## Supplement: Lagrange multipliers.

Recall our setup from class: $f: \mathbb{R}^{n} \rightarrow \mathbb{R}$ is $C^{1}$, and $g: \mathbb{R}^{n} \rightarrow \mathbb{R}^{m}$ is $C^{1}$. The function $f$ is your "cost", which you want to maximise (or minimise) subject to the constraint $g=0$.

The usual strategy is to construct a function

$$
H(x, \lambda)=f(x)+\lambda g(x)
$$

and observe that critical points of $H$ must satisfy the constraint $g=0$. Here $\lambda=\left(\lambda_{1} \lambda_{2} \ldots \lambda_{n}\right)$ is a $1 \times n$ matrix ( $n$ being the number of coordinates $g$ has, which is the same as the "number of constraints").

The typical situation (see your homework) will leave you with finitely many critical points of $H$, which you can usually deal with from other considerations. (There also exists a second derivative test involving the augmentation of the Hessian, which is described in your book and required for the last problem.)

Unfortunately, the observation that critical points of $H$ satisfy the constraint $g$ does not rule out the possibility of the existence of a constrained maximum of $f$ which is not a critical point of $H$ ! (You saw an example of this in the section, and there is one on your homework.) However, if 0 is a regular value of $g$, then any constrained maximum of $f$ must in fact correspond to a critical point of $H$. I proved this for $n=2, m=1$ in class. But the underlying idea is the same in higher dimensions, provided you know the Rank Nullity theorem. Since there are enough other problems to put on your homework this week, I write it up below.

Proposition 1. Suppose 0 is a regular value of $g$, and a constrained (local) maximum of $f$ given $g=0$ is attained at the point $x_{0}$. Then there exists an $1 \times m$ matrix $\lambda_{0}$ such that

$$
D f_{x_{0}}+\lambda_{0} D g_{x_{0}}=0
$$

Consequently $\left(x_{0}, \lambda_{0}\right)$ is a critical point of $H$.
Proof. Let $x=(y, z)$, where $y \in \mathbb{R}^{n-m}$ and $z \in \mathbb{R}^{m}$. Without loss of generality, we assume that the last $m$ columns of the matrix $D g_{x_{0}}$ are linearly independent. Thus, by the implicit function theorem there exists $\varepsilon, \delta$ and a $C^{1}$ function $h: \mathbb{R}^{n-m} \rightarrow \mathbb{R}^{m}$ such that

$$
\{g=0\} \cap B\left(x_{0}, \varepsilon\right)=\left\{(y, h(z)) \mid y \in B\left(y_{0}, \varepsilon\right)\right\}
$$

Now, $f$ attains a constrained (local) maximum at $x_{0}$ if and only if the function $\varphi: \mathbb{R}^{n-m} \rightarrow \mathbb{R}^{m}$ defined by $\varphi(y)=f(y, h(y))$ has a critical point at $y_{0}$. Thus

$$
0=D \varphi_{y_{0}}=D f_{x_{0}}\binom{I}{D h_{y_{0}}}
$$

Equivalently, this means the vector $\nabla f$ is in the kernel of the matrix $M$ defined by

$$
M \stackrel{\text { def }}{=}\left(\begin{array}{ll}
I & \left(D h_{y_{0}}\right)^{*}
\end{array}\right) .
$$

On the other hand, differentiating the identity $g(y, h(y))=0$, we see

$$
D g_{x_{0}}\binom{I}{D h_{y_{0}}}=0
$$

Equivalently, each of the vectors $\nabla g_{1}\left(x_{0}\right), \ldots, \nabla g_{m}\left(x_{0}\right)$ are in the kernel of the matrix $M$.

Now let's count dimensions. Note that $M$ is an $(n-m) \times n$ matrix, and the first $n-m$ columns are the same as that of the $(n-m) \times(n-m)$ identity matrix. This forces $\operatorname{rank}(M)=n-m$. So by the Rank-Nullity theorem, $\operatorname{dim}(\operatorname{ker}(M))=$ $n-\operatorname{rank}(M)=m$.

Observe that the $m$ vectors $\nabla g_{1}\left(x_{0}\right), \ldots, \nabla g_{m}\left(x_{0}\right)$ are all linearly independent (because rank $D g_{x_{0}}=m$ ). Since they're all in the kernel of $M$, a subspace of dimension exactly $m$, these vectors must form a basis of $\operatorname{ker}(M)$. Thus $\nabla f_{x_{0}}$ (which know is in $\operatorname{ker}(M))$ must be a linear combination of these basis vectors. So $\exists \lambda_{1}, \ldots \lambda_{m} \in \mathbb{R}$ such that

$$
\nabla f_{x_{0}}=\sum_{1}^{m} \lambda_{i} \nabla g_{x_{0}}
$$

Taking the transpose finishes the proof.

