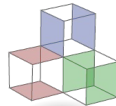


Richards equation

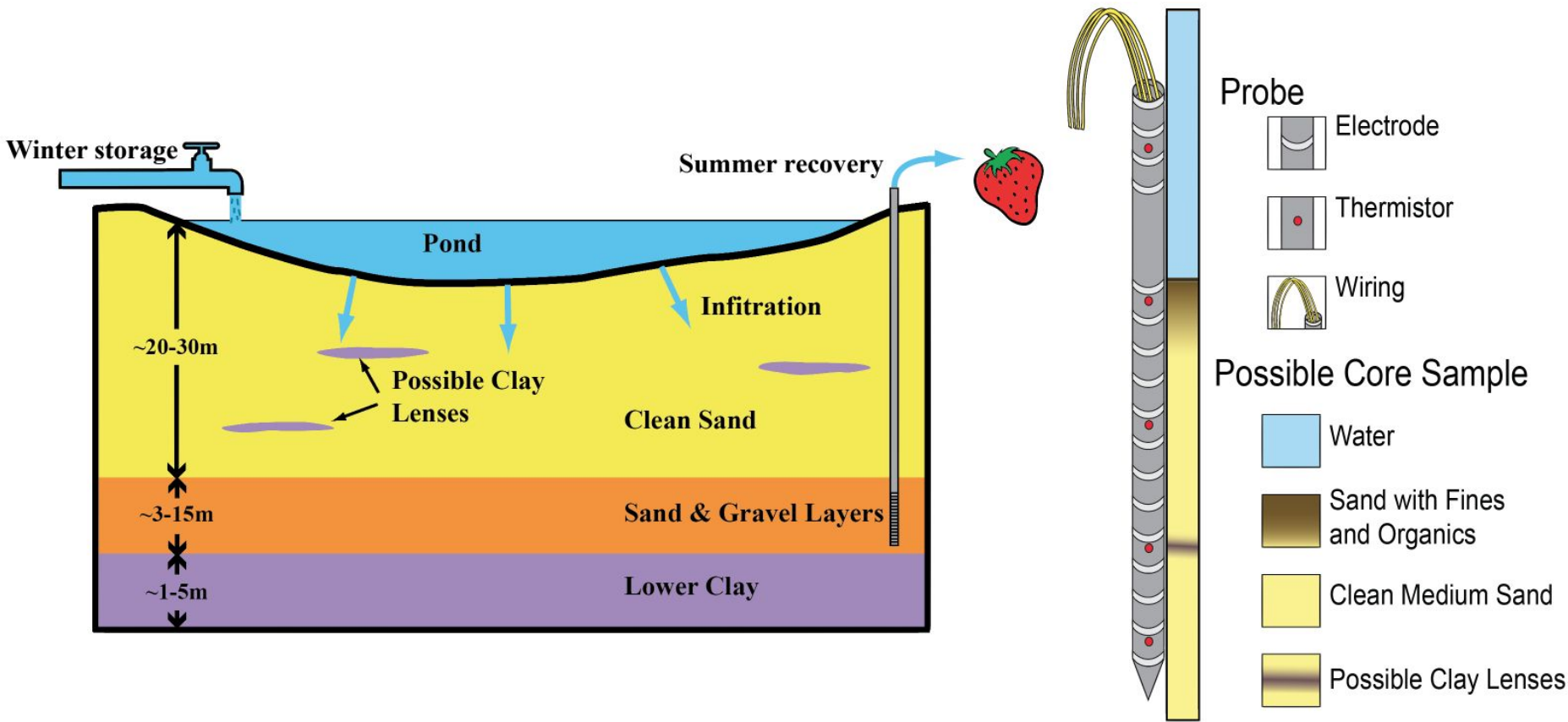
Forward modeling and inversions

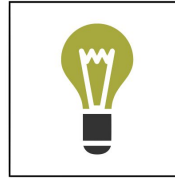
Rowan Cockett

Nov 22, 2016



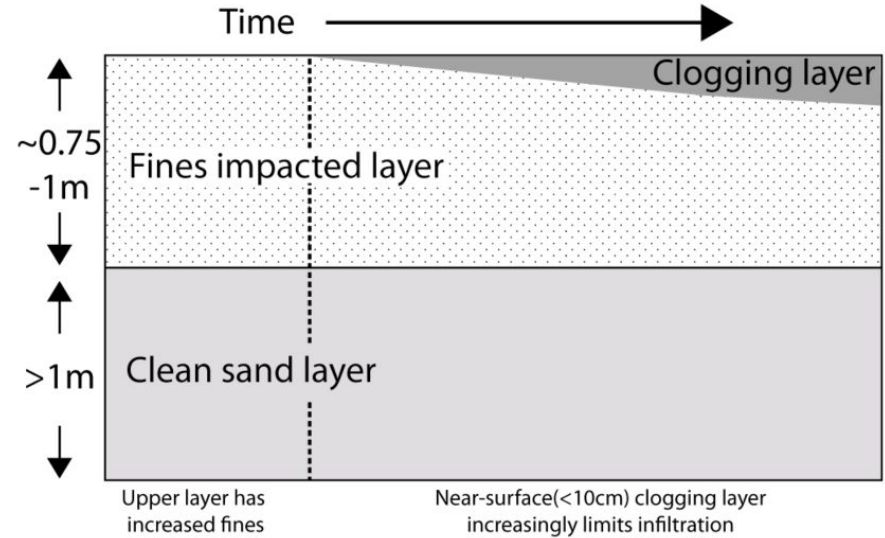
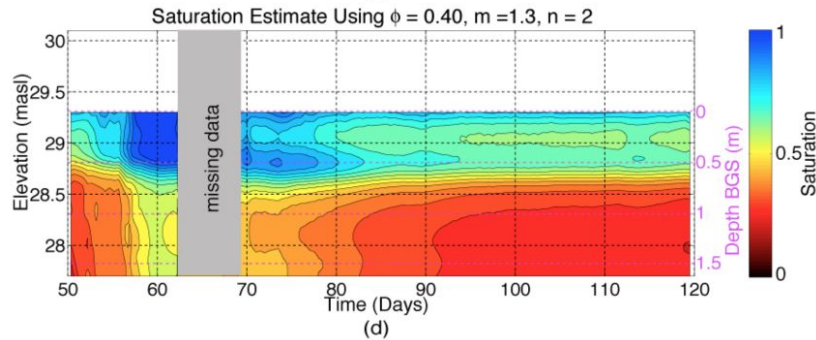
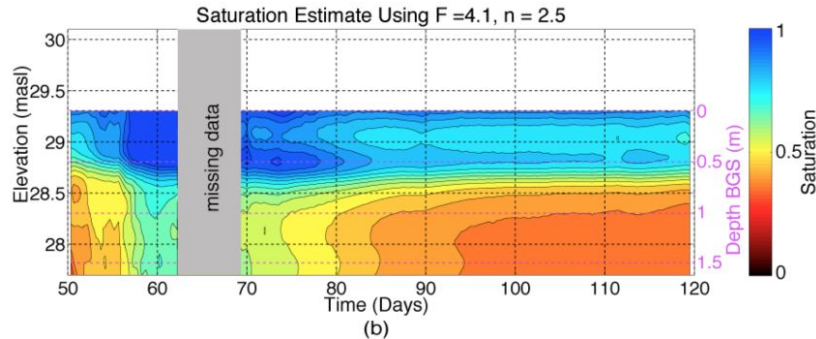
simpeg





