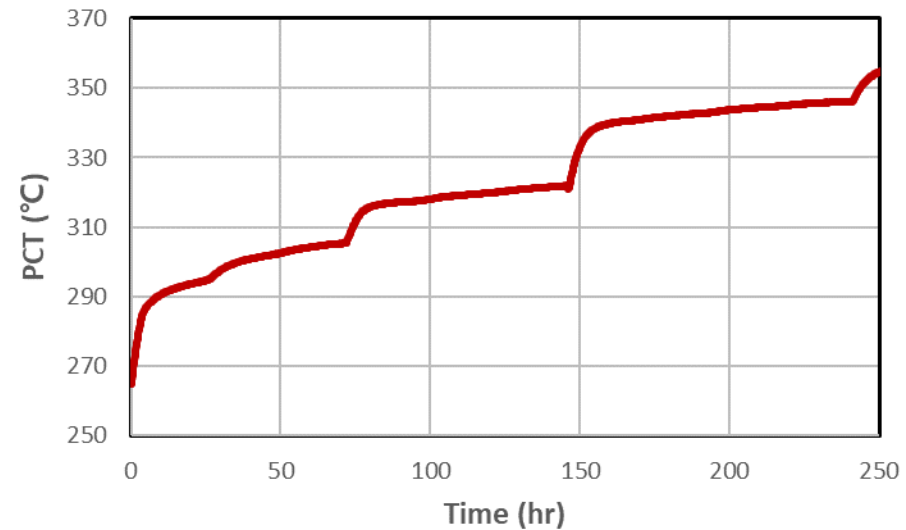
2. Pressure Variation Test and CFD Analyses (3)

● Pressure Variation Test

- To simulate helium leak in the test rig
- Assembly power of 1.7 kW
- Ambient air temperature : ~ 16 °C



