



Gravity kernel method for implicit geological modeling

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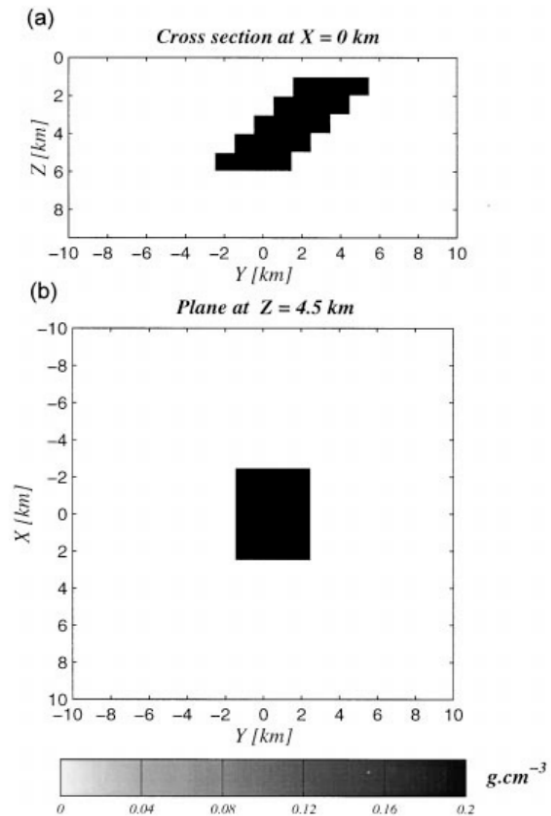
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Conventional gravity method (Voxel-based)

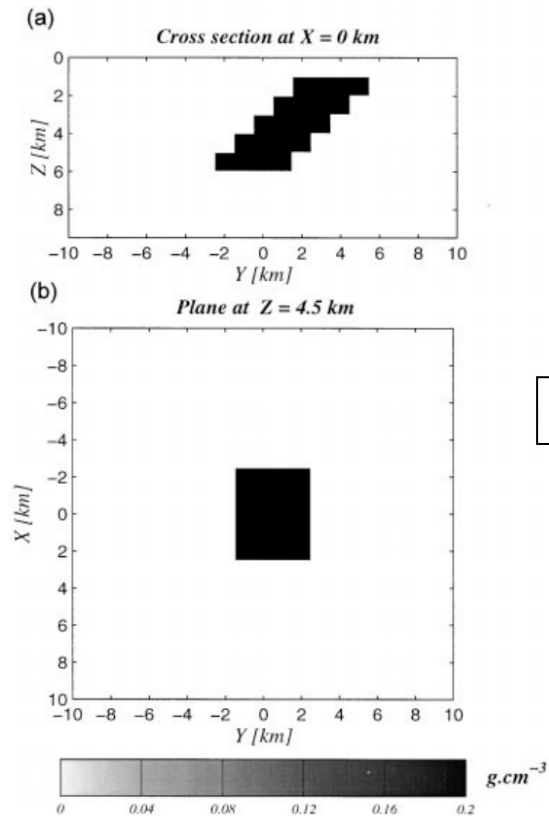
True Model



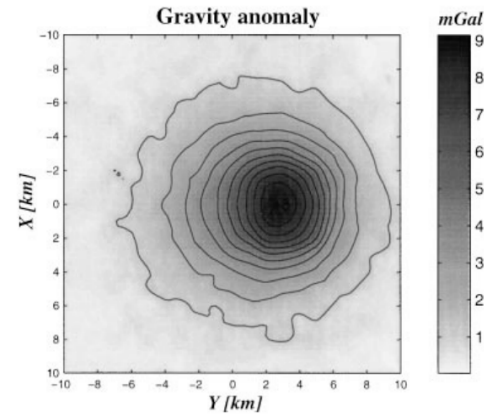
(Boulanger, O., & Chouteau, M., 2001)

Conventional gravity method (Voxel-based)

