# ICTS 4156: Introduction to ML 

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## 1 Notes on Quiz

## 2 Notes on Homework

Use the formulas on slide 18 to compute Z2, A2, and Z3 (the softmax logit scores). Use the code from the previous assignment to compute the softmax probabilities A3 and the corresponding softmax loss. The only difference is that now softmax uses A2 as input, instead of X.

## 3 Notes on NNs

- $a_{3}^{(2)}=f\left(W_{31}^{(1)} * x_{1}+W_{32}^{(1)} * x_{2}+W_{33}^{(1)} * x_{3}+b_{3}^{(1)}\right)$
- The input can be seen as layer $L_{1}$ composed of neuron activations. Let $a_{1}^{(1)}=x_{1}$. In general, $a_{j}^{(1)}=x_{j}$.
- $a_{3}^{(2)}=f\left(W_{31}^{(1)} * a_{1}^{(1)}+W_{32}^{(1)} * a_{2}^{(1)}+W_{33}^{(1)} * a_{3}^{(1)}+b_{3}^{(1)}\right)$.
- $a_{3}^{(2)}=f\left(z_{3}^{(2)}\right)$, where $z_{3}^{(2)}=W_{31}^{(1)} * a_{1}^{(1)}+W_{32}^{(1)} * a_{2}^{(1)}+W_{33}^{(1)} * a_{3}^{(1)}+b_{3}^{(1)}$
- In general, $a_{j}^{(2)}=f\left(W_{j 1}^{(1)} * a_{1}^{(1)}+W_{j 2}^{(1)} * a_{2}^{(1)}+W_{j 3}^{(1)} * a_{3}^{(1)}+b_{j}^{(1)}\right)$.
- In general, $a_{j}^{(l+1)}=f\left(W_{j 1}^{(l)} * a_{1}^{(l)}+W_{j 2}^{(l)} * a_{2}^{(l)}+\ldots W_{j s_{l}}^{(l)} * a_{s_{l}}^{(l)}+b_{j}^{(l)}\right)$.
- In general, $a_{j}^{(l+1)}=f\left(z_{j}^{(l+1)}\right)$ where $z_{j}^{(l+1)}=W_{j 1}^{(l)} * a_{1}^{(l)}+W_{j 2}^{(l)} * a_{2}^{(l)}+\ldots W_{j s_{l}}^{(l)} * a_{s_{l}}^{(l)}+b_{j}^{(l)}$.
- Vectorized, $a_{j}^{(l+1)}=f\left(z_{j}^{(l+1)}\right)$ where $z_{j}^{(l+1)}=W^{(l)}[j,:]^{T} \mathbf{a}^{(l)}+b_{j}^{(l)}$.
- Fully vectorized, $\mathbf{a}^{(l+1)}=f\left(\mathbf{z}^{(l+1)}\right)$ where $\mathbf{z}^{(l+1)}=W^{(l)} \mathbf{a}^{(l)}+\mathbf{b}^{(l)}$.
$\operatorname{softmax}(\mathbf{z})=\left[a_{1}, a_{2}, \ldots a_{K}\right]$ where $a_{j}=\frac{\exp \left(z_{j}\right)}{\sum_{k=1}^{K} \exp \left(z_{k}\right)}$, assuming $\mathbf{z}=\left[z_{1}, z_{2}, \ldots, z_{K}\right]$

