

# **GEOML & MODEL DEMOCRATIZATION WITH OSM DATA**

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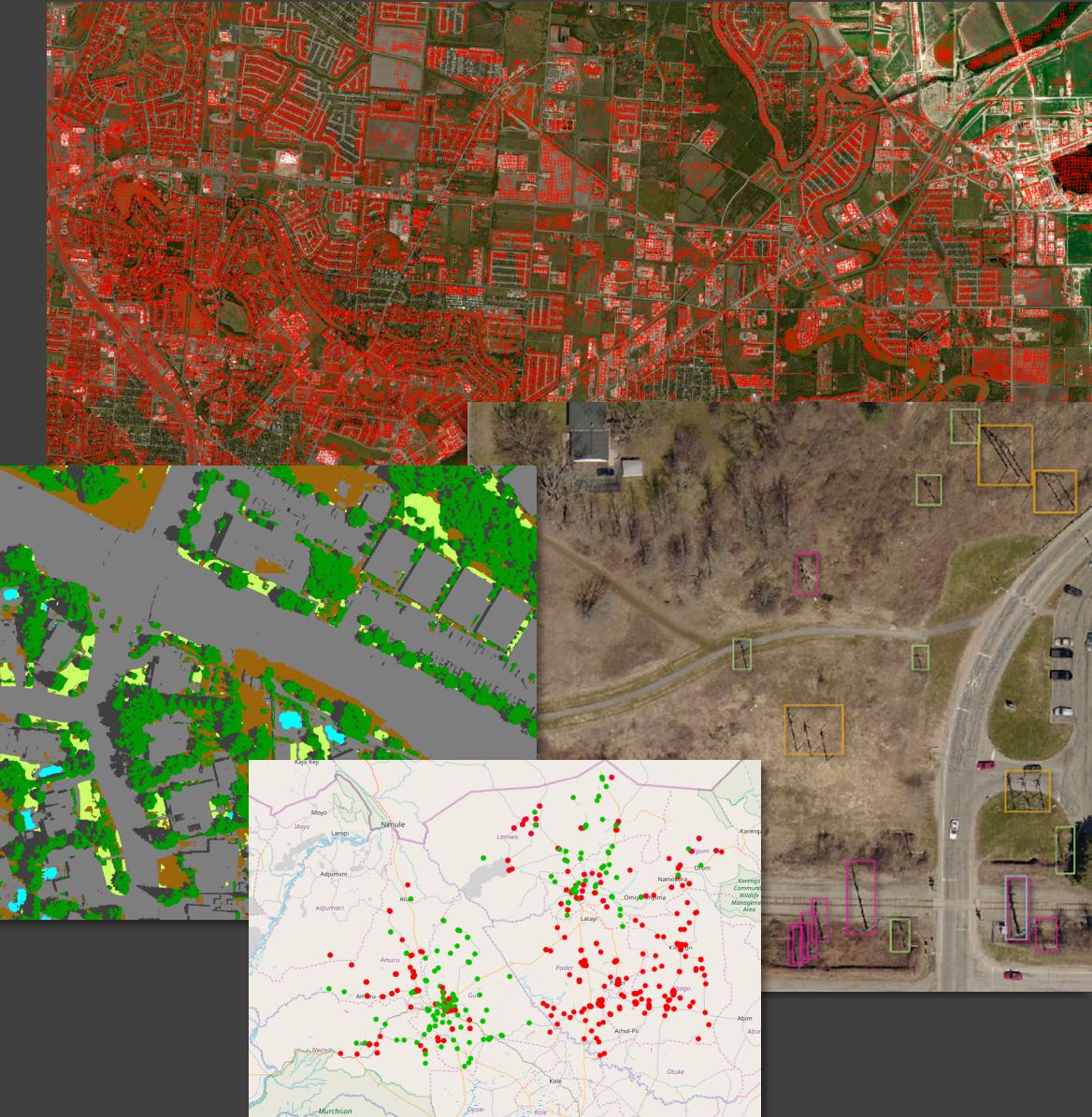
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# THE OSM + ML POTENTIAL

- ML (CNN) scales extraction of visual patterns w/ supervision
- A wealth of curated, crowd-sourced expertise that could be used for supervision
  - Still requires additional curation & extraction
- Avenue for adding contributions & education
  - Co-proposed creating a focus at UW (Geohackweek) to modify standard GIS content with end-to-end OSM to model generation
  - *Could we create a user ‘market’ of open-geoML models?* 🤔