True Model



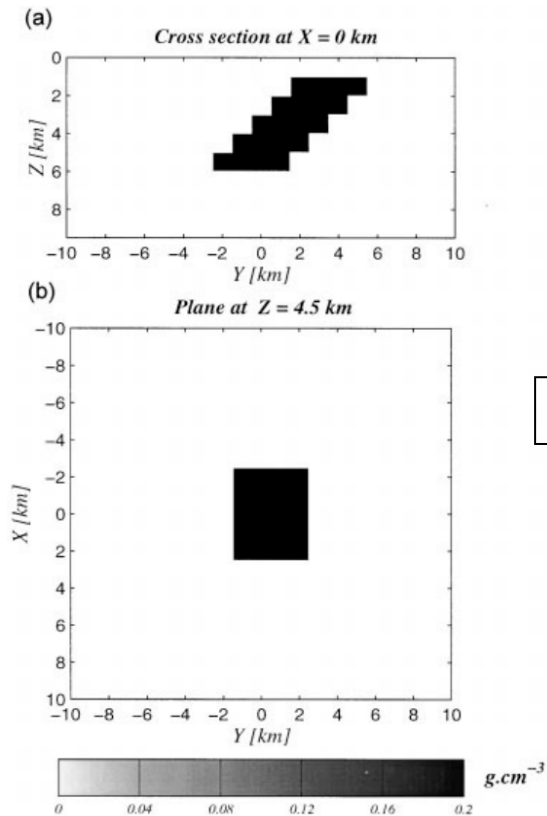
Forward



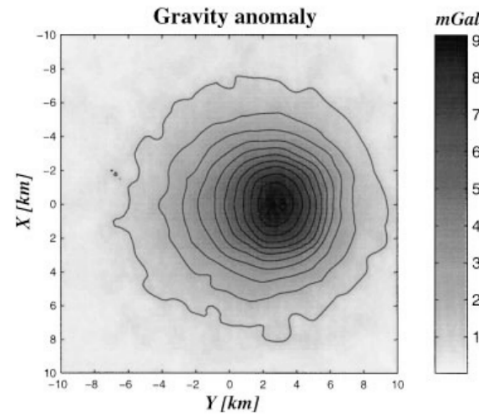
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Conventional gravity method (Voxel-based)

True Model



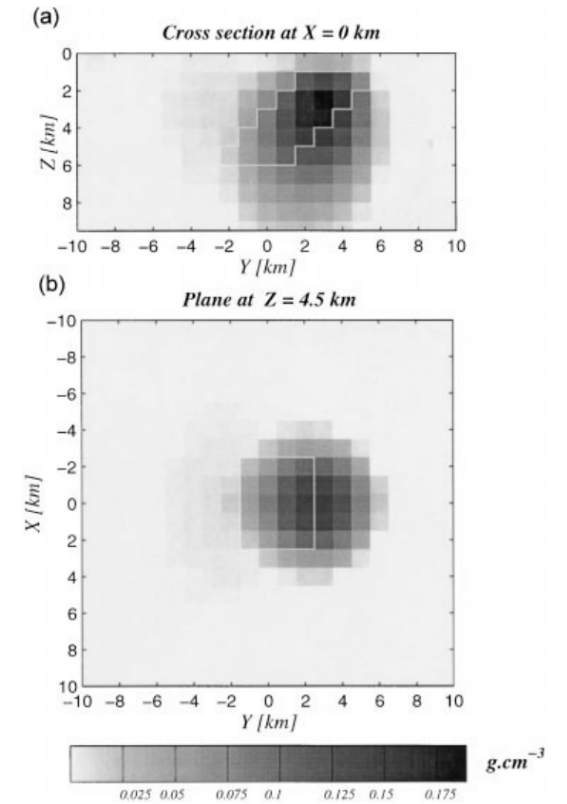
Forward



(Boulanger, O., & Chouteau, M., 2001)

Inverse

Inverse Model



Model-based inversion

- Implicit geomodeling method (GemPy)
- Interpolate a scalar field based on the interface data (surface/orientation/fault points)
- Lower dimensional parametric representation
- Query lithology at any location
- Suitable for structural geological modeling

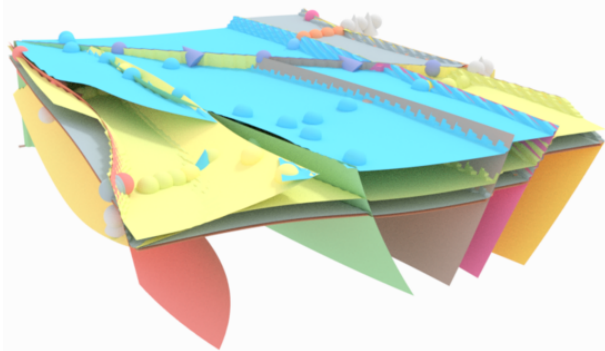


Fig. 1: Example of a 3D geological model in GemPy (de la Varga et. al., 2019)

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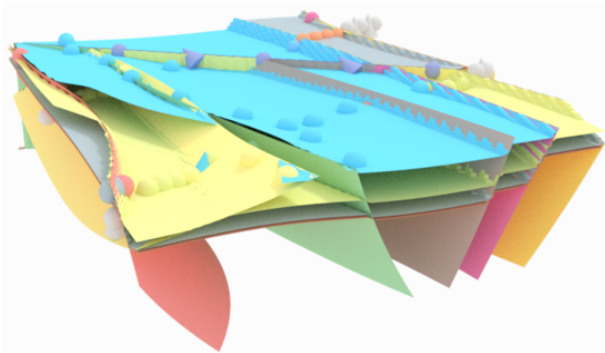


Fig. 1: Example of a 3D geological model in GemPy (de la Varga et. al., 2019)

