

# A VARIATIONAL AUTOENCODER TO GENERATE WINGED HORSES

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## ABSTRACT

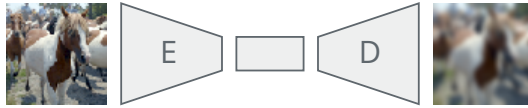
This paper proposes using a convolutional variational autoencoder to generate winged horse (Pegasus) images. To achieve this a variational autoencoder is trained on a images of horses and birds to learn a distribution over the data that permits sampling. The experiments performed suggest that this approach is poor at producing these images and produces blurry images that slightly resemble winged horses.

## 1 METHODOLOGY

The method is to train an autoencoder [1], by minimising the squared L2 loss:

$$\mathcal{L}_{\text{AE}} = \mathbb{E}_{\mathbf{x} \sim p_{\text{data}}} [\|\mathbf{x} - D(E(\mathbf{x}))\|^2] \quad (1)$$

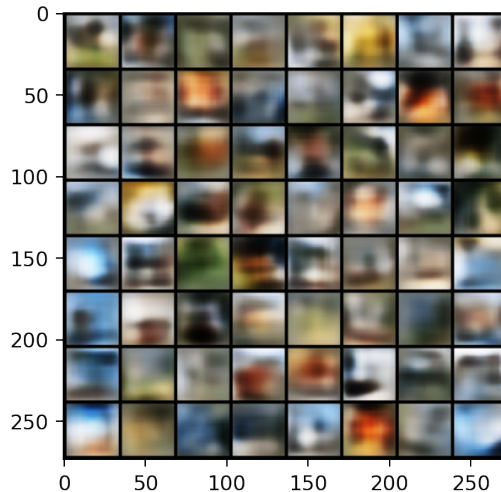
The methodology should be very concise and the mathematical notation should try to follow the ICLR conference guidelines [\[1\]](#). If you are not familiar with  $\text{\LaTeX}$ , you can use this online  $\text{\LaTeX}$  equation editor [\[2\]](#). You may want to include an architectural diagram:



The architectural diagram above was created using Inkscape and exported to a PDF. This was then uploaded to the figures directory on the left.

## 2 RESULTS

The results look very blurry, where the best batch of images looks like this:



From this batch, the most Pegasus-like image (with quite a stretch of the imagination) is:



### 3 LIMITATIONS

It's very difficult to see anything that looks like a Pegasus. In the future, this could be improved by training for more than 10 epochs, although this was not possible due to the time constraints.

### BONUSES

This submission has a total bonus of -4 marks (a penalty), as it is trained only on CIFAR-10, and the Pegasus has a dark body colour.

### REFERENCES

- [1] Mark A Kramer. "Nonlinear principal component analysis using autoassociative neural networks". In: *AIChE journal* 37.2 (1991), pp. 233–243.