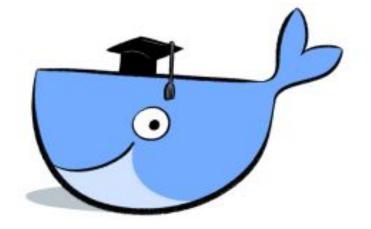
# Docker for Science



### Outline

- What is Docker?
- Workflow
- Use cases
  - Portable environments
  - Reproducibility
- Demo
- Related technologies

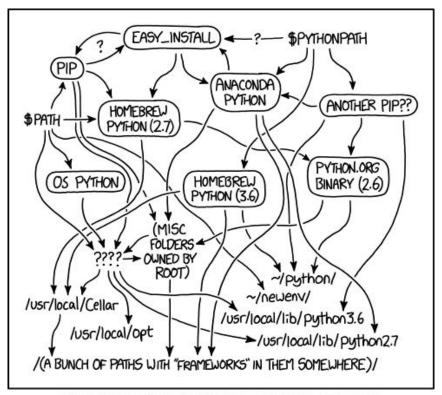
#### What is Docker?

An operating system level virtualization platform based on Linux

#### Workflow

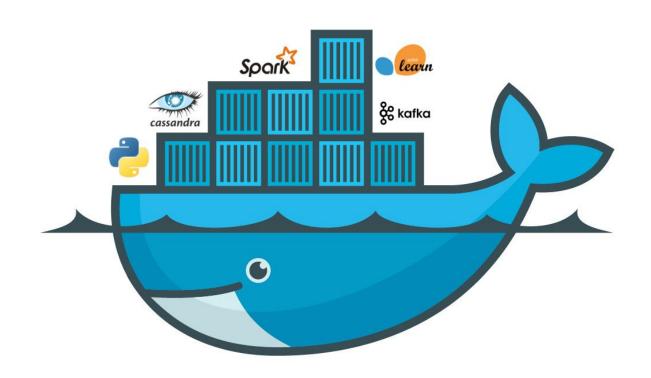
- Build Docker images
- **Push** images to Docker Hub
- Run Docker containers
- Deploy and scale applications

#### Use Case: Portable Environments



MY PYTHON ENVIRONMENT HAS BECOME SO DEGRADED THAT MY LAPTOP HAS BEEN DECLARED A SUPERFUND SITE.

### Use Case: Portable Environments



Lipton and Steinhardt: "Troubling trends of machine learning scholarship"

- Failure to distinguish between explanation and speculation
- Failure to identify sources of empirical gains
- Mathiness
- Misuse of language

the expression

$$b_{x,y}^i = a_{x,y}^i / \left( k + \alpha \sum_{j=\max(0,i-n/2)}^{\min(N-1,i+n/2)} (a_{x,y}^j)^2 \right)^{\beta}$$

where the sum runs over n "adjacent" kernel maps at the same spatial position, and N is the total number of kernels in the layer. The ordering of the kernel maps is of course arbitrary and determined before training begins. This sort of response normalization implements a form of lateral inhibition inspired by the type found in real neurons, creating competition for big activities amongst neuron outputs computed using different kernels. The constants  $k, n, \alpha$ , and  $\beta$  are hyper-parameters whose values are determined using a validation set; we used k=2, n=5,  $\alpha=10^{-4}$ , and  $\beta=0.75$ . We applied this normalization after applying the ReLU nonlinearity in certain layers (see Section 3.5).

Not ideal...

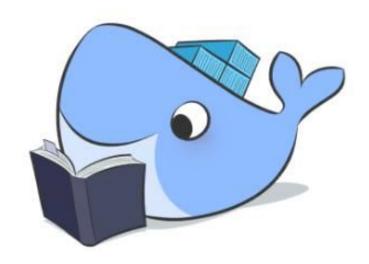
```
batch_size: 128
34
     num_train_steps: 200000
     steps_per_stats: 100
     steps_per_sample: 500
     steps_per_eval: 500
     num_buckets: 5
     sample_decodings: 4
40
     optimizer: 'adam'
41
     learning_rate: 0.0001
     start_decay_step: 20000
43
     decay_steps: 2000
44
     decay_factor: 0.5
45
46
47
     attention: True
```

Better...



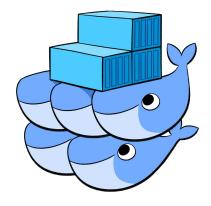
Great!

### Demo



### Related Technologies

- Docker Swarm + Compose
- Kubernetes

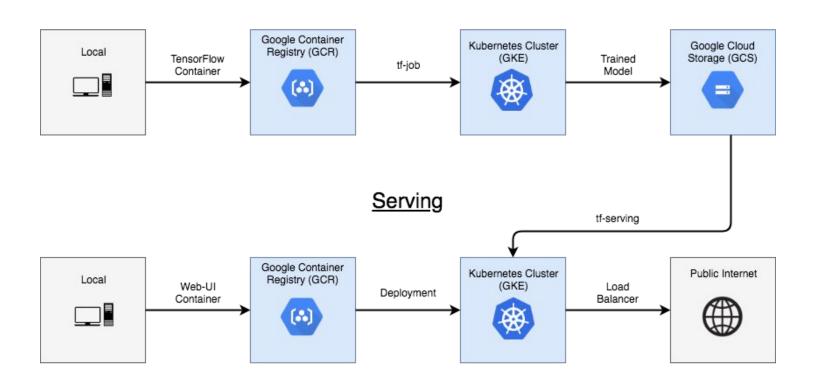




### Kubernetes

#### Kubeflow

#### **Training**



#### Links of interest

- Deep learning container: <a href="https://github.com/floydhub/dl-docker">https://github.com/floydhub/dl-docker</a>
- Troubling Trends in ML: <a href="https://arxiv.org/pdf/1807.03341.pdf">https://arxiv.org/pdf/1807.03341.pdf</a>
- Jupyter Docker stacks: <a href="https://github.com/jupyter/docker-stacks">https://github.com/jupyter/docker-stacks</a>
- Auto-sklearn:
   <a href="https://papers.nips.cc/paper/5872-efficient-and-robust-automated-machine-learning">https://papers.nips.cc/paper/5872-efficient-and-robust-automated-machine-learning</a>
- Kubeflow: <a href="https://www.kubeflow.org/">https://www.kubeflow.org/</a>