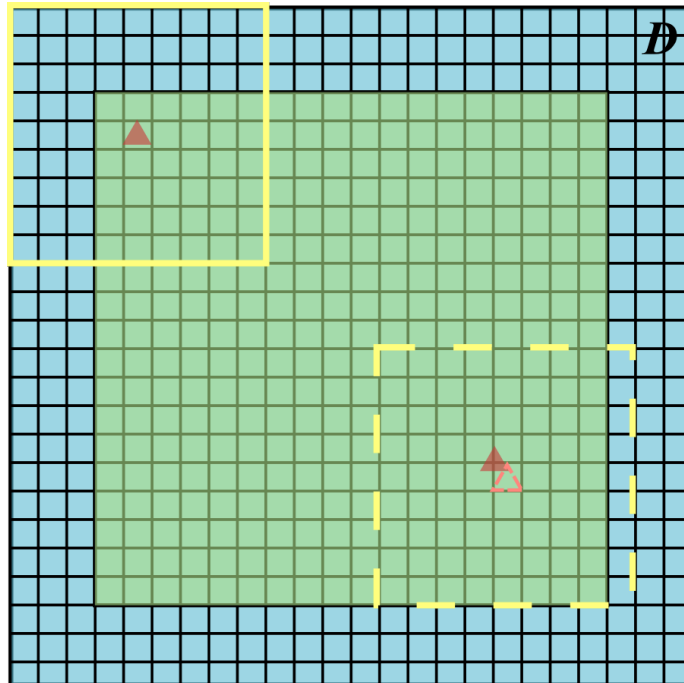
Kernel method - Gravity simulation

- Receiver - based
 - Stochastic modeling
 - Implicit methods
 - Sparse data / Highly correlated data
 - Large scale model
 - Tensor grid - analytical solution of gravity
 - Pre-compute T_z , integration with differentiable modeling framework [1]
-
- Gravity $g_z = T_z \rho$
 - Sensitivity matrix

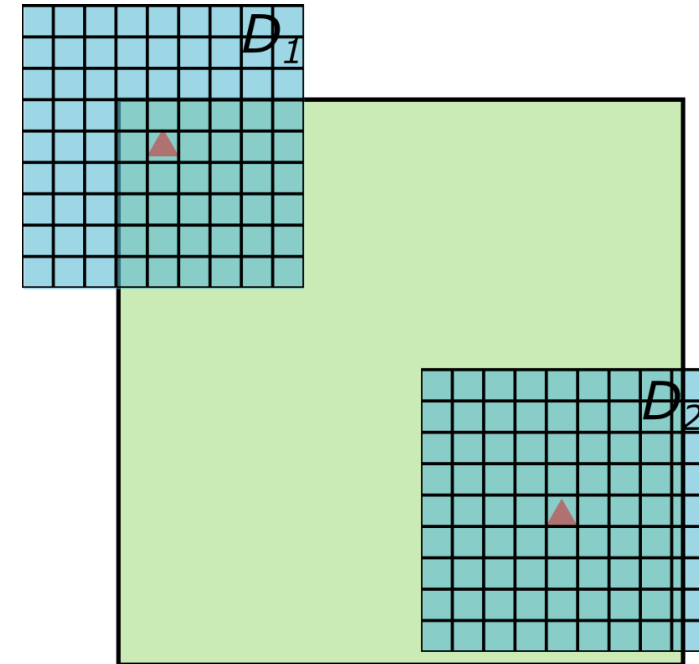
$$T_z = -|||x \ln(y + r) + y \ln(x + r) - z \arctan\left(\frac{xy}{zr}\right) \Big|_{x_1}^{x_2} \Big|_{y_1}^{y_2} \Big|_{z_1}^{z_2}$$

