

2.5d inversion recipes

ER inversion

This process is repeated for each experiment (i.e. each source s and all its receivers). Let the 2.5d weights for source s be $\{k, \omega\}$ and stored in disk,

1. compute synthetic electric potentials $\{\tilde{u}_i\}$, data d and error $e = d - d^o$,
 - (a) retrieve from memory k and ω ,
 - (b) choose $k_i \in k$ and build L_i with the right boundary conditions for that k_i ,

$$\begin{aligned} L^i &\approx -\nabla \cdot \sigma \nabla, \\ L_i &= L^i + k_i^2 \sigma, \end{aligned}$$

- (c) solve $L_i \tilde{u}_i = \frac{s}{2}$ for \tilde{u}_i and store,
 - (d) after all k has been used,

$$u = \sum_i \omega_i \tilde{u}_i \rightarrow d = Mu \rightarrow e = d - d^o.$$

2. correct error amplitude by its variance,

$$e \leftarrow \frac{e}{1 + \text{std}(d^o)}$$

3. compute g ,

- (a) choose $k_i \in k$,
 - (b) build L_i and $S_i = -((\nabla_\sigma L^i) \tilde{u}_i)^\top - k_i^2 \text{diag}(\tilde{u}_i)^\top$
 - (c) solve $L_i^\top a_i = M^\top e$ for a_i
 - (d) compute and store $g_i = S_i a_i$,
 - (e) after all k has been used,

$$g = \frac{2}{\pi} \sum \omega_i g_i.$$

4. On g filter out high spatial-frequencies and normalize by largest amplitude,

5. compute $\sigma_{err} = \sigma - \sigma_{ref}$, filter out high spatial-frequencies and normalize by largest amplitude,
6. $g \leftarrow g + \beta \sigma_{err}$ and normalize g by largest amplitude,
7. compute step-size α_s using Pica and set $\Delta\sigma_s = -\alpha_s g$
8. store $\Delta\sigma_s$ and α_s .

After all sources have their update direction,

$$\begin{aligned}\alpha &= \frac{1}{n_s} \sum \alpha_s \\ \Delta\sigma &= \sum \Delta\sigma_s \\ \Delta\sigma_{dc} &= \alpha \frac{\Delta\sigma}{\max |\Delta\sigma|}.\end{aligned}$$

The conductivity is updated as,

$$\sigma \leftarrow \sigma \odot \exp\{\sigma \odot \Delta\sigma_{dc}\}.$$

GPR inversion ε

This process is repeated for each experiment (i.e. each source s and all its receivers),

1. compute synthetic data d and error $e = d - d^o$,
2. **how to treat noise in e ?**
3. compute g_ε and apply Kurzmann preconditioner,
4. on $g_{w,\varepsilon}$ filter out high spatial-frequencies and normalize by largest amplitude,
5. compute step-size α using Pica and set $\Delta\varepsilon_s = -\alpha g_\varepsilon$
6. store $\Delta\varepsilon_s$,
7. invert for source wavelet with a **Wiener filter**.

After all sources have their update direction,

$$\Delta\varepsilon = \frac{1}{n_s} \sum \Delta\varepsilon_s.$$

The permittivity is updated as,

$$\varepsilon \leftarrow \varepsilon \odot \exp\{\varepsilon \odot \Delta\varepsilon\}.$$

GPR inversion σ

This process is repeated for each experiment (i.e. each source s and all its receivers),

1. compute synthetic data d and error $e = d - d^o$,
2. **how to treat noise in e ?**
3. compute $g_{w,\sigma}$ and apply Kurzmann preconditioner,
4. on $g_{w,\sigma}$ filter out high spatial-frequencies and normalize by largest amplitude,
5. compute step-size α using Pica and set $\Delta\sigma_s = -\alpha g_{w,\sigma}$
6. store $\Delta\sigma_s$,
7. invert for source wavelet with **Wiener filter**.

After all sources have their update direction,

$$\Delta\sigma_w = \frac{1}{n_s} \sum \Delta\sigma_s.$$

The conductivity is updated as,

$$\sigma \leftarrow \sigma \odot \exp\{\sigma \odot \Delta\sigma_w\}.$$

GPR+ER inversion σ

The update direction is

$$\Delta\sigma \leftarrow \frac{1}{2}(\Delta\sigma_w + \Delta\sigma_{dc})$$

and the update for *both* the GPR and ER is

$$\sigma \leftarrow \sigma \odot \exp\{\sigma \odot \Delta\sigma\}.$$