



TU Clausthal
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**Improving Reservoir Characterization using the Adjoint
Method in History Matching**

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(Dissertation)

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