Predict & Decide

Data → Physics → Interpret → Decide

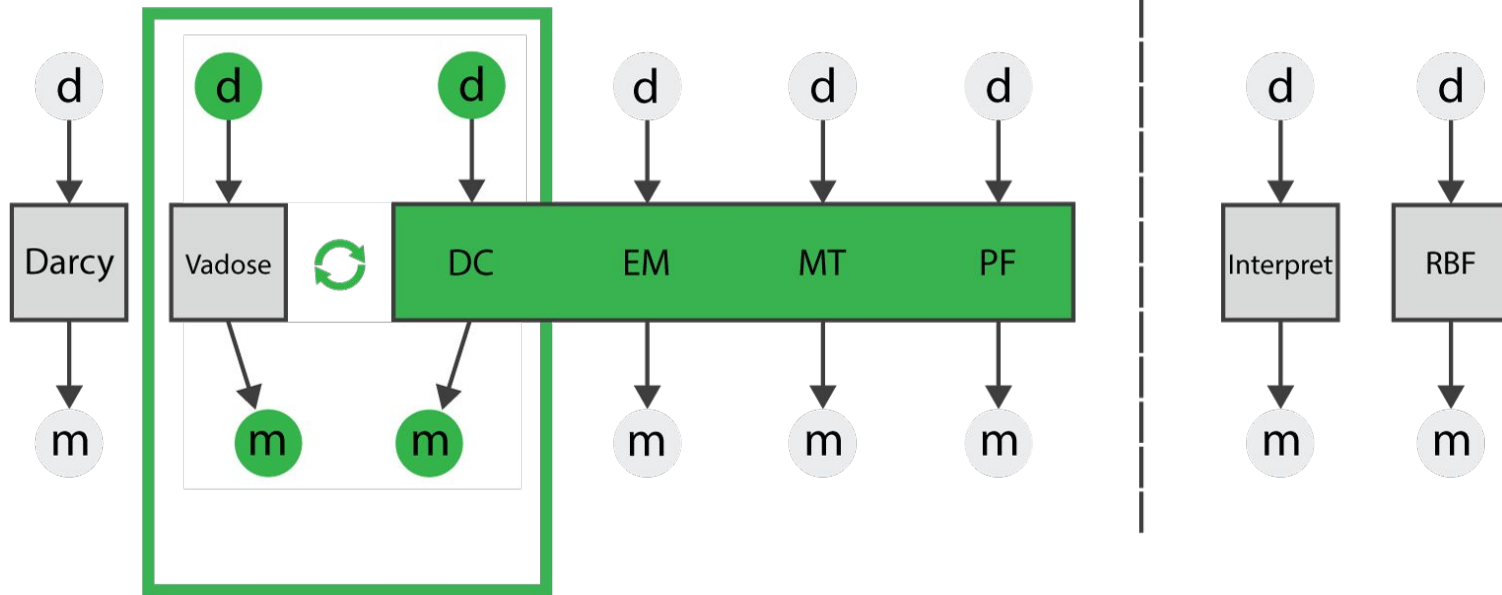


Hydrogeology

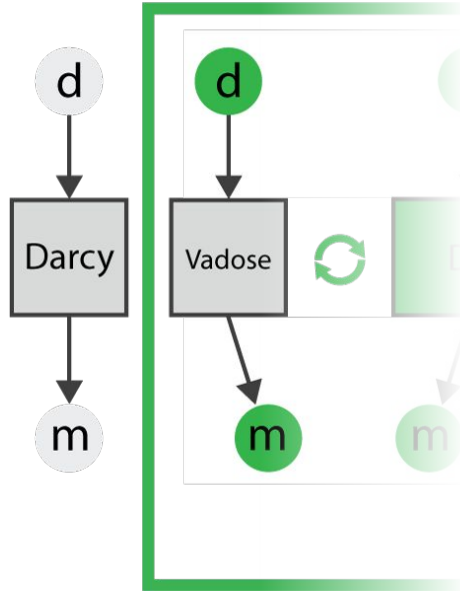
Geophysics

Geology

Inversion



Hydrogeology



Richards equation:

$$\frac{\partial \theta(\psi)}{\partial t} - \underbrace{\nabla \cdot k(\psi) \nabla \psi}_{\text{Diffusion}} - \underbrace{\frac{\partial k(\psi)}{\partial z}}_{\text{Advection}} = 0 \quad \psi \in \Omega$$

Empirical equation (e.g. van Genuchten-Mualem):

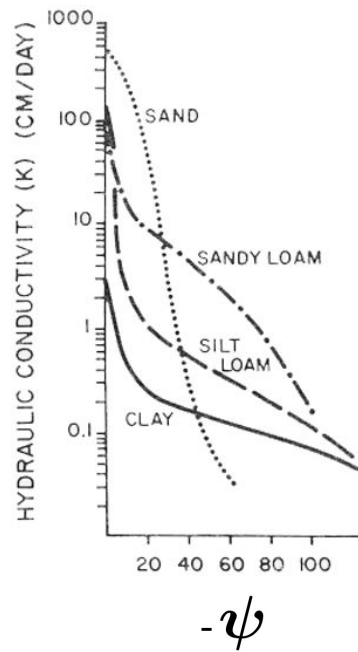
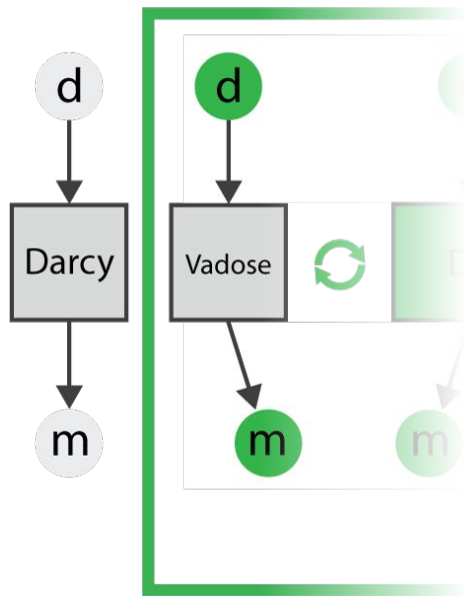
$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha\psi|^n)^m} & \psi < 0 \\ \theta_s & \psi \geq 0 \end{cases}$$

$$k(\psi) = \begin{cases} K_s \theta_e(\psi)^l (1 - (1 - \theta_e(\psi)^{-m})^m)^2 & \psi < 0 \\ K_s & \psi \geq 0 \end{cases}$$

