9 Matlab Tricks that You Probably Want to Know

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Gatsby Tea Talk

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1. Matrix storage is column-major order

Physical memory is linear.

To store a multi-dimensional array, need to arrange it linearly.

Matlab:

• $A = \begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix} \in \mathbb{R}^{r \times c}$ is internally stored as $(1, 2, 3, 4, 5, 6)^{\top}$ (column-major).

Tricks/Facts:

- A(1,2) == 3. Can also use linear index. A(3) == 3
- To flatten A, do $A(:) == (1, 2, 3, 4, 5, 6)^{\top}$. Get a column vector.
- Internally, Matlab does A((j-1)r+i) for A(i,j).
- C/C++, Python use row-major order.

2. Set diagonal elements

Task:

•
$$A = \begin{pmatrix} 1 & 4 & 7 \\ 2 & 5 & 8 \\ 3 & 6 & 9 \end{pmatrix} \in \mathbb{R}^{r \times r}$$
. Want to set the diagonal to 0.

Don't want to use (slow)

for
$$i=1:r$$

A(i, i) = 0;
end

Tricks:

- Use linear indexing. A(1:(r+1):end) = 0.
- "end" == 9.
- 1: (r+1): end == 1: 4: 9 == [1, 5, 9] == indices of the diagonal elements.

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3. reshape

reshape(..) is used to change the shape of an array.

Read elements in linear order (column-wise).

•
$$A = \begin{pmatrix} 1 & 3 & 5 \\ 2 & 4 & 6 \end{pmatrix}$$

• reshape(A, 1, 6) == (1,2,3,4,5,6). Row vector.
• reshape(A, 3, 2) == $\begin{pmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 6 \end{pmatrix}$.

■ reshape(A, 3, 3). Get an error.

- reshape(A, 3, 2) == reshape(A(:), 3, 2)
- reshape(..) is computationally very cheap.

4. Weighted average on a 3D array

Task:

• $T \in \mathbb{R}^{r \times c \times d}$, a 3d array e.g., d images of size $r \times c$.

• $v \in \mathbb{R}^d$, a weight vector.

• Want to multiply to get $M = \sum_{i=1}^{d} T(:,:,i) * v(i) \in \mathbb{R}^{r \times c}$.





Do not want to use a loop.

Tricks

- Use reshape
- $\blacksquare R = \operatorname{reshape}(T, r * c, d)$
- $\blacksquare M = \operatorname{reshape}(R * v, r, c)$

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5. Minimum element of a multi-dimensional array

Task:

- **E** $\in \mathbb{R}^{r imes c imes d}$ e.g., validation errors of param.1 imes param.2 imes param.3
- Find the minimum error, and the corresponding three parameters.

Problem:

- Matlab's min operates along one dimension.
- Tedious to find min three times.

Tricks:

[minerr, ind] = min(E(:)); [p1_ind, p2_ind, p3_ind] = ind2sub(size(E), ind);

Flatten the array E(:). Find min and its linear index (ind).

Convert the linear index back to the subscript index.

6. $\operatorname{tr}(A^{\top}B)$

Task:

• $A, B \in \mathbb{R}^{m \times n}$. Want $\operatorname{tr}(A^{\top}B)$.

Inefficient to compute $A^{\top}B$ and take the trace.

Tricks:

• Let
$$A := (\boldsymbol{a}_1 | \cdots | \boldsymbol{a}_n)$$
 and $B := (\boldsymbol{b}_1 | \cdots | \boldsymbol{b}_n)$.
 $\operatorname{tr}(A^\top B) = \operatorname{sum}(\operatorname{diag}(A^\top B))$
 $= \sum_{j=1}^n \boldsymbol{a}_j^\top \boldsymbol{b}_j = \sum_{j=1}^n \sum_{i=1}^m a_{ij} b_{ij}$
 $= \operatorname{sum}(\operatorname{sum}(A. * B))$
 $= A(:)' * B(:)$ in Matlab

• trace(A'*B) costs $O(mn^2)$.

• Compute A'*B. Then, throw away off-diagonal entries.

• A(:)'*B(:) = sum(sum(A.*B)) costs O(mn).

7. log-sum-exp trick (not specific to Matlab)

- Want $r^{(k)} = \frac{\prod_{d=1}^{D} p_d^{(k)}}{\sum_{k'=1}^{K} \prod_{d=1}^{D} p_d^{(k')}}$ where $p_d^{(k)} \in (0,1)$ and D is big.
- Example: Posterior probability of the kth-component of a mixture of Bernoulli.

Problem:

• $\prod_{d=1}^{D} p_d^{(k)}$ leads to numerical underflow. Try prod(rand(1, 1000)). Tricks:

- 1 Store log prob. $\log r^{(k)} = \sum_d \log p_d^{(k)} \log \sum_{k'} \prod_d p_d^{(k')}$
- 2 Introduce c

$$\begin{split} \log \sum_{k'} \prod_d p_d^{(k')} &= \log \exp(c) + \log \exp(-c) + \log \sum_{k'} \exp\left(\log \prod_d p_d^{(k')}\right) \\ &= c + \log \sum_{k'} \exp\left(\sum_d \log p_d^{(k')} - c\right), \\ \text{choose } c \text{ so that } \exp\left(\sum_d \log p_d^{(k')} - c\right) > 0. \\ \text{One way is } c &:= \max_{k'} \sum_d \log p_d^{(k')} < 0. \end{split}$$

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8. bsxfun and repmat

Task:

- $\blacksquare A \in \mathbb{R}^{m \times n}, v \in \mathbb{R}^m.$
- Want B = f(A, v) (f: element-wise) such that $B_{ij} = f(A_{ij}, v_i)$.
- Example: Subtract mean from each column.

Tricks:



- **Trick 1:** f(A, repmat(v, [1, n]))
- Trick 2: bsxfun(@f, A, v)
 - Same effect as Trick 1 without replicating v. Memory efficient.
- bsxfun can only take in simple f
 - $f \in \{$ @plus, @minus, @times, @max, @eq, ...}, not any arbitrary f

See "doc bsxfun".

bsxfun also works for

9. Embarassingly parallel for-loop

- Want to run an embarassingly parallel for-loop on multiple machines.
- **Example:** validation_error(θ_i) for *i* in a long list.

Tricks:

Download Multicore package (open source).

 $\verb+http://uk.mathworks.com/matlabcentral/fileexchange/13775-multicore-parallel-processing-on-multiple-coresistic structure and the struct$

- Master/slave machines need to share temp_dir for passing information.
- On slave Matlab's, run

startmulticoreslave(temp_dir);

On the master,

```
v_error_func = .. (some func. of theta) ..
thetas = {t1, t2, ...}
resultCell = startmulticoremaster(v_error_func, thetas, setting);
```

- resultCell{i} == validation error of θ_i .
- Master/slave machines can be on the same or different machines. Need to share the same file system. Work at Gatsby.
- Should launch slave Matlab's through the job queue (slurm).