<Pressure variation during the test>

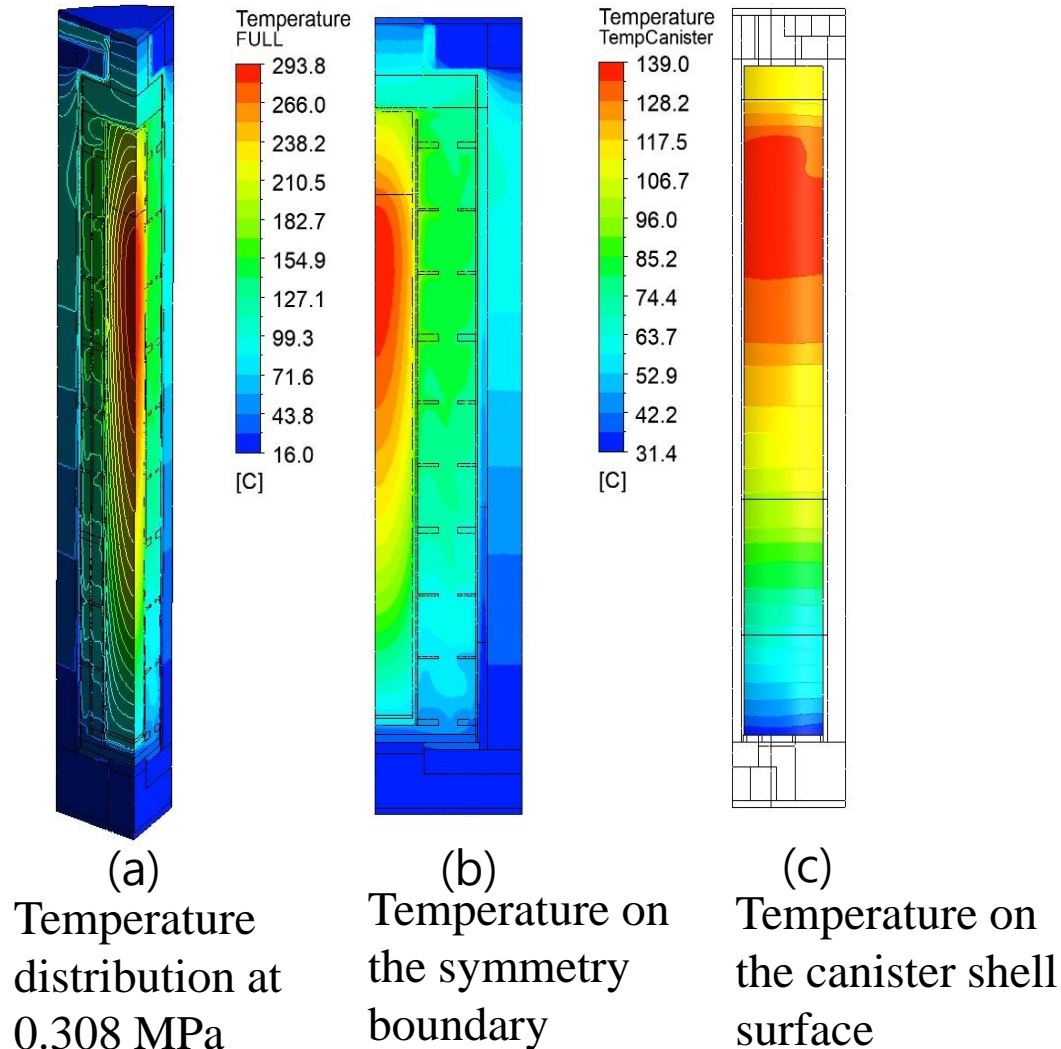


<Measured PCT during the test>

2. Pressure Variation Test and CFD Analyses (4)

- Thermal Analysis of the Test

- Steady-state FLUENT results of 1/8 symmetry 3D model



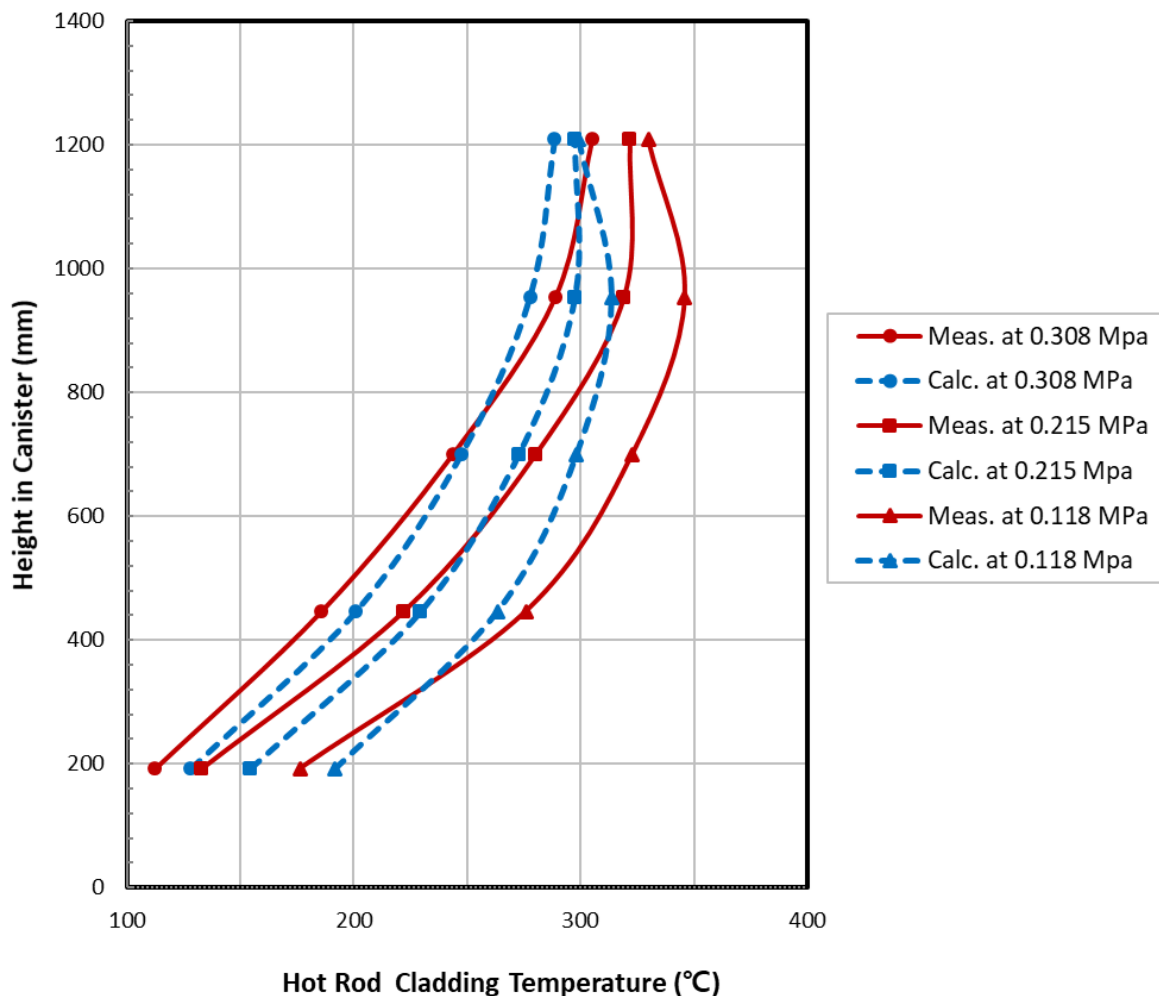
2. Pressure Variation Test and CFD Analyses (5)