Forward gravity

Spatial convolution scheme

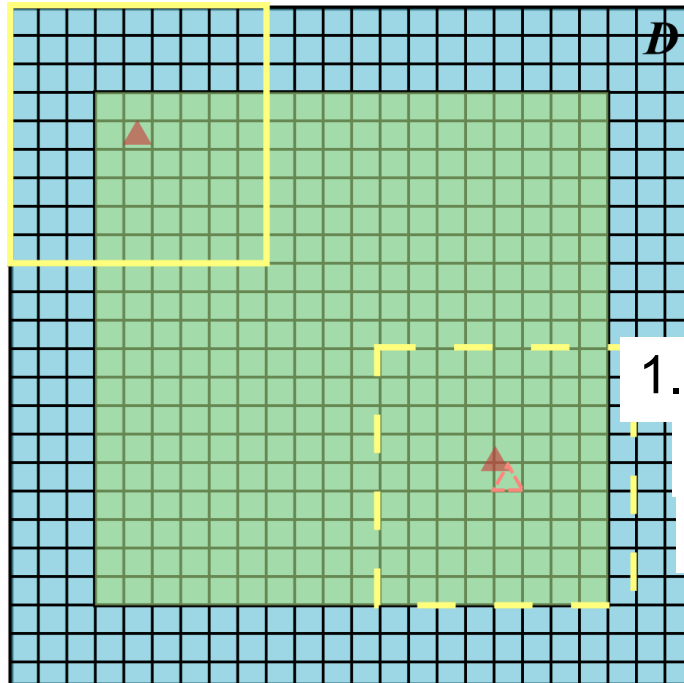


Kernel method



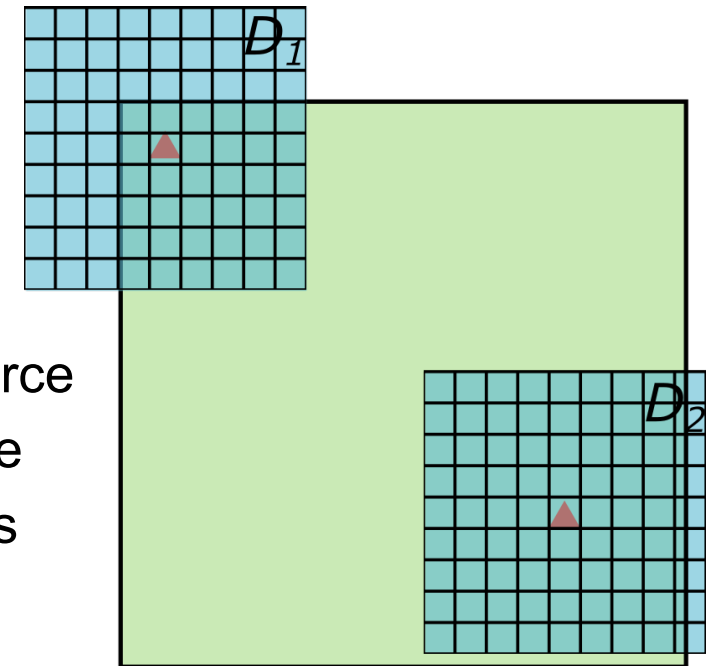
Forward gravity

Spatial convolution scheme



1. Save computation resource
2. Integrate padding zone
3. Eliminate slicing errors

Kernel method



Kernel method - refinement strategy

Optimal grid design without knowing the underlying geometry is not straightforward before evaluating the model.

- Tensor grid allows a parametric representation
- Refine based on the sensitivity T_z of the cell to the corresponding receiver
- Average the risk of aliasing
- Use AD to minimize $std(T_z)$

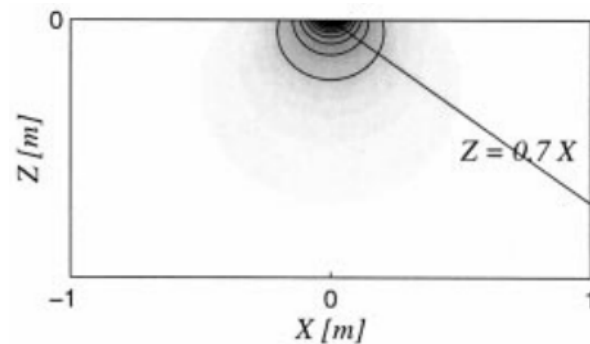
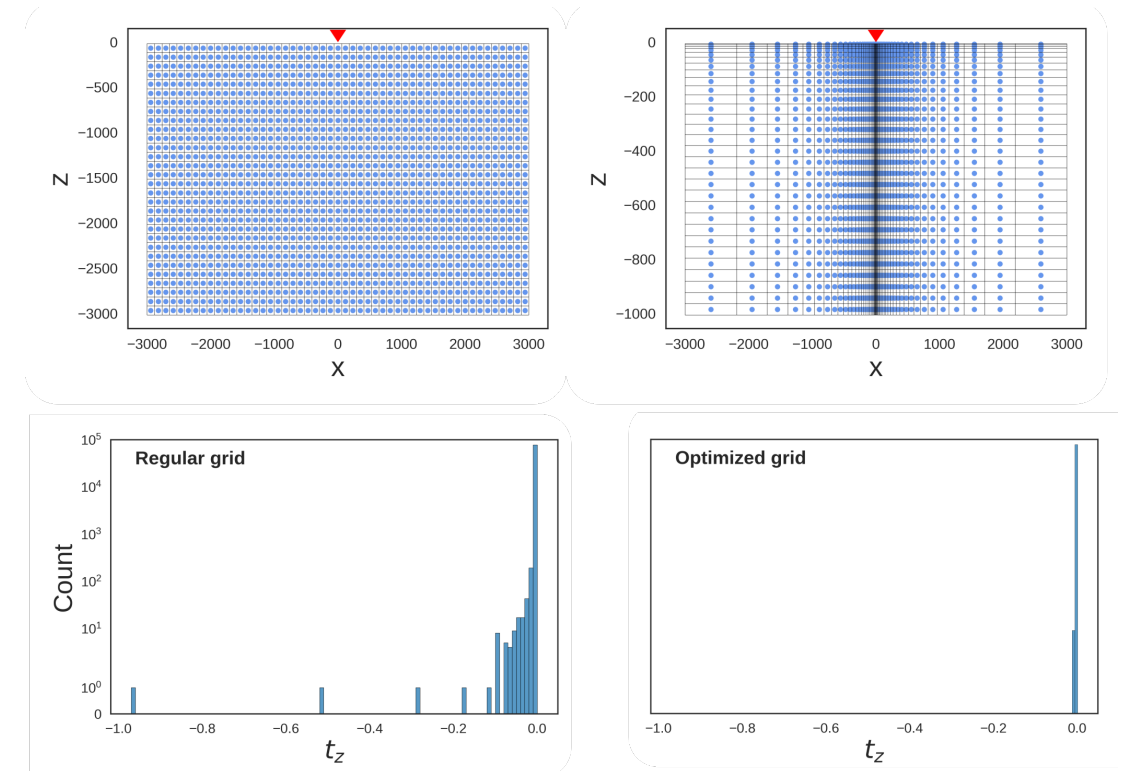


Fig. 2: Normalized sensitivity coefficients (Boulanger, O., & Chouteau, M., 2001)



