# GEORGIA INSTITUTE OF TECHNOLOGY SCHOOL of ELECTRICAL and COMPUTER ENGINEERING <br> ECE $6250 \quad$ Spring 2017 <br> Problem Set \#9 

Assigned: 13-Apr-17
Due Date: 24 -Apr-17

The Final Exam will be held in our regular class room on Friday, April 28 at 3:30 $P M$.

This assignment is due at the beginning of class on Monday, April 28.
As stated in the syllabus, unauthorized use of previous semester course materials is strictly prohibited in this course.

## A note on MATLAB function handles

The two of the problems below require you to write code that uses function handles. These may be unfamiliar, but are very straightforward. Here is an example:

```
>> f = @(z) z + 2;
```

This defines a function $f$ that takes a scalar (or vector) and adds two to it (or every entry):

```
>> f(5)
ans =
```

    7
    Suppose that integers N and K are already defined in the workspace, and we also define the boxcar filter
h = ones(K,1);
We can define a function handle A that takes a vector of length $N$ and convolves it with h:

```
>> A = @(z) ifft(fft(z,N+K-1).*fft(h,N+K-1));
```

Of course since A is a linear mapping from $\mathbb{R}^{N}$ to $\mathbb{R}^{N+K-1}$, there is a corresponding matrix (you have generated this matrix on previous homeworks). Here $A(x)$ implements the action of this matrix on x without having to generate and store the matrix.

## PROBLEM 9.1:

Using you class notes, prepare a 1-2 paragraph summary of what we talked about in class in the last week. I do not want just a bulleted list of topics, I want you to use complete sentences and establish context (Why is what we have learned relevant? How does it connect with other things you have learned here or in other classes?). The more insight you give, the better.

## PROBLEM 9.2:

We are using a radar to track a truck moving in a 2D plane with coordinates $\left(p_{x}, p_{y}\right)$. At a series of times $t_{k}$ indexed by $k$, we are interesting in estimating its position $\left(p_{x}\left(t_{k}\right), p_{y}\left(t_{k}\right)\right.$ in the plane, and its velocity $\left(v_{x}\left(t_{k}\right), v_{y}\left(t_{k}\right)\right)$ along each coordinate. We stack these into a single vector

$$
\boldsymbol{x}_{k}=\left[\begin{array}{c}
p_{x}\left(t_{k}\right) \\
p_{y}\left(t_{k}\right) \\
v_{x}\left(t_{k}\right) \\
v_{y}\left(t_{k}\right)
\end{array}\right] .
$$

Although the velocity of the truck will drift, we expect it to remain close to a constant - that is, our best guess for $\left(v_{x}\left(t_{k+1}\right), v_{y}\left(t_{k+1}\right)\right)$ is simply $\left(v_{x}\left(t_{k}\right), v_{y}\left(t_{k}\right)\right)$. Our best guess for the position at time $t_{k+1}$ is determined by the previous position $\left(p_{x}\left(t_{k}\right), p_{y}\left(t_{k}\right)\right)$, previous velocity $\left(v_{x}\left(t_{k}\right), v_{y}\left(t_{k}\right)\right)$, and the time $t_{k+1}-t_{k}$ that has elapsed between the samples. The evolution of the parameters can be modeled by

$$
\boldsymbol{x}_{k+1}=\boldsymbol{F}_{k} \boldsymbol{x}_{k}+\boldsymbol{\epsilon}_{k} .
$$

At time $t=0$, we make a direct observation that

$$
\left(p_{x}(0), p_{y}(0)\right)=(0,0), \quad \text { and } \quad\left(v_{x}(0), v_{y}(0)\right)=(1, \pi / 2) .
$$

You might write this as $\boldsymbol{y}_{0}=\boldsymbol{C}_{0} \boldsymbol{x}_{0}$, where $\boldsymbol{C}_{0}=\mathbf{I}$. At subsequent times $t_{k}>0$, we make a single measurement of the position and velocity of the truck. This measurement is of the form

$$
y[k]=\cos \left(1150 \pi t_{k}\right) p_{x}\left(t_{k}\right)+\sin \left(1150 \pi t_{k}\right) p_{y}\left(t_{k}\right)+\cos \left(1250 \pi t_{k}\right) v_{x}\left(t_{k}\right)+\sin \left(1250 \pi t_{k}\right) v_{y}\left(t_{k}\right)
$$

1. Write down the "state evolution equations." That is, write down the $4 \times 4$ matrix $\boldsymbol{F}_{k}$ explicitly (it should depend on $t_{k+1}-t_{k}$ ).
2. The file kalman_data.mat contains 499 measurements (in the vector y) for different times (in the vector t ). Implement a Kalman filter to track the truck, and plot each of your 500 estimates of the position in the $\left(p_{x}, p_{y}\right)$ plane from $\hat{\boldsymbol{x}}_{k \mid k}$ on a set of axes. To plot discrete points rather than a connected line, use something like
```
plot(pxhat,pyhat,'o')
```

3. Construct and solve the (large) system to estimate all 500 positions with knowledge of all of the measurements. That is, find $\hat{\boldsymbol{x}}_{0 \mid 499}, \hat{\boldsymbol{x}}_{1 \mid 499}, \ldots, \hat{\boldsymbol{x}}_{499 \mid 499}$. Plot the corresponding position estimates on another set of axes (maybe using ' $x$ ' for the plot command), and overlay your answer from the previous part (maybe using ' 0 ' for the plot command again). Comment on what you see.