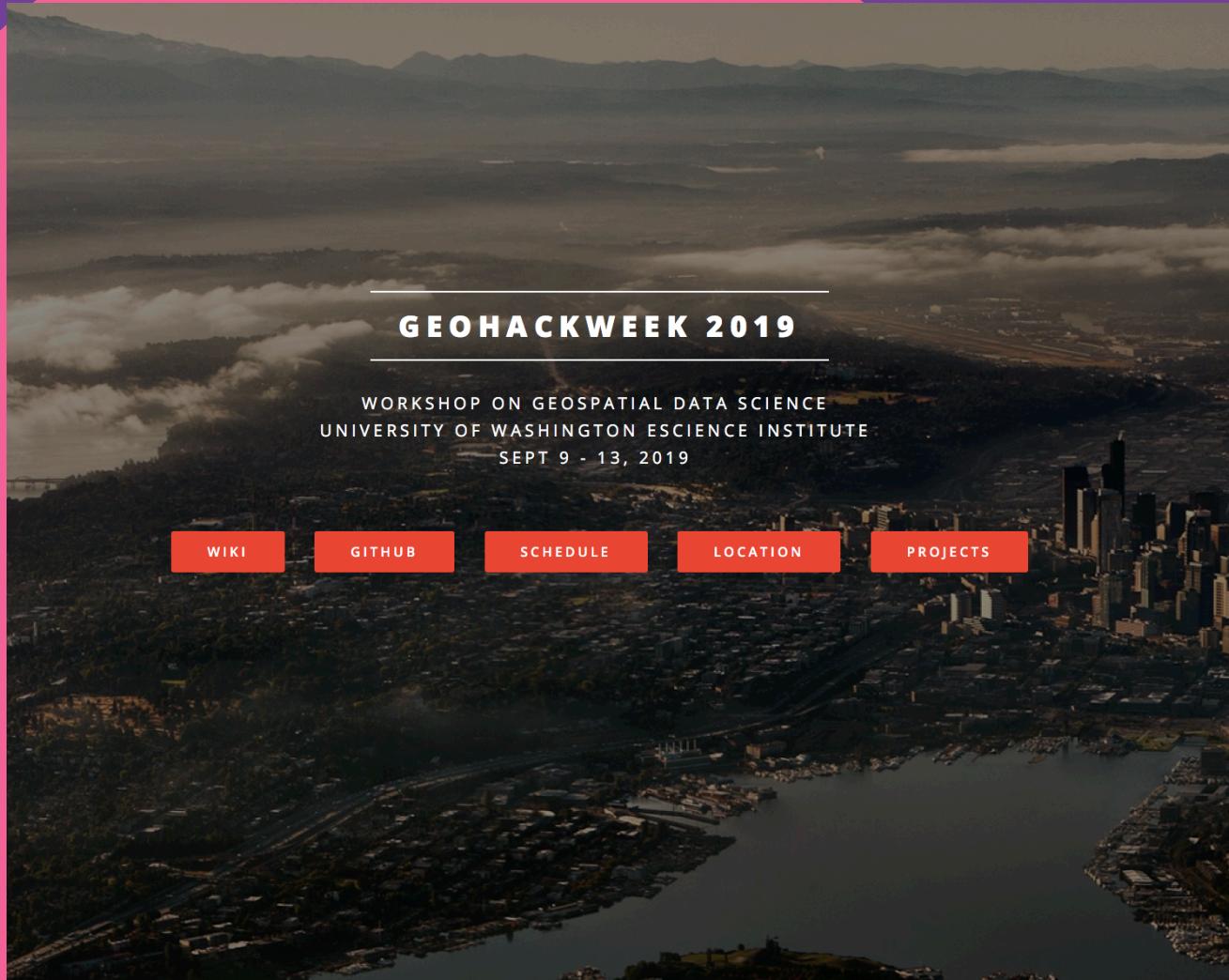
# AN OVER-ABUNDANCE OF ML PLATFORMS

- Democratize dev of ‘geoML’, lower the barrier to entry for students & practitioners, & obliterate the ‘practice’ of platform commercialization
- geoML is not hard, but daunting
  - Difficulty as a DS with limited cloud-orchestration resources
  - Opensource/open-framework desires (common tools, community driven)
  - The POC is easy. The scaling is hard.
  - Big spenders ‘own’ the university programs (drive content)
  - Model marketplaces & commercialization that force you into platform use



