### 18.06 Problem Set 10 Solutions

1. Do problem 5 from section 6.5 .

Solution $f=(x+3 y)(x+y)=(x+2 y+y)(x+2 y-y)=(x+2 y)^{2}-y^{2}$. There are many points where this is negative, say $(-1,2)$, where the above is $0^{2}-2^{2}=-4$.
This goes to show that not everything is positive-definite, even if all the entries are positive.
2. Do problem 26 from section 6.5.

Solution Let's do an LU of $A=\left(\begin{array}{lll}9 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 2 & 8\end{array}\right)$. We should immediately get $A=\left(\begin{array}{lll}1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 1\end{array}\right)\left(\begin{array}{lll}9 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 4\end{array}\right)$.
By symmetry, we don't have to do much more. We know $D=\left(\begin{array}{ccc}9 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 4\end{array}\right)$ from the pivots, so to get $C^{T}$ we should multiply $L$ by $\left(\begin{array}{ccc}3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 2\end{array}\right)$ to get $\left(\begin{array}{lll}3 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 2 & 2\end{array}\right)$, and $C=\left(\begin{array}{lll}3 & 0 & 0 \\ 0 & 1 & 2 \\ 0 & 0 & 2\end{array}\right)$.
The second matrix is similar. LU gives $A=\left(\begin{array}{ccc}1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1\end{array}\right)\left(\begin{array}{lll}1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 5\end{array}\right)$. We know the square roots of $D$ are $1,1, \sqrt{5}$, so $C^{T}=\left(\begin{array}{ccc}1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & \sqrt{5}\end{array}\right)$ and $C=\left(\begin{array}{ccc}1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & \sqrt{5}\end{array}\right)$.
The point of this is just that the cholesky decomposition really is just $L U$ for a symmetric matrix don't need to think of them as separate things.
3. Do problem 6 from section 6.7.

Solution $A^{T} A=\left(\begin{array}{lll}1 & 1 & 0 \\ 1 & 2 & 1 \\ 0 & 1 & 1\end{array}\right)$ and $A A^{T}=\left(\begin{array}{ll}2 & 1 \\ 1 & 2\end{array}\right)$. Because we know that they "basically" have the same eigenvalues I'm going to save work by using the eigenvalues of $A A^{T}$, which are 3 and 1 (so we know $A^{T} A$ has eigenvalues $3,1,0$ ).
In the order of $3,1,0$, the normalized eigenvectors of $A^{T} A$ are $(1 / \sqrt{6}, 2 \sqrt{6}, 1 \sqrt{6}),(1 / \sqrt{2}, 0,-1 / \sqrt{2}),(1 / \sqrt{6},-1 \sqrt{6}, 1 \sqrt{6})$ and the normalized eigenvectors of $A A^{T}$ are $(1 / \sqrt{2}, 1 / \sqrt{2}),(1 / \sqrt{2},-1 / \sqrt{2})$. Multiplying, we get

$$
A=\left(\begin{array}{cc}
1 / \sqrt{2} & 1 / \sqrt{2} \\
1 / \sqrt{2} & -1 / \sqrt{2}
\end{array}\right)\left(\begin{array}{ccc}
\sqrt{3} & 0 & 0 \\
0 & 1 & 0
\end{array}\right)\left(\begin{array}{ccc}
1 / \sqrt{6} & 2 / \sqrt{6} & 1 / \sqrt{6} \\
1 / \sqrt{2} & 0 & -1 / \sqrt{2} \\
1 / \sqrt{6} & -1 / \sqrt{6} & 1 / \sqrt{6}
\end{array}\right)
$$

which we can check to be correct.
Okay, I just realized you guys already did this in the last pset. Ugh.
4. Do problem 11 from section 6.7.