## PROBLEM 9.3:

Write a MATLAB function sdsolve.m that implements the steepest descent algorithm. The function should be called as

```
[x, iter] = sdsolve(H, b, tol, maxiter);
```

where H is a function handle that implements the action of an $N \times N$ symmetric positive definite matrix, b is a vector of length $N$, and tol and maxiter specify the halting conditions. Your algorithm should iterate until $\left\|r_{k}\right\|_{2} /\|b\|_{2}$ is less than tol or the maximum number of iterations maxiter has been reached. For the outputs: x is your solution, and iter is the number of iterations that were performed.

Using function handles instead of explicit matrices is easy. Instead of writing
$\mathrm{r}=\mathrm{b}-\mathrm{H} * \mathrm{x}$;
like you would if H is a matrix, you write
r $=\mathrm{b}-\mathrm{H}(\mathrm{x})$;

You will want to debug your code using small (sym+def) matrices that you can construct the inverses to explicitly.

When your code is ready for the big-time, download the files imagedeconv_experiment.m, imconv.m, imagedeconv_data.mat, and imconv_transpose.m. In the mat file, you will find a $305 \times 305$ image Y and a $50 \times 50$ kernel W . The image Y was created by convolving a $256 \times 256$ image X with W . It is your job to figure out what X must have been.

Of course, the code you wrote for sdsolve operates on and returns vectors, not images. But is is easy to turn a $N \times N$ image X into a vector x of length $N^{2}$ :

```
>> x = reshape(X, N^2, 1);
```

(the shorter $\mathrm{x}=\mathrm{X}(:)$; also works) and vice versa:

```
>> X = reshape(x, N, N);
```

The 2D convolution and its transpose have been implemented for you in the files imconv.m and imconv_transpose.m, and created some function handles at the beginning of imagedeconv_experiment.m that will help you. All you have to do is add a few lines to imagedeconv_experiment.m that calls your code and does the recovery. Your solution must have relative residual error $\left\|r_{k}\right\|_{2} /\|b\|_{2}$ less than $10^{-4}$

Turn in your code, the original image (created using imagesc(Y); colormap(gray)), your recovered image, and the number of iterations it took you to reduce the relative residual error to less than $10^{-4}$ from a starting guess of $\mathbf{0}$.

Important note: The pseudo-code below should be your guideline. Note that this uses only one application of H per iteration by using our trick for updating the residual (instead of explicitly
calculating it). In practice, you will probably want to calculate the residual explicitly once every 50 iterations. To do this, just put in an if-then that substitutes $r k=b-H(x k)$ for rk $=r k o l d$ - ak*q every fifty iterations.

## Steepest Descent, more efficient version 2

Initialize: $\boldsymbol{x}_{0}=$ some guess, $k=0, \boldsymbol{r}_{0}=\boldsymbol{b}-\boldsymbol{H} \boldsymbol{x}_{0}$.
while not converged, $\left\|\boldsymbol{r}_{k}\right\|_{2} \geq \delta$ do
$\boldsymbol{q}=\boldsymbol{H} \boldsymbol{r}_{k}$
$\alpha_{k}=\boldsymbol{r}_{k}^{\mathrm{T}} \boldsymbol{r}_{k} / \boldsymbol{r}_{k}^{\mathrm{T}} \boldsymbol{q}$
$\boldsymbol{x}_{k+1}=\boldsymbol{x}_{k}+\alpha_{k} \boldsymbol{r}_{k}$
$\boldsymbol{r}_{k+1}=\boldsymbol{r}_{k}-\alpha_{k} \boldsymbol{q}$
$k=k+1$
end while

## PROBLEM 9.4:

(Optional) Write a MATLAB function cgsolve.m that implements the method of conjugate gradients. The function should be called as

```
[x, iter] = cgsolve(H, b, tol, maxiter);
```

where the inputs and outputs have the same interpretation as in the previous problem.
Use your code to solve the same image deconvolution problem, and comment on the number of iterations required relative to steepest descent. Turn in your code, your recovered image, and the number of iterations it took you to reduce the relative residual error to less than $10^{-4}$ from a starting guess of $\mathbf{0}$.

You will want to explicitly calculate the residual every 50 iterations here as well.