# UW GEOHACKWEEK

(<https://geohackweek.github.io>)



## GEOHACKWEEK 2019

WORKSHOP ON GEOSPATIAL DATA SCIENCE  
UNIVERSITY OF WASHINGTON ESCIENCE INSTITUTE  
SEPT 9 - 13, 2019

WIKI

GITHUB

SCHEDULE

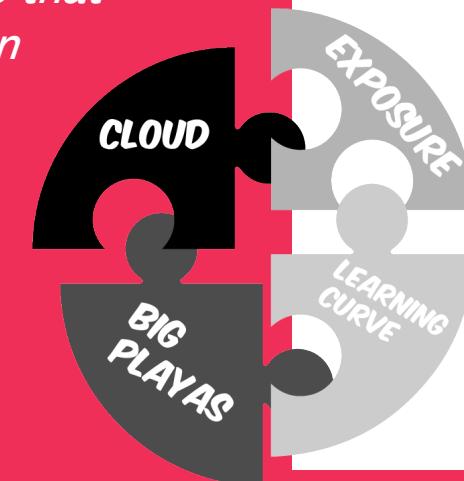
LOCATION

PROJECTS

- **2018 Software Carpentry Approach**
  - Last minute volunteer to teach ML in 1 hr 1 month after moving to Seattle
  - 1 week to cover many aspects of geospatial + github + opensource
  - Team projects
- **2019 ML Revolution**
  - We can do so much more!
  - Leverage HOT-OSM data sets as vector training data
  - **Raster -> Vector -> Machine Learning**
  - <https://github.com/scottyhq/geohackweek2019-raster>
  - [https://github.com/geohackweek/tutorial\\_contents/tree/master/vector/notebooks](https://github.com/geohackweek/tutorial_contents/tree/master/vector/notebooks)
  - <https://github.com/shaystrong/sagely/>

# EDUCATION

- Leveraging OSM + ML + cloud compute
  - University of Washington's GeoHackweek Multi-disciplinary, multi-background
  - HOT-OSM tasking (vector) → ML outcome building prediction in 1 week!
  - 50 new OSM mappers!!
  - AWS sponsorship (negotiated free compute!)
  - Personal goal (unaffiliated): *Advance stagnant geospatial activities and create a community invested in producing optimal solutions that become foundational to advanced, open endeavors.*



Screenshot of the GitHub repository page for "University of Washington Geohackweek".

The page shows the following details:

- Repository:** University of Washington Geohackweek
- Owner:** eScience Institute
- URL:** <https://geohackweek.github.io/>
- Repositories:** 42
- Packages:** 0
- People:** 95
- Teams:** 7
- Projects:** 0

Search bar: Find a repository... Type: All Language: All

New button

Top languages: Jupyter Notebook (red), CSS (blue), JavaScript (yellow), HTML (orange), CSS (purple), Dockerfile (black).

People section: 95 > [grid of 15 user profile pictures].

Repository list:

- preliminary**: Preliminary Tutorial, Jupyter Notebook, 3 views, 0 stars, 1 forks, 0 issues, Updated 5 days ago.
- geohackweek.github.io**: CSS, 27 views, 14 stars, 0 forks, 0 issues, Updated 5 days ago.
- sample\_project\_repository**: Jupyter Notebook, 0 views, 0 stars, 0 forks, 0 issues, Updated 17 days ago.
- databsharing**: JavaScript, 2 views, 0 stars, 0 forks, 0 issues, Updated 17 days ago.
- administrative**: Hackweek project management, Jupyter Notebook, 0 views, 0 stars, 22 forks, 0 issues, Updated 18 days ago.
- wiki**: 5 views, 0 stars, 0 forks, 0 issues, Updated 18 days ago.
- hackweeks**: 0 views, 0 stars, 0 forks, 0 issues, Updated 20 days ago.
- reproducible-research**: JavaScript, 2 views, 0 stars, 0 forks, 0 issues, Updated 28 days ago.
- ghw2018**: Archived, CSS, 2 views, 0 stars, 0 forks, 0 issues, Updated on Jun 26.