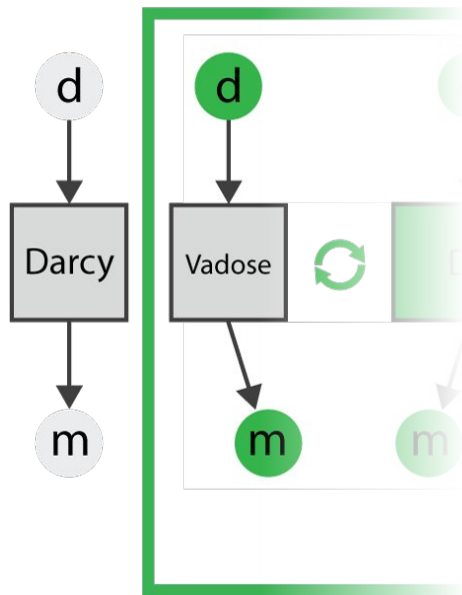
$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 - \frac{1}{n}, \quad n > 1$$

Hydrogeology

Inversion



Hydrogeology



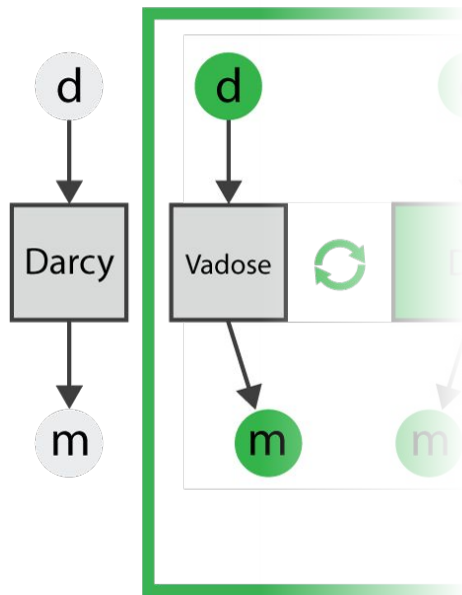
$$\frac{\partial \theta(\psi)}{\partial t} - \nabla \cdot k(\psi) \nabla \psi - \frac{\partial k(\psi)}{\partial z} = 0 \quad \psi \in \Omega$$

$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha\psi|^n)^m} & \psi < 0 \\ \theta_s & \psi \geq 0 \end{cases}$$

$$k(\psi) = \begin{cases} K_s \theta_e(\psi)^l (1 - (1 - \theta_e(\psi)^{-m})^m)^2 & \psi < 0 \\ K_s & \psi \geq 0 \end{cases}$$

$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 - \frac{1}{n}, \quad n > 1$$

Hydrogeology



$$\frac{\partial \theta(\psi)}{\partial t} - \nabla \cdot k(\psi) \nabla \psi - \frac{\partial k(\psi)}{\partial z} = 0 \quad \psi \in \Omega$$

Heterogeneous?

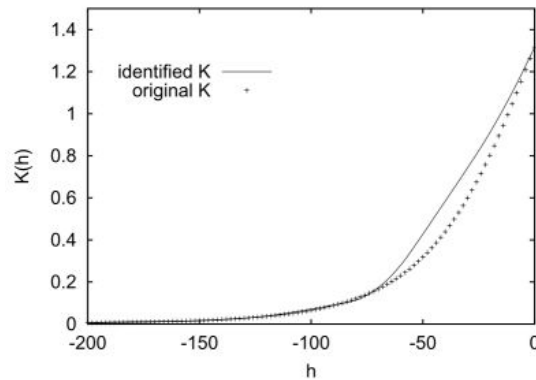
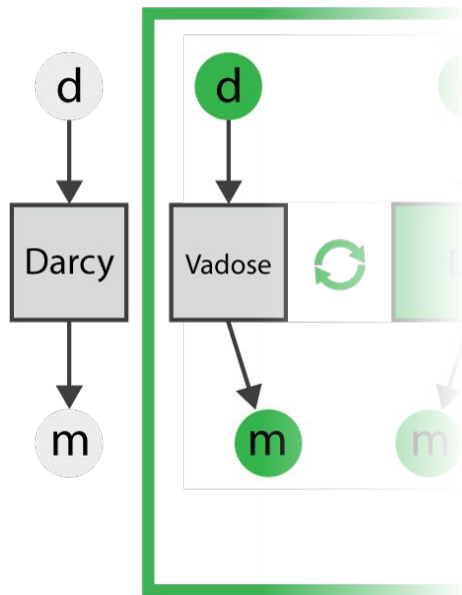
$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha \psi|^n)^m} & \psi < 0 \\ \theta_s & \psi \geq 0 \end{cases}$$

$$k(\psi) = \begin{cases} \frac{K_s \theta_e(\psi)^l}{K_s} (1 - (1 - \theta_e(\psi)^{-m})^m)^2 & \psi < 0 \\ \frac{K_s}{K_s} & \psi \geq 0 \end{cases}$$

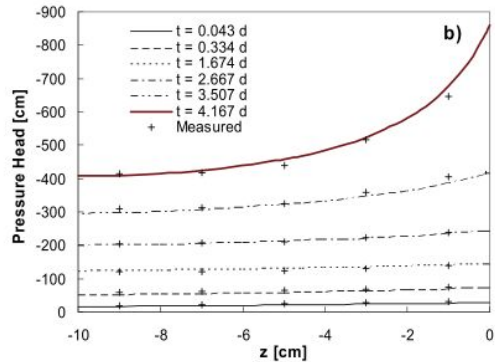
$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 - \frac{1}{n}, \quad n > 1$$

Anisotropic?