Examples

Greenstone belt, Western Australia

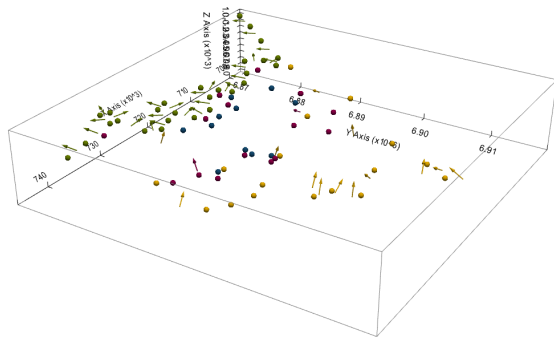


Fig. 3: Input data

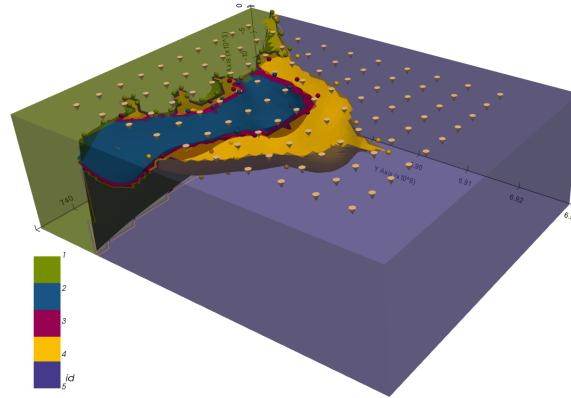


Fig. 4: 3D GemPy Greenstone model

- Archaean greenstone belt in the Youanmi Terrane of the Yilgarn Craton in Western Australia
- Interest for mineral exploration
- 69 surfaces points and 40 orientation points