geotiff  
COG  
mbtiles

pixels!

raster

gdal  
slippy-map

rasterio  
georectification

satellite  
vs.  
aerial

super-  
mercado

geoML

vector

postgres/  
postGIS

geopandas  
shapely

ogr2ogr

OSMNX

over  
pass  
turbo

tippecanoe  
HOT

Pascal  
VOC  
sage-  
maker

conv.  
neural  
networks  
semantic  
segmentation

ML  
object  
detection

mxnet  
image  
classification  
data  
munging

data  
curation

AWS

cloud  
scaling

## OSM vector data as ML training labels

**Cyclone Kenneth 2019-04-25**

### Part I

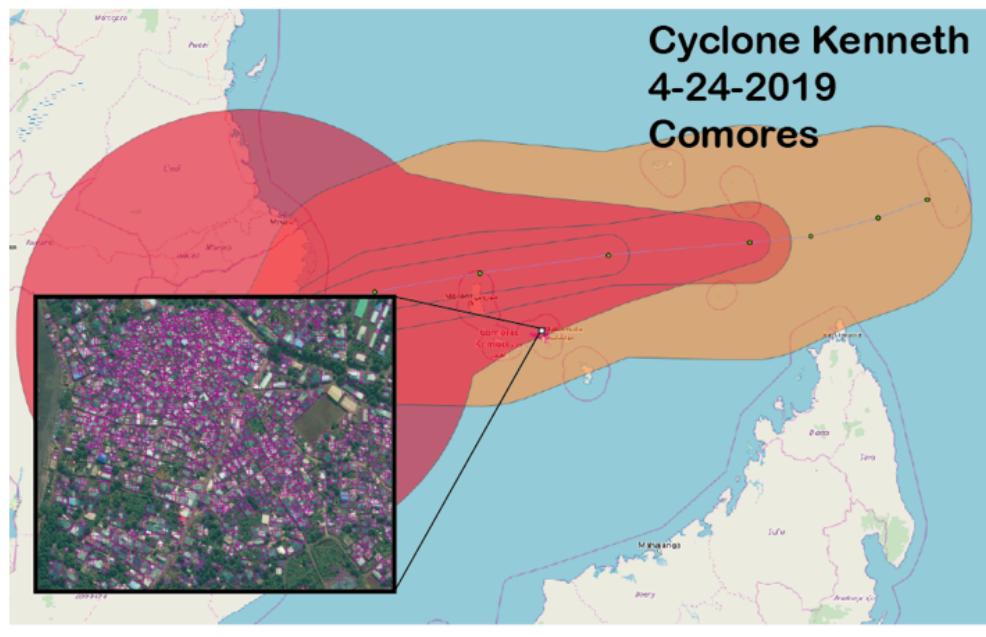
So I wanted to create a seamless tutorial for taking OpenStreetMap (OSM) vector data and converting it for use with machine learning (ML) models. In particular, I am really interested in creating a tight, clean pipeline for disaster relief applications, where we can use something like crowd sourced building polygons from OSM to train a supervised object detector to discover buildings in an unmapped location.

The recipe for building a basic deep learning object detector is to have two components: (1) training data (image (raster) + label (vector) pairs) and (2) model framework. The deep learning model itself will be a Single Shot Detector (SSD) object detector. We will use OSM polygons as the basis of our label data and Digital Globe imagery for the raster data. We won't go into the details of an SSD here, as there are plenty sources available. We will run the object detector in [AWS Sagemaker](#).

In this part I you will:

1. Get vector data from OSM
2. Convert them to labels for a CNN object detection (using Apache MXNET)
3. Store them in VOC style
4. Create optimized .rec files for porting them into the AWS Sagemaker world
5. AWS S3 & EC2

I anticipate using this tutorial in conjunction with [HOT-OSM](#) related tasks -- where we may have drawn vector data as part of a specific project and know it exists. For the purpose of establishing a demo, we will use a recent HOT OSM task area that was impacted by Cyclone Kenneth in 2019, Nzwani, Comores.