- **CFD Options for Thermal Analysis**

- **The viscous model : Realizable k- ϵ model with standard wall function option.**
- **Thermal radiation : discrete ordinates model with 5x5 divisions.**
- **The pressure-velocity coupling scheme : 'coupled' algorithm.**
- **He density property : Ideal-gas law**
 - for driving force for natural convection flow
- **Fuel assembly : modeled by a porous media**
 - separate FLUENT calculations
 - Transverse effective thermal conductivity
 - Flow resistance

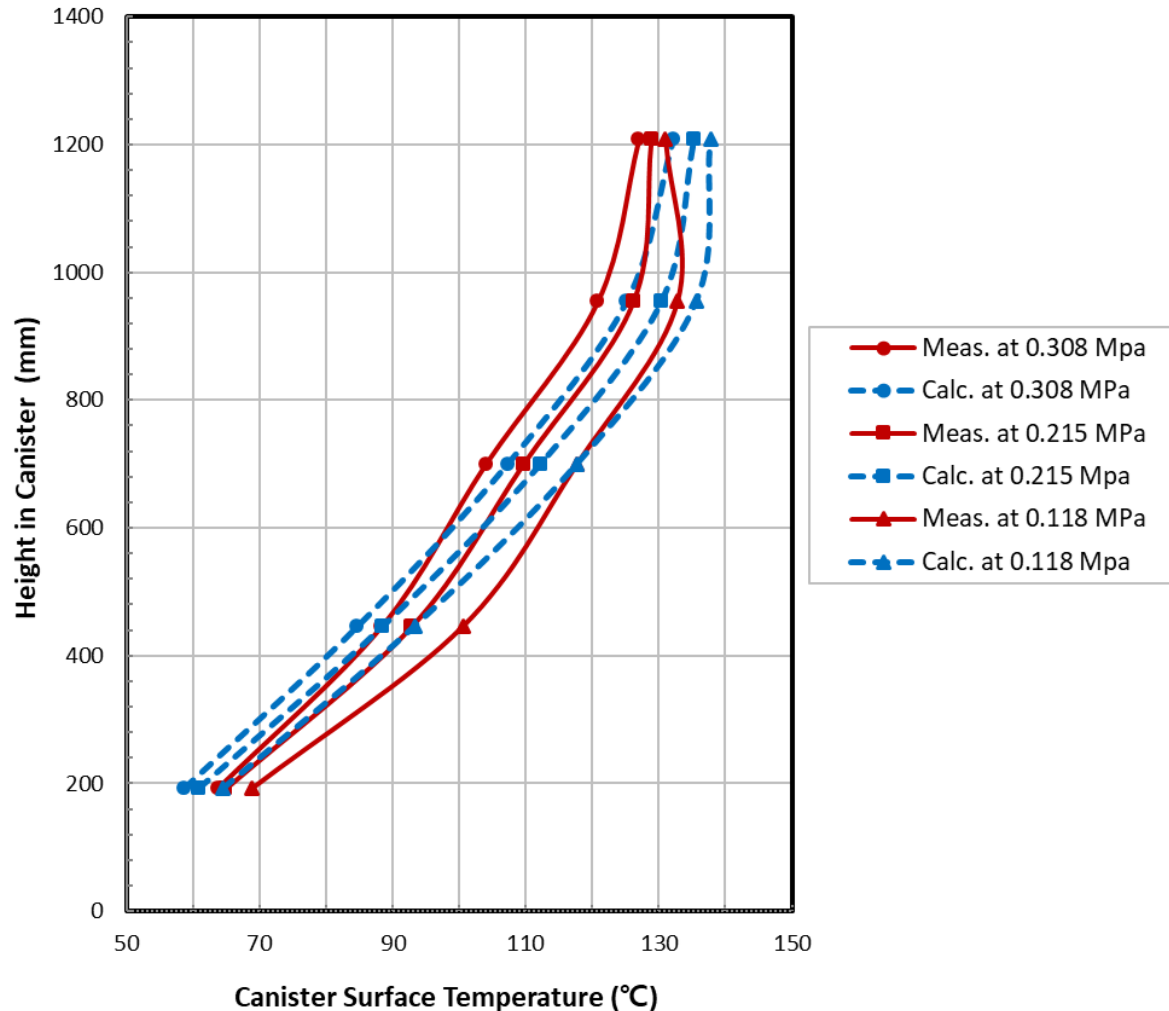
2. Pressure Variation Test and CFD Analyses (6)