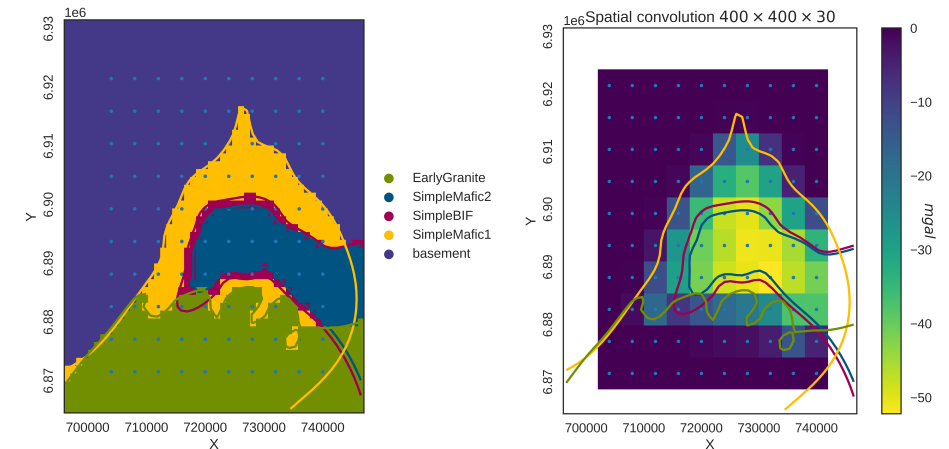


Fig. 5: Lithologies at the surface and gravity forward simulation

Examples

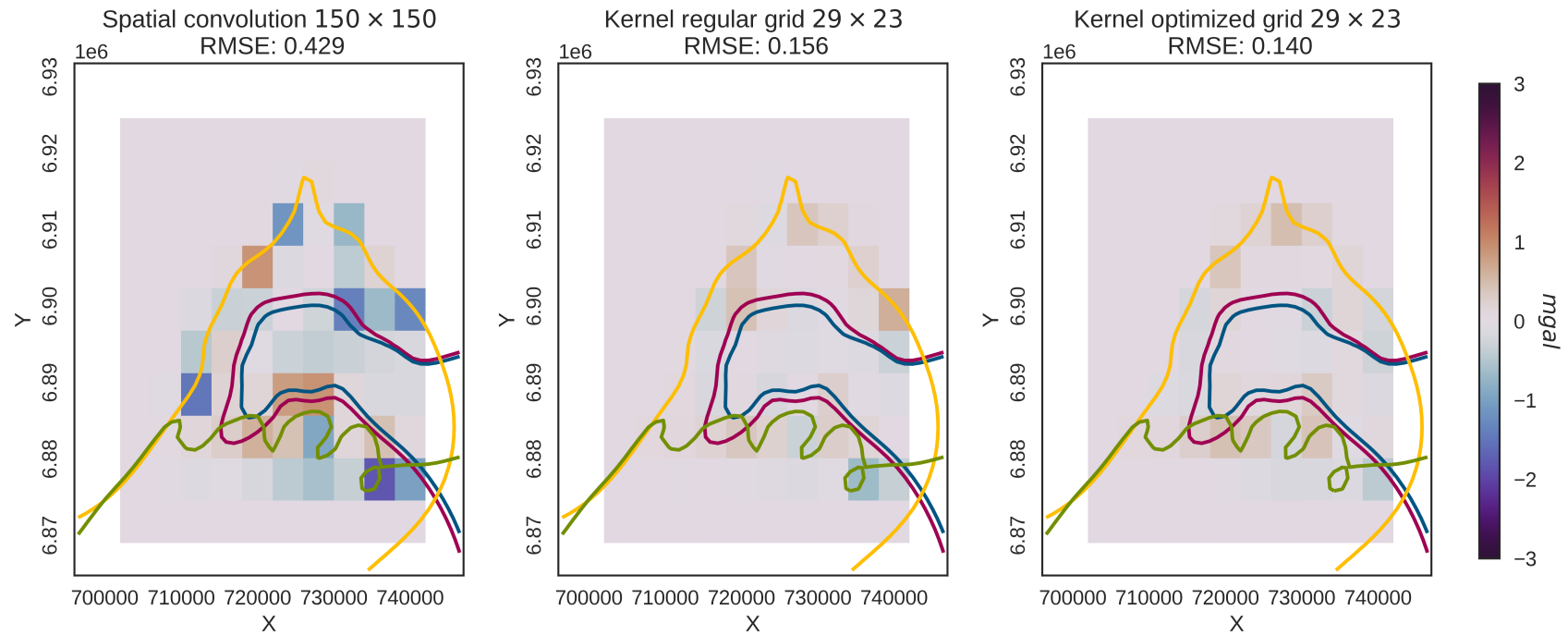


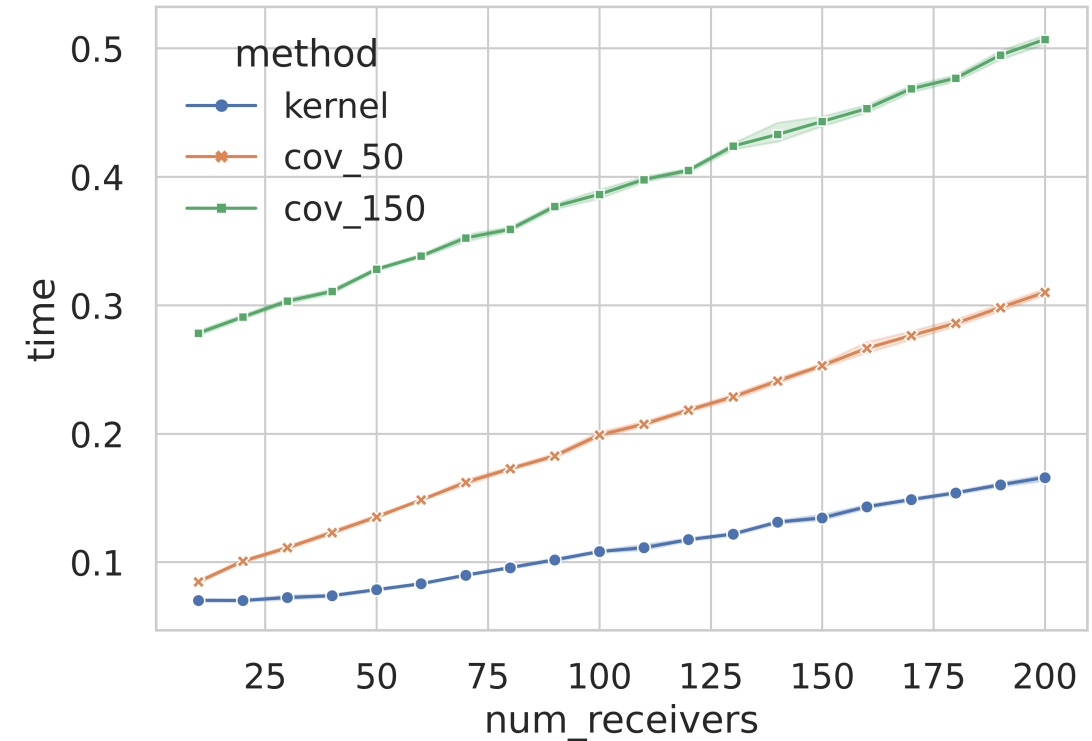
Fig. 6: Gravity forward simulation using spatial convolution scheme and kernel method on centered regular grid and optimized grid

Conclusion

- Kernel method
 - Individual grid for receivers
 - Less error than the spatial convolution scheme
 - Time complexity of $\mathcal{O}(1)$ compared to spatial convolution scheme $\mathcal{O}(n)$ (although memory overhead still exists)
- Optimized grid
 - Average risk of aliasing due to low resolution
 - Optimization only need to be run once for a given grid setup