<https://github.com/shaystrong/sagely>

## AWS Sagemaker Training and Deploying

**Cyclone Kenneth 2019-04-25**

### Part II

In this part II notebook, we will upload the data to AWS S3 that we generated for training in the previous notebook. We will kick off an AWS Sagemaker object detection job and monitor the results. At the end of this notebook, you will have trained your own OSM-based CNN object detector!



A couple of things worth noting:

- 💡 ML models are not super useful unless they are scaled across a large amount of data
- 💡 To effectively scale across data, you need to be efficient
- 💡 Because we will be passing sensitive data to this notebook in order to scale our cloud compute through Sagemaker, we will use papermill to run this notebook from within python. It creates a simple wrapper around the notebook so that we can specify variables.

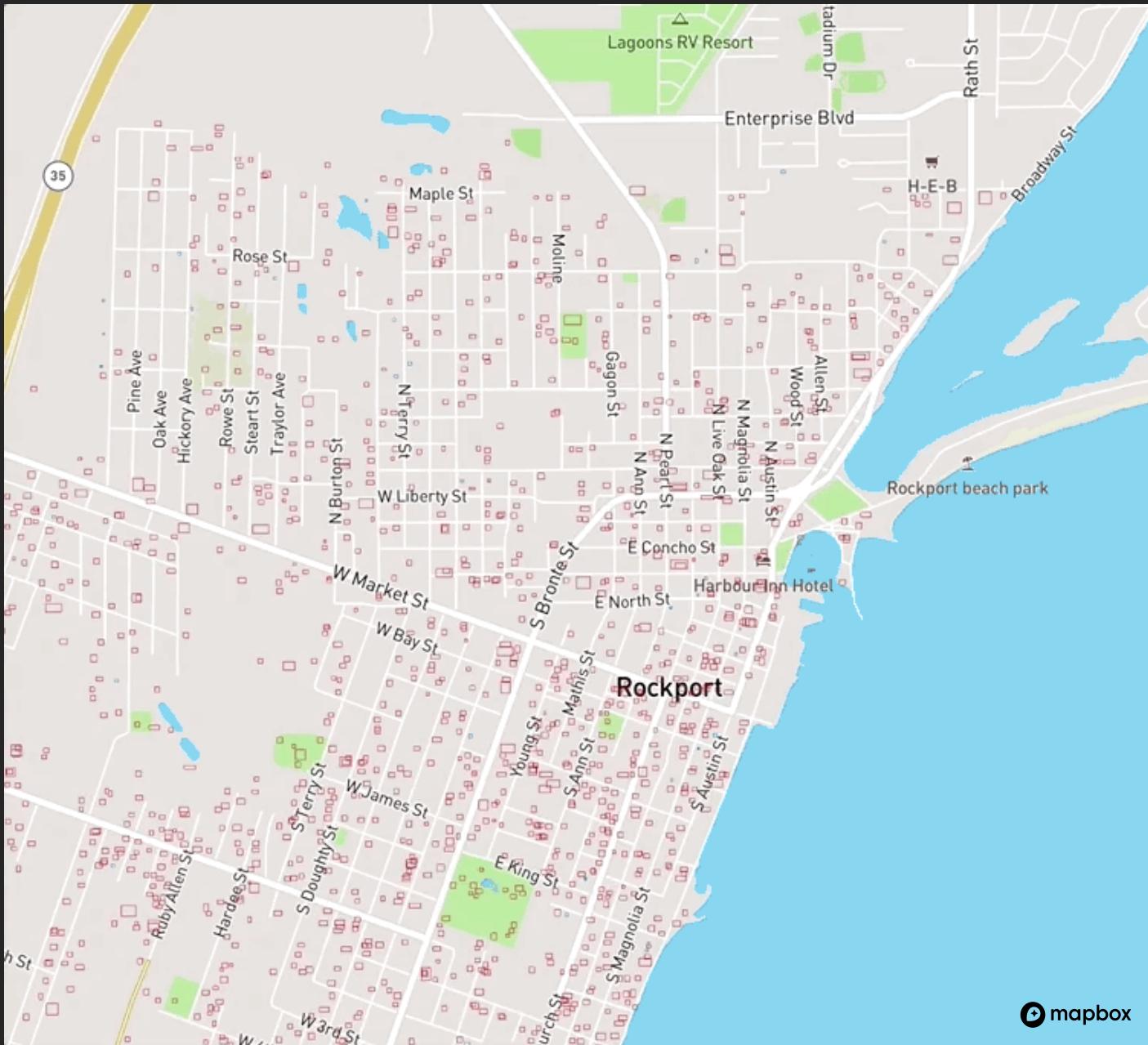
e.g.

```
import papermill as pm
pm.execute_notebook('osm_ml_training_pt2.ipynb','osm_ml_training_pt2_out.ipynb', parameters = dict(sage_bucket='',my_bucket='', role=''))
```

```
In [9]: import sagemaker
from sagemaker import get_execution_role
from sagemaker.amazon.amazon_estimator import get_image_uri
```