Hydrogeology



Bitterlich and Knabner, 2002



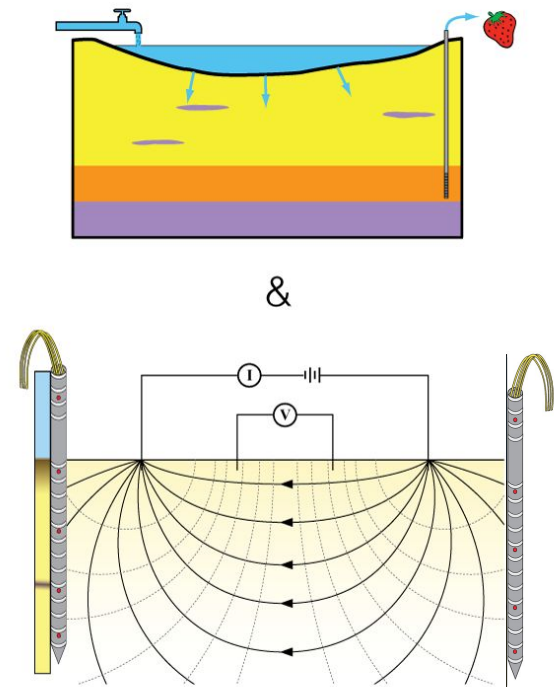
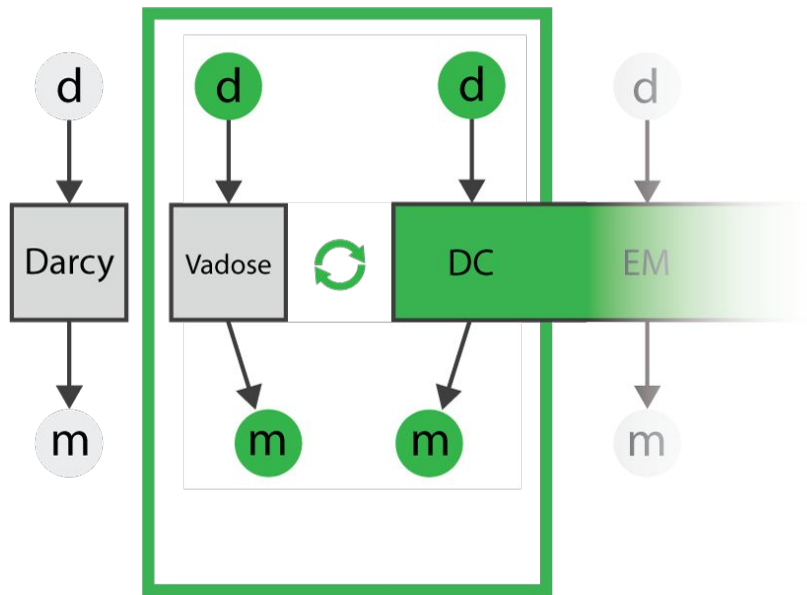
Šimůnek, van Genuchten, and Senja, 2012

Inversion

Hydrogeology

Geophysics

Inversion



Number of parameters:



