Differential programming

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Outline

- 1 Non-linear optimization and SABR model
 - Levenberg-Marquad
 - SABR model

2 Diffential programming for detectors

Levenberg-Marquad algorithm - is an optimization algorithm which is basically combination of two objective minimization algorithms: **Gauss-Newton** and **Gradient Descent** It is very very well suited for the optimization problemsswhere model is **non-linear** in its parameters

Non-linear least squares

$$\hat{y}(t; oldsymbol{p})$$
 - fitted model $t \in \mathbb{R}$ - independent variable $\{t_i, y_i\}_{i=1}^m$ - observed data points W_{ij} - inverse covariance matrix $oldsymbol{p} \in \mathbb{R}^n$ model parameters

Optimized objective is nothing else but chi-squared statistic

$$\chi^{2}(\boldsymbol{p}) = \sum_{i=1}^{m} \left[\frac{y(t_{i}) - \hat{y}(t_{i}; \boldsymbol{p})}{\sigma_{y_{i}}} \right]^{2}$$
$$= (\boldsymbol{y} - \hat{\boldsymbol{y}}(\boldsymbol{p}))^{\top} \boldsymbol{W}(\boldsymbol{y} - \hat{\boldsymbol{y}}(\boldsymbol{p}))$$

Gradient Descent and Jacobian update

Gradient Descent step update

$$\frac{\partial}{\partial \boldsymbol{p}} \chi^2 = -2(\boldsymbol{y} - \hat{\boldsymbol{y}})^\top \boldsymbol{W} \boldsymbol{J}$$
$$\boldsymbol{h}_{\mathrm{gd}} = \alpha \boldsymbol{J}^\top \boldsymbol{W} (\boldsymbol{y} - \hat{\boldsymbol{y}})$$

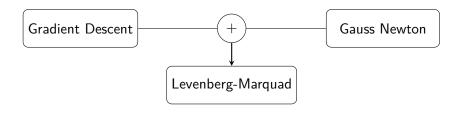
Gauss-Newton step update

$$\hat{\mathbf{y}}(\mathbf{p} + \mathbf{h}) \approx \hat{\mathbf{y}}(\mathbf{p}) + \left[\frac{\partial \hat{\mathbf{y}}}{\partial \mathbf{p}}\right] \mathbf{h} = \hat{\mathbf{y}} + \mathbf{J}\mathbf{h}$$

$$\frac{\partial}{\partial \mathbf{h}} \chi^{2}(\mathbf{p} + \mathbf{h}) \approx -2(\mathbf{y} - \hat{\mathbf{y}})^{\top} \mathbf{W} \mathbf{J} + 2\mathbf{h}^{\top} \mathbf{J}^{\top} \mathbf{W} \mathbf{J}$$

$$\left[\mathbf{J}^{\top} \mathbf{W} \mathbf{J}\right] \mathbf{h}_{gn} = \mathbf{J}^{\top} \mathbf{W} (\mathbf{y} - \hat{\mathbf{y}})$$

Levenberg-Marquad update



$$\left[\mathbf{J}^{\top} \mathbf{W} \mathbf{J} + \lambda \operatorname{diag} \left(\mathbf{J}^{\top} \mathbf{W} \mathbf{J} \right) \right] \mathbf{h}_{\mathrm{lm}} = \mathbf{J}^{\top} \mathbf{W} (\mathbf{y} - \hat{\mathbf{y}})$$

 λ is dumping parameter

Motivation for the model

Model Definition

Exact solution