We will use 'papermill' (<https://github.com/nteract/papermill>) to pass sensitive variables to this jupyter notebook. Things like passwords, cloud locations, etc, should be parameterized as a best practice -- Never stored in a repo (especially public facing).

# OSM TO STORM RESPONSE



- Flood commercialized marketplaces with models leverage-able by a broader community
  - Extricate 'free' models from their marriage to black-box platforms
- What could incentivization or a social currency approach look like in the long run?
- Challenges: Imagery, cloud compute, annotation cleansing

The screenshot shows the Algorithmia AI Layer interface. At the top, there are navigation links: Product, Pricing, Learn, Sign In, Try It For Free, and Get Your Demo. Below this is a search bar and a large 'Share' button. The main area is titled 'Algorithms' and features four categories with icons: Text Analysis (book icon), Machine Learning (brain icon), Computer Vision (camera icon), and Deep Learning (neuron icon). Each category has a brief description. Below these are tabs for Utilities, Microservices, Web Tools, Time Series, and Sentiment Analysis. A 'Top Rated' section lists 'Colorful Image Colorization' and 'Summarizer'. At the bottom, there are filters for Top Rated, Most Called, and Recently Added.

The screenshot shows a news article from WIRED magazine. The title is 'NEED SOME AI? YEAH, THERE'S A MARKETPLACE FOR THAT' by DAVEY ALBA, published on 09.15.16 at 01:27 PM. The article discusses the emergence of AI marketplaces. The WIRED logo is at the top left, and there are social sharing buttons for Facebook and Twitter.

The screenshot shows the AWS Marketplace interface. The search bar is set to 'Machine Learning (382 results)'. The results page lists several items, including:
 

- Deep Learning Notebook (Python 3, Tensorflow)** by 1.13, Pytorch 1.1. Starting from \$0.00 to \$0.20/h for software + AWS usage fees. Description: Jupyter notebook instance ready to train deep learning models - Start coding in minutes - Runs on GPU automatically if available (choose p2 or p3 instances), otherwise runs on CPU - Latest versions of pytorch and tensorflow - Automatically starts a jupyter notebook server on https port 8888. The password... Linux/Unix, Amazon Linux Amazon Linux 2 - 4.14 kernel (ami-0b898040803850657 - 64-bit x86) - 64-bit Amazon Machine Image (AMI).
- HERE Location Services** by HERE. Starting from \$0.00 to \$0.20/h for software + AWS usage fees. Description: HERE Location Services (HLS) make it easy for developers to bring enterprise-grade location intelligence to AWS applications and solutions. HLS portfolio of REST and JavaScript APIs allow users to quickly address business problems. Geocoding and Search - Geocoder: Forward and Reverse - Batch Geocoder...
- H2O.ai Driverless AI Latest (BYOL)** by H2O.ai. Starting from \$0.00 to \$0.20/h for software + AWS usage fees. Description: H2O Driverless AI is an artificial intelligence (AI) platform that automates some of the most difficult data science and machine learning workflows such as feature engineering, model validation, model tuning, model selection, and model deployment. It aims to achieve highest predictive accuracy, comparable... Linux/Unix, Ubuntu/Ubuntu - 64-bit Amazon Machine Image (AMI).
- AWS Deep Learning Containers for TensorFlow** by AWS. Starting from \$0.00 to \$0.20/h for software + AWS usage fees. Description: AWS Deep Learning Containers for TensorFlow are pre-configured Amazon Lambda functions that provide a simple way to run TensorFlow models on AWS Lambda. They are designed to be used with the AWS Lambda API and can be triggered by various AWS services, such as Amazon S3, Amazon Kinesis, and Amazon CloudWatch Metrics.