100s → 1000s

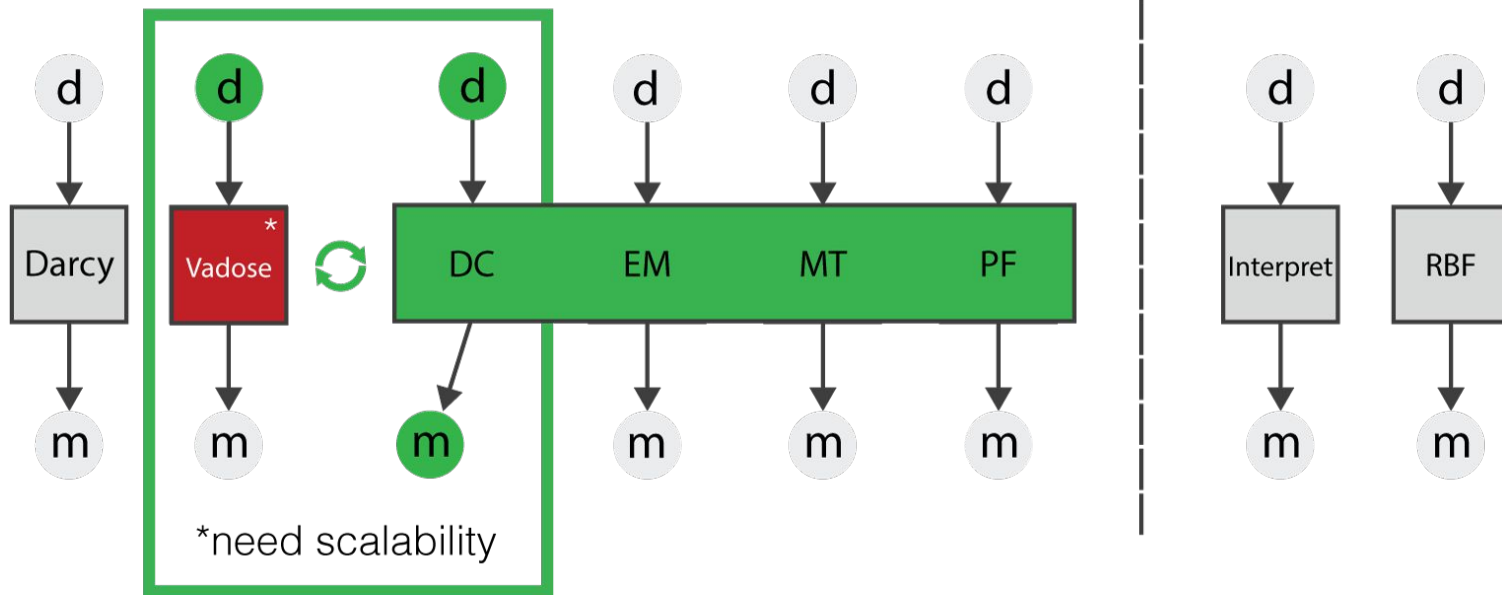
100,000s → 10,000,000s

Hydrogeology

Geophysics

Geology

Inversion

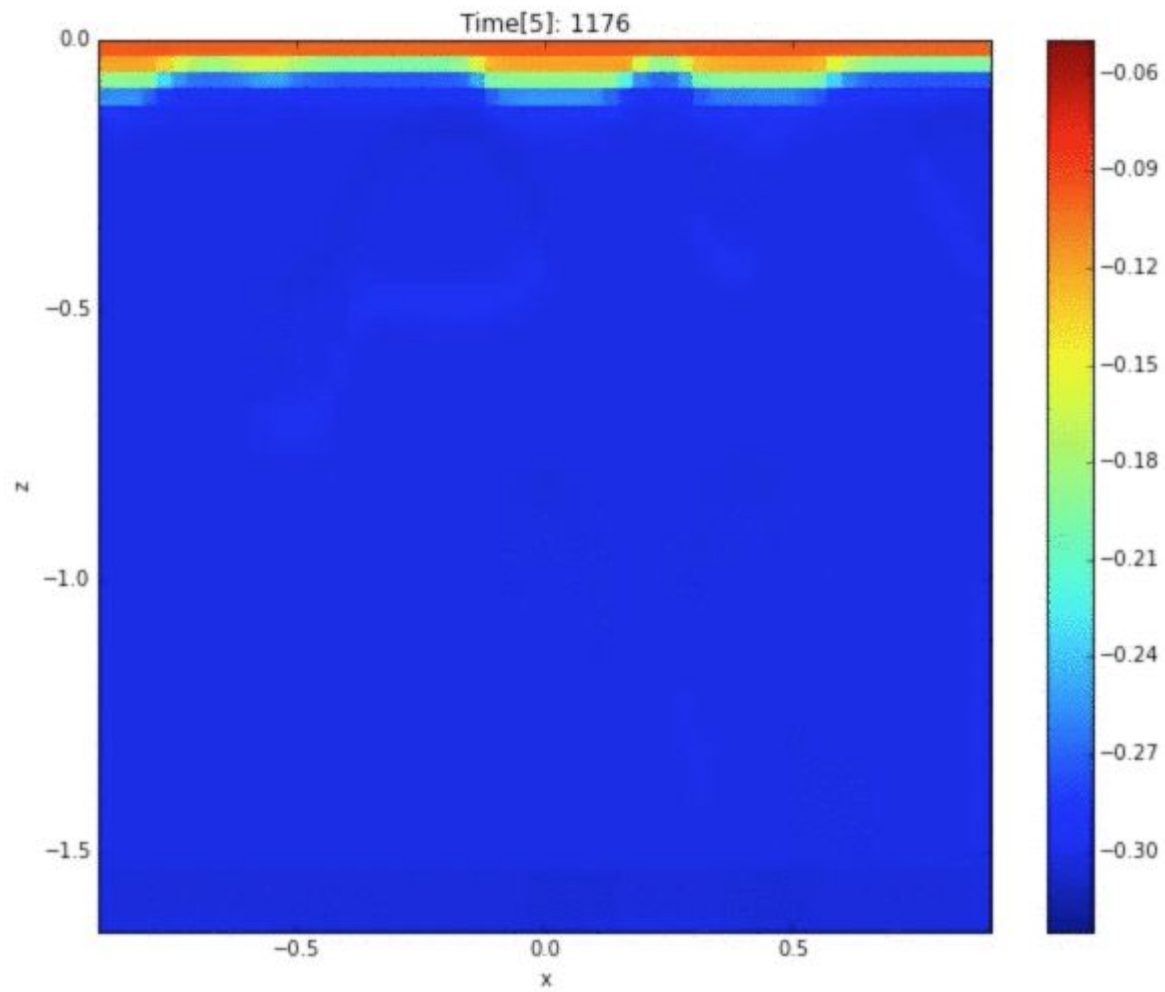


Scalable Sensitivity?

- 1) Investigate forward
- 2) Implicit sensitivity
- 3) For example



(Cockett, Heagy & Haber, in prep)



$$\frac{\partial \theta(\psi)}{\partial t} - \underbrace{\nabla \cdot K(\psi) \nabla \psi}_{\text{Diffusion}} - \underbrace{\frac{\partial K(\psi)}{\partial z}}_{\text{Advection}} = 0$$

$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha\psi|^n)^m} & \psi < 0 \\ \theta_s & \psi \geq 0 \end{cases}$$

$$k(\psi) = \begin{cases} K_s \theta_e(\psi)^l (1 - (1 - \theta_e(\psi)^{-m})^m)^2 & \psi < 0 \\ K_s & \psi \geq 0 \end{cases}$$

$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 - \frac{1}{n}, \quad n > 1$$

`class SimPEG.FLOW.Richards.RichardsProblem.Richards`

Bases: `SimPEG.Problem.BaseTimeProblem`

docstring for RichardsProblem

Required

- Parameters:
- **boundaryConditions** (`Array`) – boundary conditions, <type 'float'>, <type 'int'> with shape (*)
 - **Solver** (`Property`) – Numerical Solver, Default: `Newton`
 - **mapping** (`Property`) – the mapping
 - **tolRootFinder** (`Float`) – Maximum iteration for rootFinder, Default: 0.0001
 - **maxIterRootFinder** (`Integer`) – Maximum iteration for rootFinder, Default: 30
 - **initialConditions** (`Array`) – boundary conditions, <type 'float'>, <type 'int'> with shape (*)
 - **debug** (`Bool`) – Show all messages, a boolean
 - **model** (`Model`) – Inversion model, a numpy array with shape (*)
 - **doNewton** (`Bool`) – Do a Newton iteration vs. a Picard iteration, Default: False
 - **method** (`StringChoice`) – Formulation used, choices: 'mixed' or 'head'

```
class RichardsProblem(Problem.BaseTimeProblem):
```

```
    """RichardsProblem"""
```

```
    mapping = properties.Property("the mapping")
```

```
    boundaryConditions = properties.Array("boundary conditions.")
```

```
    initialConditions = properties.Array("boundary conditions.")
```

```
    surveyPair = RichardsSurvey
```

```
    mapPair = RichardsMap
```

```
    debug = properties.Bool("Show all messages")
```

```
    Solver = properties.Property("Numerical Solver", default=lambda: Solver)
```

```
    solverOpts = {}
```

```
    method = properties.StringChoice(
```

```
        "Formulation used, See notes in Celia et al., 1990.",
```

```
        choices=['mixed', 'head']
```

```
)
```

```
    doNewton = properties.Bool(
```

```
        "Do a Newton iteration vs. a Picard iteration ",
```

```
        default=False
```

```
)
```

```
    maxIterRootFinder = properties.Integer(
```

```
        "Maximum iterations for rootFinder iteration.",
```

```
        default=30
```

```
)
```

```
    tolRootFinder = properties.Float(
```

```
        "Maximum iterations for rootFinder iteration.",
```

```
        default=1e-4
```


```
)
```

Getting rid of mapping

```
prob = Richards.RichardsProblem(  
    mesh,  
    mapping=E,  
    timeSteps=[(40, 3), (60, 3)],  
    Solver=Solver,  
    boundaryConditions=bc,  
    initialConditions=h,  
    doNewton=False,  
    method='mixed',  
    tolRootFinder=1e-6,  
    debug=False  
)
```

```
class RichardsMap(object):  
    """docstring for RichardsMap"""  
  
    mesh = None #: SimPEG mesh  
  
    @property  
    def thetaModel(self):  
        """Model for moisture content"""  
        return self._thetaModel  
  
    @property  
    def kModel(self):  
        """Model for hydraulic conductivity"""  
        return self._kModel  
  
    def __init__(self, mesh, thetaModel, kModel):  
        self.mesh = mesh  
        assert isinstance(thetaModel, NonLinearMap)  
        assert isinstance(kModel, NonLinearMap)  
  
        self._thetaModel = thetaModel  
        self._kModel = kModel  
  
    def theta(self, u, m):  
        return self.thetaModel(u, m)  
  
    def thetaDerivM(self, u, m):  
        return self.thetaModel.derivM(u, m)  
  
    def thetaDerivU(self, u, m):  
        return self.thetaModel.derivU(u, m)
```