## Solution

Orthogonality tells us that $A^{T} A$ is going to be diagonal with entries $\rho_{1}^{2}, \ldots, \rho_{n}^{2}$. Thus, the columns of $V$ (or the rows of $V^{T}$ ) are just going to be the eigenvectors $e_{i}=(0, \ldots, 1, \ldots, 0)$, so $V=V^{T}$ is going to be the $n$ by $n$ identity matrix.
$\Sigma$ is going to have the lengths on the diagonal, because those are exactly the positive square roots of the eigenvalues.

Finally, $u_{j}=A v_{j} / \rho_{j}$, which in our case is exactly the normalized column $w_{j} /\left|w_{j}\right|$. So $U$ is just going to have columns of $A$, but normalized.
5. Do problem 13 from section 6.7.

Solution Let's go through the process for the SVD of $R$. note that $R^{T} R=R^{T} Q^{T} Q R=A^{T} A$, so the eigenvalues (and thus $\Sigma$ ) and eigenvectors (and thus $V$ ) remain the same in the two calculations the only thing that changes is $U$.
An alternate way to see this is to note that if we multiply $U$ on the left by a n orthonormal $Q$, the result $Q U$ is still orthonormal because $(Q U)^{T} Q U=U^{T} Q^{T} Q U=I$. Thus, since $R=U S V^{T}$, $A=Q R=(Q U) S V^{T}$ is a valid SVD for $A$.
6. Do problem 2 from section 8.1.

Solution Here we have $A_{1}=\left(\begin{array}{ccc}1 & 0 & 0 \\ -1 & 1 & 0 \\ 0 & -1 & 1\end{array}\right) \cdot A_{1}^{-1}=\left(\begin{array}{ccc}1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1\end{array}\right)$, so the inverse is just $A_{1}^{-1} C_{1}^{-1}\left(A_{1}^{T}\right)^{-1}=\left(\begin{array}{ccc}1 / c_{1} & 1 / c_{1} & 1 / c_{1} \\ 1 / c_{1} & 1 / c_{1}+1 / c_{2} & 1 / c_{1}+1 / c_{2} \\ 1 / c_{1} & 1 / c_{1}+1 / c_{2} & 1 / c_{1}+1 / c_{2}+1 / c_{3}\end{array}\right)$.
7. Do problem 5 from section 8.1.

Solution The solution of this is $y=-\int f(x)+C$, with $C$ determined by $y(1)=0$. For $f(x)=1$ we get $y=-x+1$.
8. Do problem 6 from section 7.1.

Solution Let's call the conditions "additivity" and "scaling" respectively.
[a]: This is scaling the vector into a normal vector. Thus it is impossible that we get additivity, because the sums of normal vectors don't have to be normal. Take $T(0,1)$ and $T(1,0)$ for instance. Howver, true to its name this does have the scaling property, as whatever $c$ we introduce will be canceled from $v$ and $\|v\|$.
[b]: This satisfies both. One immediate way to see this is to see that this is exactly matrix multiplication by $[1,1,1]$, which is a linear operation and thus satisfies both properties.
[c]: This also satisfies both. Again, this is jsut because this is matrix multiplication by $\left(\begin{array}{lll}1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3\end{array}\right)$.
[d]: This doesn't satisfy additivity $((0,1)$ and $(1,0)$ still work). Furthermore, scaling doesn't work either (if we scale by -1 we now pick out the negative of the smallest component, which doesn't have to be related in any way to the largest component.
9. Do problem 27 from section 7.2.

## Solution

The question statement is kinda confusing. I'm parsing it as: "Suppose some linear transformation $T$ sends a basis of $v_{i}$ to a basis of $w_{i}$ via $T\left(v_{i}\right)=w_{i}$. Why must $T$ be invertible?"
$T$ is invertible because we can give an explicit inverse from its image: take the $w_{i}=T\left(v_{i}\right)$ and construct the map $T^{\prime}$ that sends $w_{i}$ to $v_{i}$. This is a well-defined map because there is only one way to define what $T$ does on any vector $w$ (since $w_{i}$ form a basis there is only one way to decompose $w$ into $w_{i}$, which is the heart of the problem). This is easily checked to be linear.

