Weekly Report III for Laboratory Research

University of Houston

June 18, 2018





Outline

Forward model: Curves

Plan for next step

Forward model: Curves

- C++ Prototype
- Matlab Result
- Python Result



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The forward model API

The basic API is

curves(Layers, Rh, Rv, Zbed, Dip, TVD)

- This project is aimed at simulating the response of an azimuthal resistivity LWD tool.
 - Layers: The number of layers.
 - Rh: [ohm * m] Resistivity \in [0.1,1000].
 - Rv: Usually the same as Rh.
 - Zbed: [feet] Boundary Position ∈ TVD+[-100, 100].
 - Dip: [deg] Dip Angle tool/boundary. When 90°, parallel to boundary.
 - TVD: True Vertical Depth. When 0, relative.



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Figure 1: The generated random model.

- Use a script to generate a random 3-layer model.
- The resistivity is {10, 50, 1} respectively.
- The Z-bed position is adaptive according to *t*.

100



10

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Figure 2: The response of the model.

The sensor number is 92. TVD is kept 0 and Dip is kept 90°.

- *t* ∈ (1, 80).
- Each sensor gives a different response.



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Figure 3: The response of the model (selective).

- We select *s*₁, *s*₂₅, *s*₅₀, *s*₇₅ and *s*₉₂ to show part of the results.
- The response is added with a small noise.



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Figure 4: The initialized model during the inversion.

- We use a flat model to predict the whole model during the initialization.
- The target is to minimize ||𝔅(p) r||, where 𝔅 is the model, p is the input parameters and r is the response.



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Figure 5: The inversed model.

The inversed model is slightly different from the real one.



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Figure 6: The python model.

- We have migrate the matlab-c-api to python-c-api.
- This is the result given by python and of the same real model.



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Figure 7: The python model response compared to the original one.

We have compared the results of the forward model defined by python API. It is almost the same with the matlab one (the difference is caused by small noise).



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Figure 8: The python model response compared to the original one (selective).

- To show the details, we compare the results from same sensors.
- We does not add random noise to python results.



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Figure 9: The model generated by migrated generator script.

- We have migrated the random generator to python script, too.
- This is another different model generated by the script.



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Figure 10: The response of generated model.

The is the simulated response from the forward model.



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- 1 Try to use the tensorflow to simulating the primal inversion algorithm.
 - Maybe tf.py_func would be a solution.
- 2 Try to implement the deep-learning scheme by introducing the forward model.