Getting rid of mapping

```
prob = Richards.RichardsProblem(  
  mesh,  
  mapping=E,   
  timeSteps=[(40, 3), (60, 3)],  
  Solver=Solver,  
  boundaryConditions=bc,  
  initialConditions=h,  
  doNewton=False,  
  method='mixed',  
  tolRootFinder=1e-6,  
  debug=False  
)
```

```
kModel = Richards.Empirical.Haverkamp_k(mesh)  
thetaModel = Richards.Empirical.Haverkamp_theta(mesh)  
kModel.KsMap = Maps.ExpMap(np=mesh.nC)]  
  
prob = Richards.RichardsProblem(  
  mesh,  
  { kModel=kModel,  
    thetaModel=thetaModel,  
    timeSteps=[(40, 3), (60, 3)],  
    Solver=Solver,  
    boundaryConditions=bc,  
    initialConditions=h,  
    doNewton=False,  
    method='mixed',  
    tolRootFinder=1e-6,  
    debug=False  
)
```

$$\frac{\partial \theta(\psi)}{\partial t} - \underbrace{\nabla \cdot K(\psi) \nabla \psi}_{\text{Diffusion}} - \underbrace{\frac{\partial K(\psi)}{\partial z}}_{\text{Advection}} = 0$$

```
class NonLinearModel(Props.BaseSimPEG):

    model = Props.Model("model")

    def __init__(self, mesh, **kwargs):
        self.mesh = mesh
        super(NonLinearModel, self).__init__(**kwargs)

    @property
    def nP(self):
        """Number of parameters in the model."""
        return self.mesh.nC

    def derivU(self, u, model):
        self.model = model
        return df_du

    def derivM(self, u, model):
        self.model = model
        return df_dm
```

$$\frac{\partial \theta(\psi)}{\partial t} - \underbrace{\nabla \cdot K(\psi) \nabla \psi}_{\text{Diffusion}} - \underbrace{\frac{\partial K(\psi)}{\partial z}}_{\text{Advection}} = 0$$

```
class BaseThetaModel(NonLinearModel):
    def __call__(self, u, model):
        self.model = model
        f = theta(u)
        return f

class BaseKModel(NonLinearModel):
    def __call__(self, u, model):
        self.model = model
        f = k(u)
        return f
```

$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha\psi|^n)^m} & \psi < 0 \\ \theta_s & \psi \geq 0 \end{cases}$$

$$k(\psi) = \begin{cases} K_s \theta_e(\psi)^l (1 - (1 - \theta_e(\psi)^{-m})^m)^2 & \psi < 0 \\ K_s & \psi \geq 0 \end{cases}$$

$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 - \frac{1}{n}, \quad n > 1$$

$$\frac{\partial \theta(\psi)}{\partial t} - \underbrace{\nabla \cdot K(\psi) \nabla \psi}_{\text{Diffusion}} - \underbrace{\frac{\partial K(\psi)}{\partial z}}_{\text{Advection}} = 0$$

```
class BaseThetaModel(NonLinearModel):
    def __call__(self, u, model):
        self.model = model
        f = theta(u)
        return f

class BaseKModel(NonLinearModel):
    def __call__(self, u, model):
        self.model = model
        f = k(u)
        return f
```

$$\theta(\psi) = \begin{cases} \theta_r + \frac{\theta_s - \theta_r}{(1 + |\alpha\psi|^n)^m} \\ \theta_s \end{cases}$$

$$k(\psi) = \begin{cases} K_s \theta_e(\psi)^l (1 - (1 - \theta_e(\psi))^m) \\ K_s \end{cases}$$

$$\theta_e(\psi) = \frac{\theta(\psi) - \theta_r}{\theta_s - \theta_r}, \quad m = 1 -$$

```
class Haverkamp_k(BaseKModel):
    Ks, KsMap, KsDeriv = Props.Invertible(
        "Saturated hydraulic conductivity",
        default=24.96
    )
    A, AMap, ADeriv = Props.Invertible(
        "fitting parameter",
        default=1.175e+06
    )
    gamma, gammaMap, gammaDeriv = Props.Invertible(
        "fitting parameter",
        default=4.74
    )

    def derivM(self, u, model):
        self.model = model
        return self._derivKs(u) + self._derivA(u) + self._derivGamma(u)
```

Inverting for multiple parameters

```
opts = [  
    ('Ks',  
     dict(KsMap=expmap), 1),  
    ('A',  
     dict(AMap=expmap), 1),  
    ('gamma',  
     dict(gammaMap=expmap), 1),  
    ('Ks-A',  
     dict(KsMap=expmap*wires2.one, AMap=expmap*wires2.two), 2),  
    ('Ks-gamma',  
     dict(KsMap=expmap*wires2.one, gammaMap=expmap*wires2.two), 2),  
    ('A-gamma',  
     dict(AMap=expmap*wires2.one, gammaMap=expmap*wires2.two), 2),  
    ('Ks-A-gamma', dict(  
        KsMap=expmap*wires3.one,  
        AMap=expmap*wires3.one,  
        gammaMap=expmap*wires3.two), 3),  
]
```



J is large and requires dense linear algebra

$$\mathbf{J} = \mathbf{P} \frac{\partial \Psi(\mathbf{m})}{\partial \mathbf{m}}$$

To build **J**, look at function over every time-step:

$$F(\psi^n, \psi^{n+1}, m) = \frac{\theta^{n+1}(\psi^{n+1}) - \theta^n(\psi^n)}{\Delta t} - \nabla \cdot K(\psi^{n+1}, m) \nabla \psi^{n+1} - \frac{\partial}{\partial z} K(\psi^{n+1}, m) = 0$$

