

Demo_4_Lapis_2019

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In [3]: `using` PyPlot, LinearAlgebra

1 LAPIS 2019 - UNLP

Use SD to solve sparse inversion problem with Preconding. Rather than operator \mathbf{W} I will use a low-pass operator \mathbf{P}

In [4]: *# Set forward problem and compute mock (synthetic) data*

```
N = 200
M = 500

# Make a sparse signal

m = zeros(M)
m[70:100] = m[70:100] .+ 1.0
m[130:160] = m[130:160] .- 1.0
m[230:260] = m[230:260] .- 0.4
x_max = 100.
x_min = 0.

r_max = 100.
r_min = 0.

alpha = 0.01
A = .1

dx = (x_max-x_min)/(M-1)
dr = (r_max-r_min)/(N-1)

x = [x_min+dx*(i-1) for i in 1:M]
r = [r_min+dr*(i-1) for i in 1:N]

L = A*[exp(-alpha*(( x[i]-r[j] )^2)) for j in 1:N, i in 1:M]

# Make the synthetic data and add noise to it
```

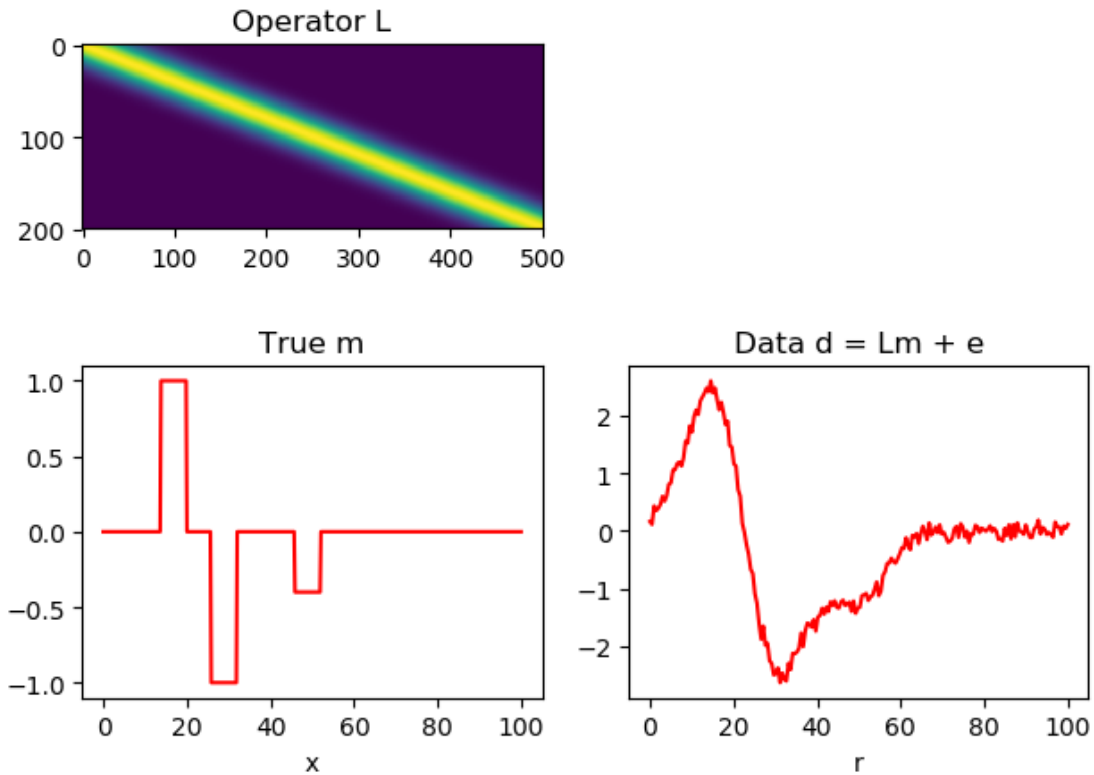
```

d = L*m + 0.1*randn(N)

subplot(221); imshow(L); title("Operator L");
subplot(223); plot(x,m,"r"); title("True m"); xlabel("x")
subplot(224); plot(r,d,"r"); title("Data d = Lm + e"); xlabel("r")

tight_layout()

```



```
In [10]: # Solve via SD
```

```

a = 0.01
mu = 0.01

```

```
# Low pass-filter
```

```
P = 0.25*diagm(-1=> ones(M-1)) + 0.5*diagm(0=> ones(M)) + 0.25*diagm(1=> ones(M-1))
```

```
# SD method
```

```

m_sol = zeros(M)
u = zeros(M)

```

```

for k = 1:300
    u = u - a*(P'*L'*(L*P*u-d)+mu*u)
    m_sol = P*u
end

d_pred = L*m_sol
subplot(221); plot(x,m,"r", x,m_sol, "g"); xlabel("x"); title("True and Estimated Model")
subplot(222); plot(r,d,"r", r,d_pred, "g"); xlabel("r"); title("Observed and Predicted Data")
tight_layout()

```