● Hot Rod Temperature Results compared with Measurements



2. Pressure Variation Test and CFD Analyses (7)

- Surface Temperature Results compared with Measurements



2. Pressure Variation Test and CFD Analyses (8)

- **Summary of Analysis Results**

- **As the helium pressure decreases,**
 - **Heat conduction and convection become less active**
 - **Fuel and canister surface temperatures increase**
 - **Peak temperature tends to move towards the central zone**
- **FLUENT model underestimates the cladding temperature while overestimates the canister surface temperature**
- **Prediction performance deteriorates as the canister pressure is lowered**
- **Thermal behavioral trends over canister pressure are predictable to some degree**

3. Neural Network Prediction Engine Based on CFD Analysis Data

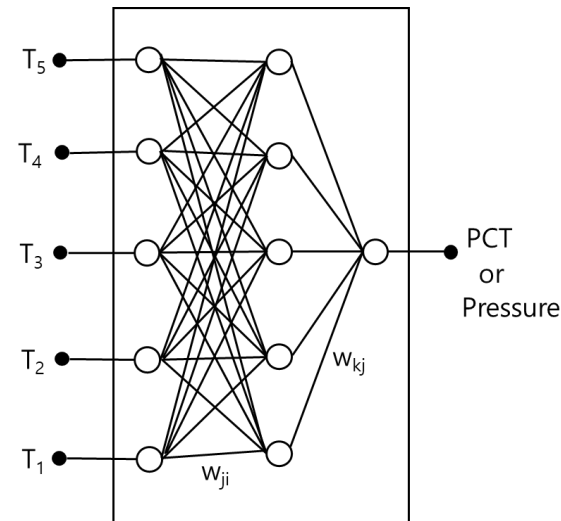
- **Database for Neural Network Training**

- 84 calculation cases
 - 7 pressures (0 ~ 0.46 MPa)
 - 3 assembly powers (1.6 ~ 1.84 kW)
 - 4 ambient temperatures (10 ~ 35 °C)

- **Neural Network Prediction Model**

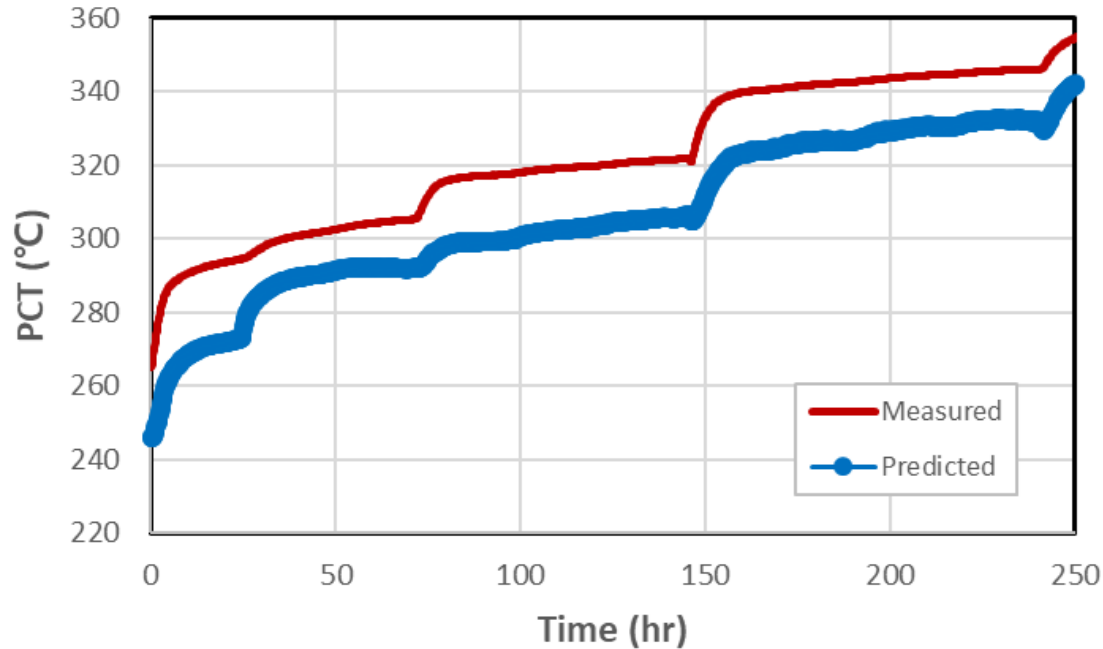
- Delta rule with the backpropagation algorithm
- Separate prediction engines for PCT and pressure
- Additional secondary inputs of temperature slopes

$$\Delta w_{ij} = -\eta \frac{\partial E}{\partial w_{ij}}$$
$$E = \frac{1}{2} \sum_p \sum_k (t_{pk} - o_{pk})^2$$