Take the devivative, $\frac{\partial F(\psi^n, \psi^{n+1}, m)}{\partial m}$

$$\begin{aligned} \frac{1}{\Delta t} \left(\frac{\partial \theta^{n+1}}{\partial \psi^{n+1}} \frac{\partial \psi^{n+1}}{\partial m} - \frac{\partial \theta^n}{\partial \psi^n} \frac{\partial \psi^n}{\partial m} \right) - \nabla \cdot \text{diag}(\nabla \psi^{n+1}) \left(\frac{\partial K}{\partial m} + \frac{\partial K}{\partial \psi^{n+1}} \frac{\partial \psi^{n+1}}{\partial m} \right) \\ - \nabla \cdot K(\psi^{n+1}) \nabla \frac{\partial \psi^{n+1}}{\partial m} - \frac{\partial}{\partial z} \left(\frac{\partial K}{\partial m} + \frac{\partial K}{\partial \psi^{n+1}} \frac{\partial \psi^{n+1}}{\partial m} \right) = 0 \end{aligned}$$

$$\underbrace{\begin{bmatrix} \mathbf{A}_{-1}(\psi^n) \\ -\frac{1}{\Delta t} \frac{\partial \theta^n}{\partial \psi^n} \end{bmatrix}}_{\mathbf{A}_{-1}(\psi^n)} \underbrace{\frac{\partial \psi^n}{\partial m}}_{\text{green}} + \underbrace{\begin{bmatrix} \mathbf{A}_0(\psi^{n+1}) \\ \frac{1}{\Delta t} \frac{\partial \theta^n}{\partial \psi^{n+1}} - \nabla \cdot \text{diag}(\nabla \psi^{n+1}) \frac{\partial K}{\partial \psi^{n+1}} - \nabla \cdot K(\psi^{n+1}) \nabla - \frac{\partial}{\partial z} \frac{\partial K}{\partial \psi^{n+1}} \end{bmatrix}}_{\mathbf{A}_0(\psi^{n+1})} \underbrace{\frac{\partial \psi^{n+1}}{\partial m}}_{\text{orange}}$$

$$= \underbrace{\nabla \cdot \text{diag}(\nabla \psi^{n+1}) \frac{\partial K}{\partial m} + \frac{\partial}{\partial z} \frac{\partial K}{\partial m}}_{\mathbf{b}(\psi^{n+1}, m)}$$

$$\underbrace{\begin{bmatrix} \mathbf{A}_0(\psi_1) & & & & \\ \mathbf{A}_{-1}(\psi_1) & \mathbf{A}_0(\psi_2) & & & \\ & \mathbf{A}_{-1}(\psi_2) & \mathbf{A}_0(\psi_3) & & \\ & & \ddots & \ddots & \\ & & & \mathbf{A}_{-1}(\psi_{n-1}) & \mathbf{A}_0(\psi_n) \end{bmatrix}}_{\mathbf{A}(\Psi, m)} \underbrace{\begin{bmatrix} \frac{\partial \psi_1}{\partial m} \\ \frac{\partial \psi_2}{\partial m} \\ \vdots \\ \frac{\partial \psi_{n-1}}{\partial m} \\ \frac{\partial \psi_n}{\partial m} \end{bmatrix}}_{\frac{\partial \Psi(m)}{\partial m}} = \underbrace{\begin{bmatrix} \mathbf{b}(\psi_1, m) \\ \mathbf{b}(\psi_2, m) \\ \vdots \\ \mathbf{b}(\psi_{n-1}, m) \\ \mathbf{b}(\psi_n, m) \end{bmatrix}}_{\mathbf{B}(\Psi, m)}$$

$$\mathbf{J} = \mathbf{P} \frac{\partial \Psi(m)}{\partial m}$$

$$\mathbf{J} = \mathbf{P} \mathbf{A}(\Psi, m)^{-1} \mathbf{B}(\Psi, m)$$



$$\left[\begin{array}{c} A_{-1}(\psi) \\ -\frac{1}{\Delta t} \frac{\partial}{\partial \psi} \end{array} \right]$$

```
# Compute part of the derivative of:
#
#      DIV*diag( GRAD*hn1+BC*bc)*(AV*(1.0/K))^(-1)
```

```
DdiagGh1 = DIV*Utils.sdiag( GRAD*hn1+BC*bc)
diagAVk2_AVdiagK2 = (
    Utils.sdiag((AV*(1./K1))**(-2)) *
    AV*Utils.sdiag(K1**(-2))
)
```

$$\frac{\partial \psi^{n+1}}{\partial m}$$

$$\left[\begin{array}{c} A_0(\psi) \\ A_{-1}(\psi) \end{array} \right]$$

```
# The matrix that we are computing has the form:
```

```
#      -
#      |
#      |  Adiag      |
#      |  Asub      Adiag      |
#      |      Asub      Adiag      |
#      |      ...      ...      |
#      |      ...      Asub      Adiag      |
#      |      -
```

$$\left[\begin{array}{c} m) \\ m) \\ m) \\ m) \\ m) \\ m) \end{array} \right]$$

```
Asub = (-1.0/dt)*dT
```

```
Adiag = (
    (1.0/dt)*dT1 -
    DdiagGh1*diagAVk2_AVdiagK2*dK1 -
    DIV*Utils.sdiag(1./(AV*(1./K1)))*GRAD -
    Dz*diagAVk2_AVdiagK2*dK1
)
```

```
B = DdiagGh1*diagAVk2_AVdiagK2*dKm1 + Dz*diagAVk2_AVdiagK2*dKm1
```



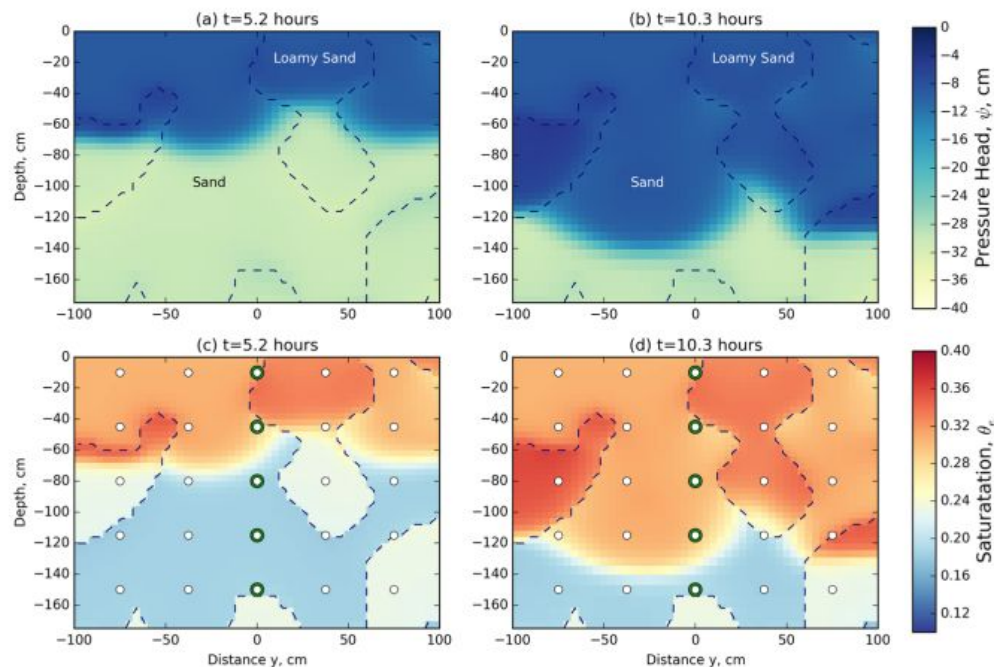
Literature scraping of hydrogeologic params

```
In [1]: from SimPEG import FLOW
```

```
In [2]: VG = FLOW.Richards.Empirical.VanGenuchtenParams()
```

```
In [7]: print VG.sand  
        print VG.loamySand
```

```
{'Ks': 5.833333333333333e-05, 'alpha': 13.8, 'theta_r': 0.02, 'theta_s': 0.417, 'n': 1.592}  
{'Ks': 1.696759259259259e-05, 'alpha': 11.5, 'theta_r': 0.035, 'theta_s': 0.401, 'n': 1.474}
```