Future work

- Memory scalability with the number of data
- Inversion

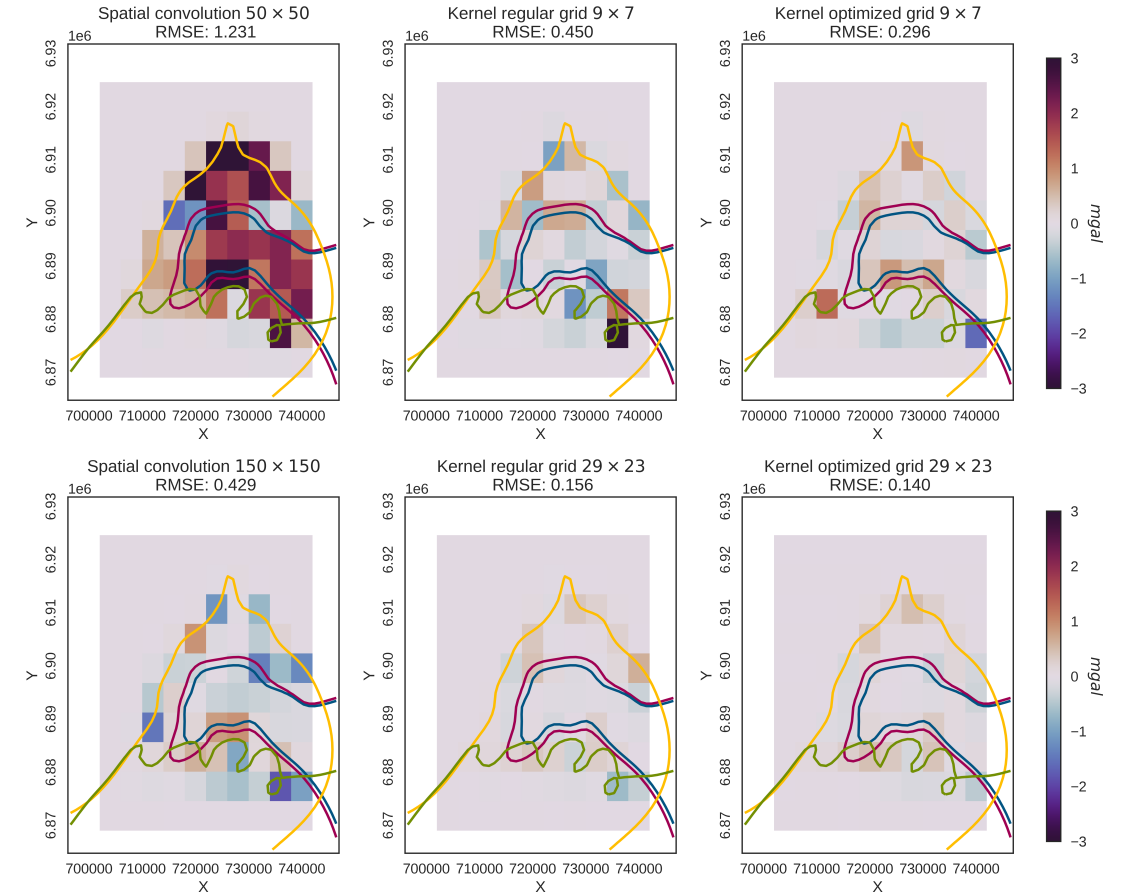


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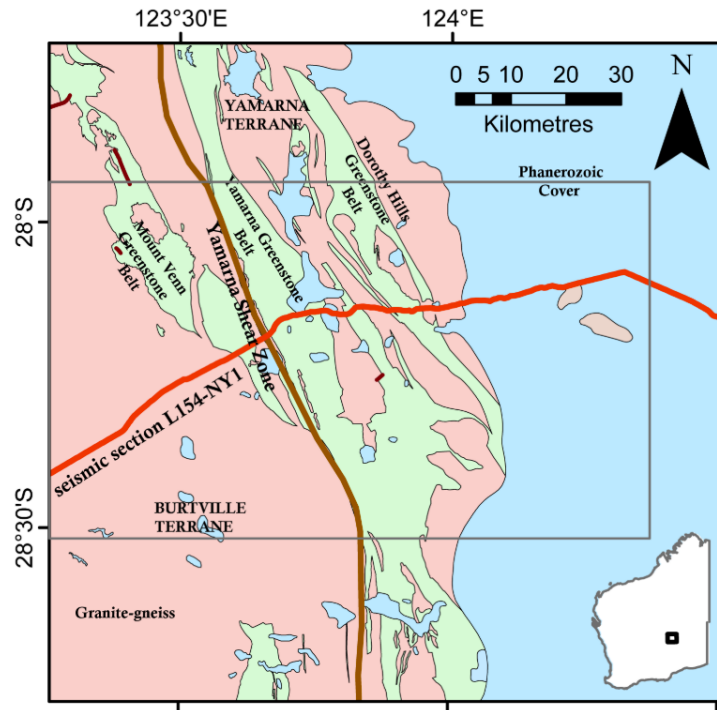
Thank you for your attention

Any questions?