4. Prediction Engine Application to Leak Detection (1)

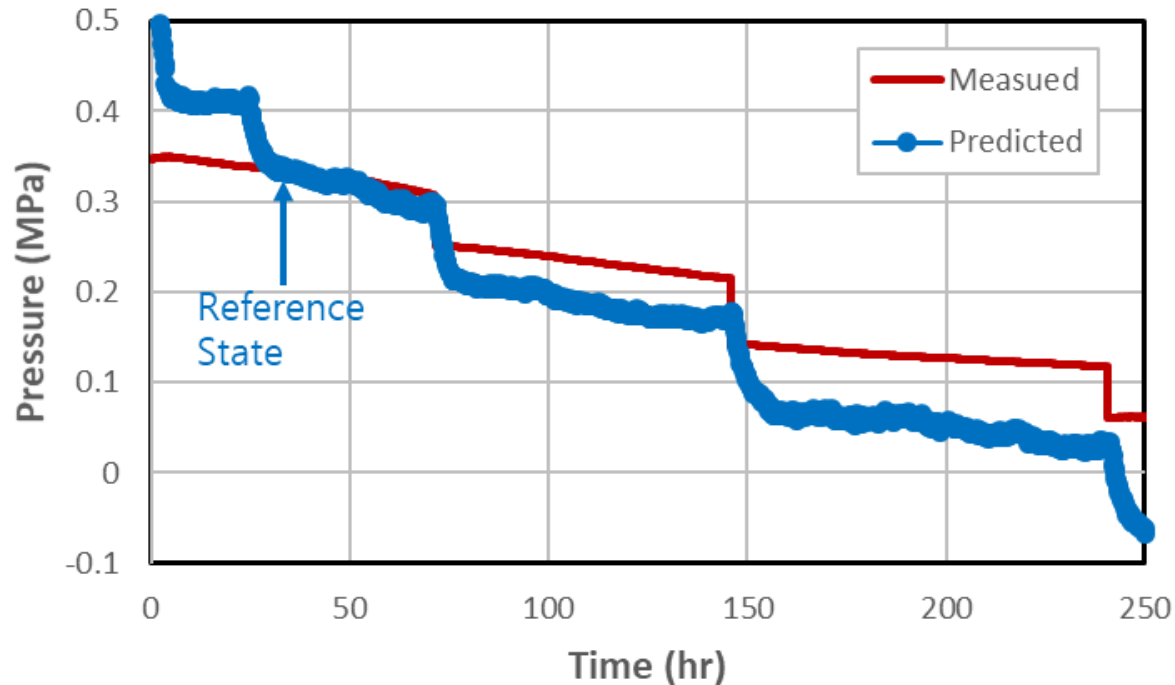
- Predicted PCT Results Compared with Measurements



<Monitoring results of peak cladding temperature>

4. Prediction Engine Application to Leak Detection (2)

- Predicted Canister Pressure Compared with Measurements



<Monitoring results of canister pressure>

4. Prediction Engine Application to Leak Detection (3)

● Assessment of Monitoring Results

- The primary goal of the integrity monitoring is not to obtain the PCT and the pressure with a high degree of accuracy, but to detect any leakage occurrence in the canister resulting from confinement degradation.
- It is important to detect any changes in the PCT and the canister pressure from the reference values.
- Initial He pressure is assumed known according to the relevant work procedure.
- In the test, a stabilized state was obtained after 33 hours of initial transient.
- For improved prediction, input temperature signals were compensated by the deviations obtained at the reference state.
- After initial dead band period, the prediction engine can perceive even a very slow loss of pressure as well as a sudden drop of pressure.
- He leak is detectable by a pressure decrease accompanied with PCT increase.
- When the pressure and the PCT changes exceed specified limit values, the system should raise an alarm for maintenance actions.

5. Conclusions

● Concluding Remarks

- A leak detection method has been developed for integrity monitoring of dry storage casks.
- The method employs a prediction engine comprising of neural network models for predicting PCT and canister pressure.
- The prediction engine uses only canister surface temperatures as input for prediction without need of sensors installed through canister wall.
- Database for training neural networks was generated by CFD calculations.
- The prediction engine has been applied to a pressure variation test for validation.
- The test results showed that helium leak in the cask is detectable by monitored pressure and PCT changes with the proposed method.



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