4cm discretization
 50*50*45 cells in space (1.125E5)
 12.3 hours
 53 times, (2 - 15 min)
 Fields 5.9 million elements

125 saturation estimates (35 cm apart)
 40 times (18 min apart)
 5000 data points

What can we get out?!

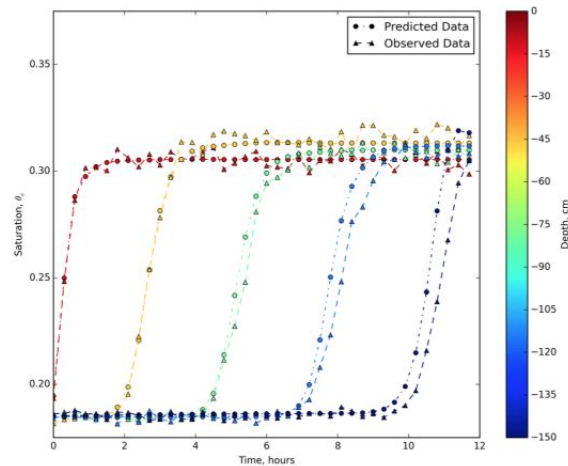
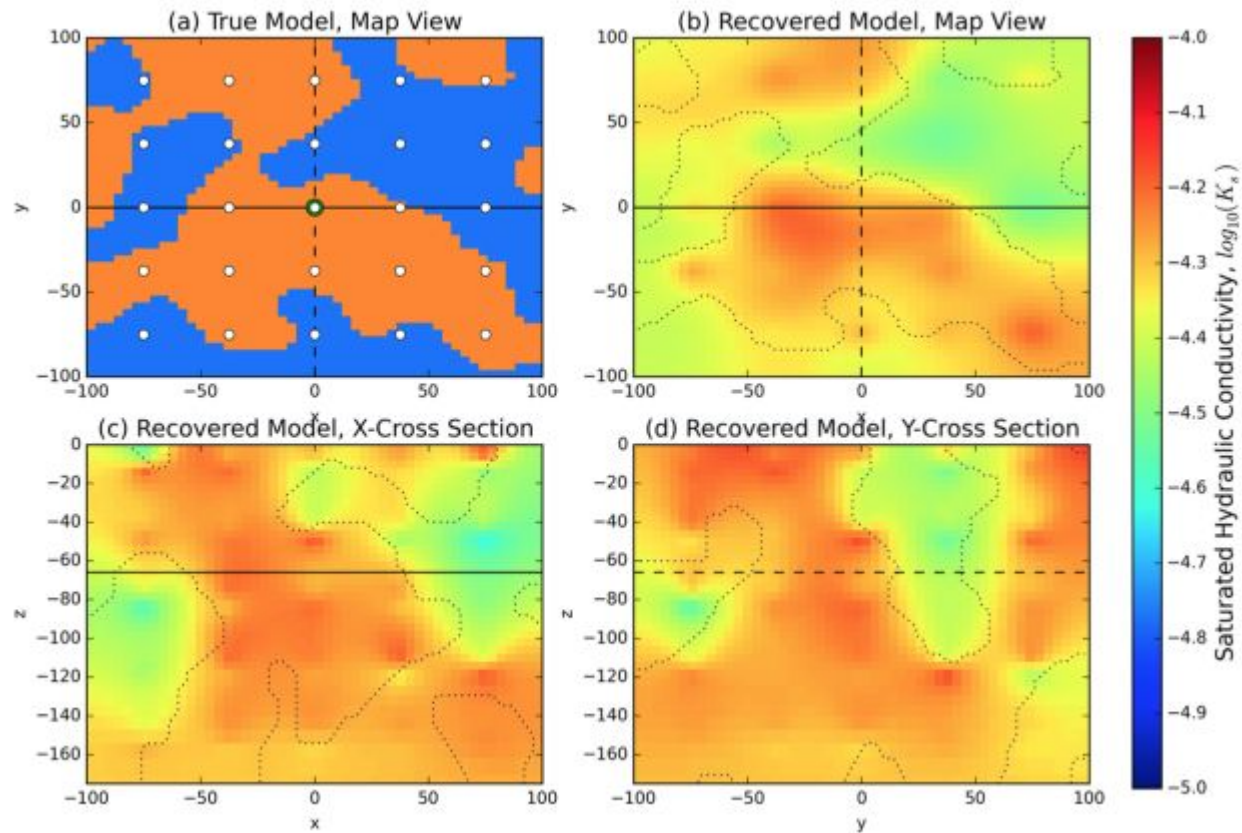
Figure 3: Vertical cross sections through the pressure head and saturation fields from the numerical simulation at two times: (a) pressure head field at $t = 5.2$ hours and (b) $t = 10.3$ hours; and (c) saturation field at $t = 5.2$ hours and (d) $t = 10.3$ hours. The saturation field plots also show measurement locations and green highlighted regions that are shown in Figure 4. The true location of the two soils used are shown with a dashed outline.

```
In [1]: from SimPEG import FLOW
```

```
In [2]: VG = FLOW.Richards.Empirical.VanGenuchtenParams()
```

```
In [7]: print VG.sand
        print VG.loamySand
```

```
{'Ks': 5.833333333333333e-05, 'alpha': 13.8, 'theta_r': 0.02, 'theta_s': 0.417, 'n': 1.592}
{'Ks': 1.696759259259259e-05, 'alpha': 11.5, 'theta_r': 0.035, 'theta_s': 0.401, 'n': 1.474}
```



5000 saturation estimates

112 500 parameters

The dense Jacobian matrix would have 562 million elements

Assume everything is sand, and invert for K_s