Additional Material

Yamarna Terrane, Western Australia Geographical Map



Gravity anomaly

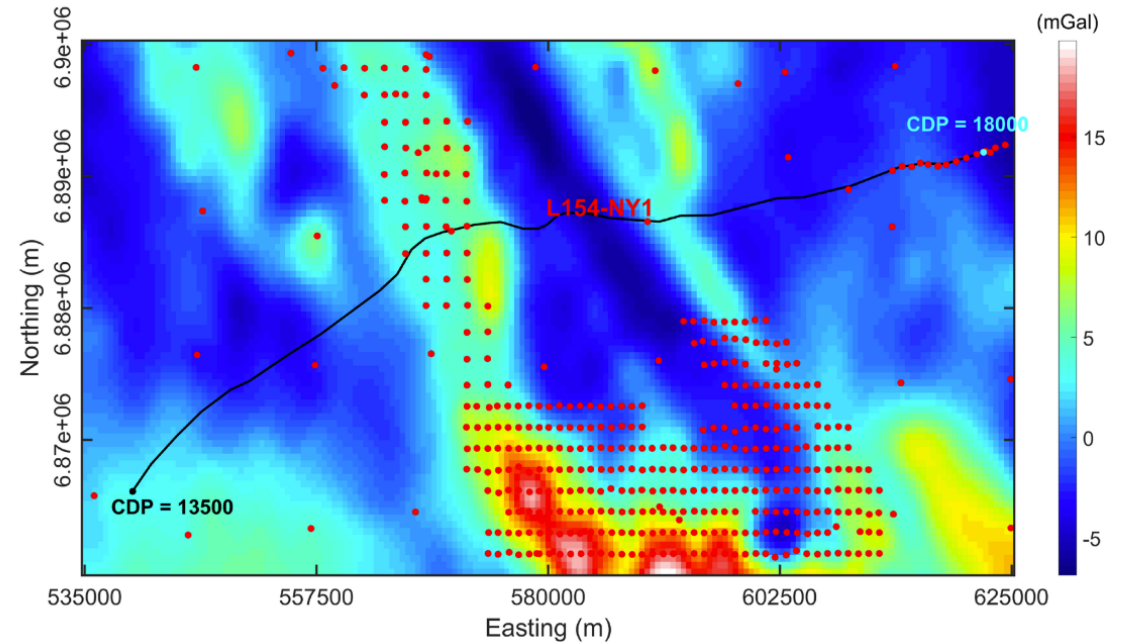


Fig. 8: Rashidifard et. al. (2021)

Synthetic spherical geobody

$$Z = (x^2 + y^2 + z^2), \rho(Z) = \begin{cases} \rho_1, & Z < Z_0 \\ \rho_2, & Z > Z_0 \end{cases}$$

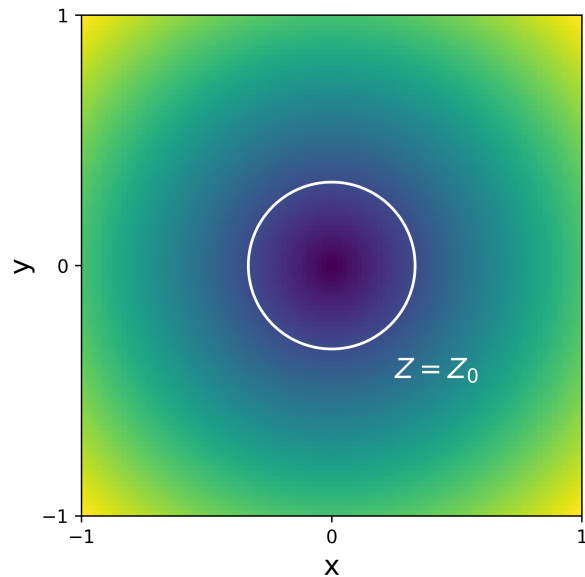


Fig. 9: 2D scalar field of a spherical geobody

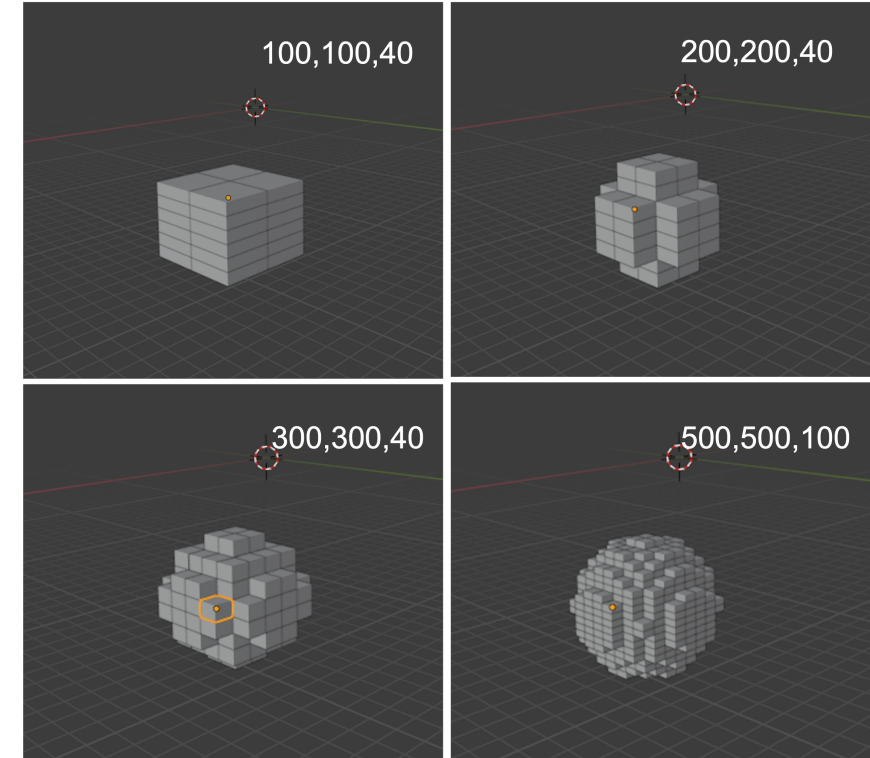


Fig. 10: 3D grid representation of sphere

Synthetic spherical geobody

sphere position: $x = 5000$, $z = 870$; radius = 100

Synthetic spherical geobody

sphere position: $x = 4100$, $z = 110$; radius = 100