

Painless machine learning in production

H. Chase Stevens

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[@hchasestevens](https://twitter.com/hchasestevens)

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Lessons from industry
regarding pain reduction
and data scientist

empowerment in the

productionization of

machine learning models

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@hchasestevens





teikametrics

teikametrics Welcome! Jan 1, 2020 → Jun 1, 2020

Estimated Gross Profit	Total Sales	Total Ad Sales	Total Ad Spend	TACoS
\$283,512 +30%	\$1,142,872 +30%	\$350,239 +30%	\$31,082 +31%	6.5% -2.4

Workflow opportunities

- New Keywords** 6,155
Add the top search terms from your Sponsored Products campaigns as new keywords so that your ads appear in the best locations.
[Add 6,155 search terms as new keywords →](#)
- Negative Keywords** 43
Start removing the keywords from your ad groups that are costing you money without converting into sales!
[Take action on 43 negative keywords →](#)
- Automatic to Manual**
Take your best keywords from Automatic campaigns and add them to Manual to lower ad spend.

Discover Flywheel

Flywheel Bid Changes 84,391 This month
Flywheel data science is constantly tuning your keyword bids to maximize sales and increase overall profitability.
[See Flywheel's recent bid changes →](#)

Additional dashboard elements:

- Home
- My Products
- Advertising
- Analytics
- Promotions
- Expert Services
- Help Center
- Expert Chat
- Settings

Note: A woman wearing safety glasses is overlaid on the bottom left of the dashboard screenshot.

Contents

- Motivation
- Developer experience
- Our stack
- Lessons learned

Motivation

I. Ops is intrinsic to ML

Motivation

- I. Ops is intrinsic to ML
- II. MLOps is unsustainable

Motivation

I. Ops is intrinsic to ML

II. MLOps is unsustainable



Data scientists need to productionize their own models

Motivation

I. Ops is intrinsic to ML

II. MLOps is unsustainable



Data scientists need to productionize their own models

III. Data scientists want to do data science

Motivation

I. Ops is intrinsic to ML

II. MLOps is unsustainable



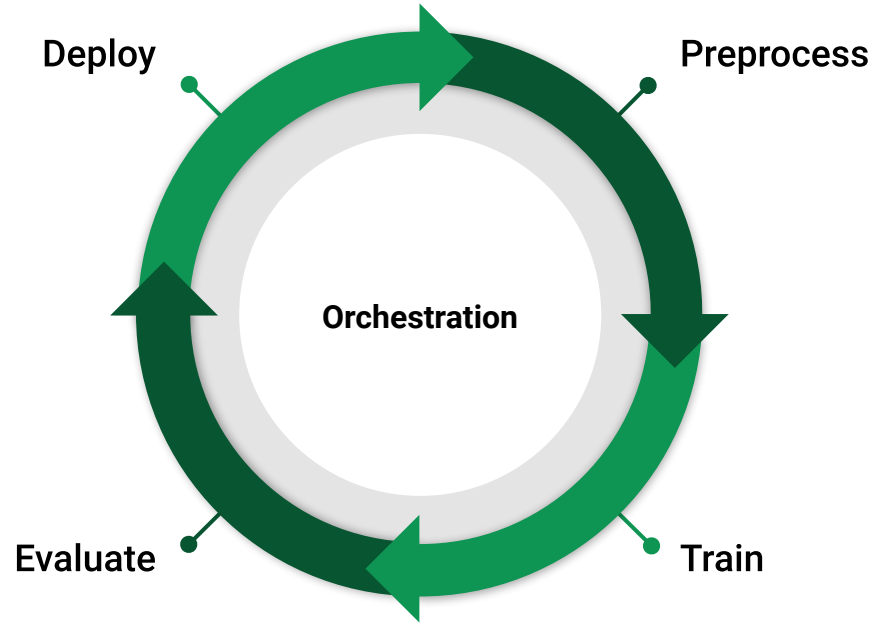
Data scientists need to productionize their own models

III. Data scientists want to do data science

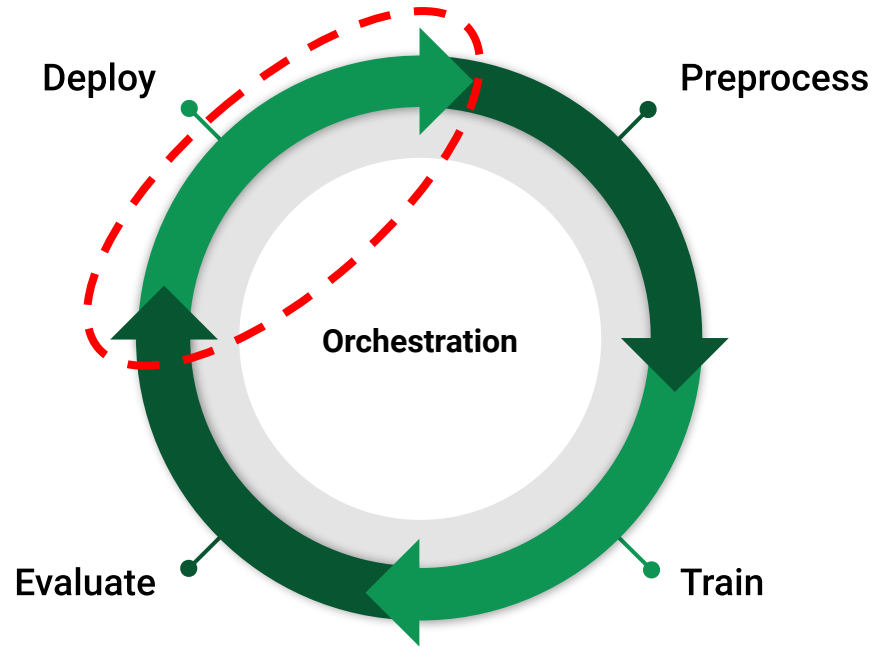


We need tooling and services to minimize “ops” overhead

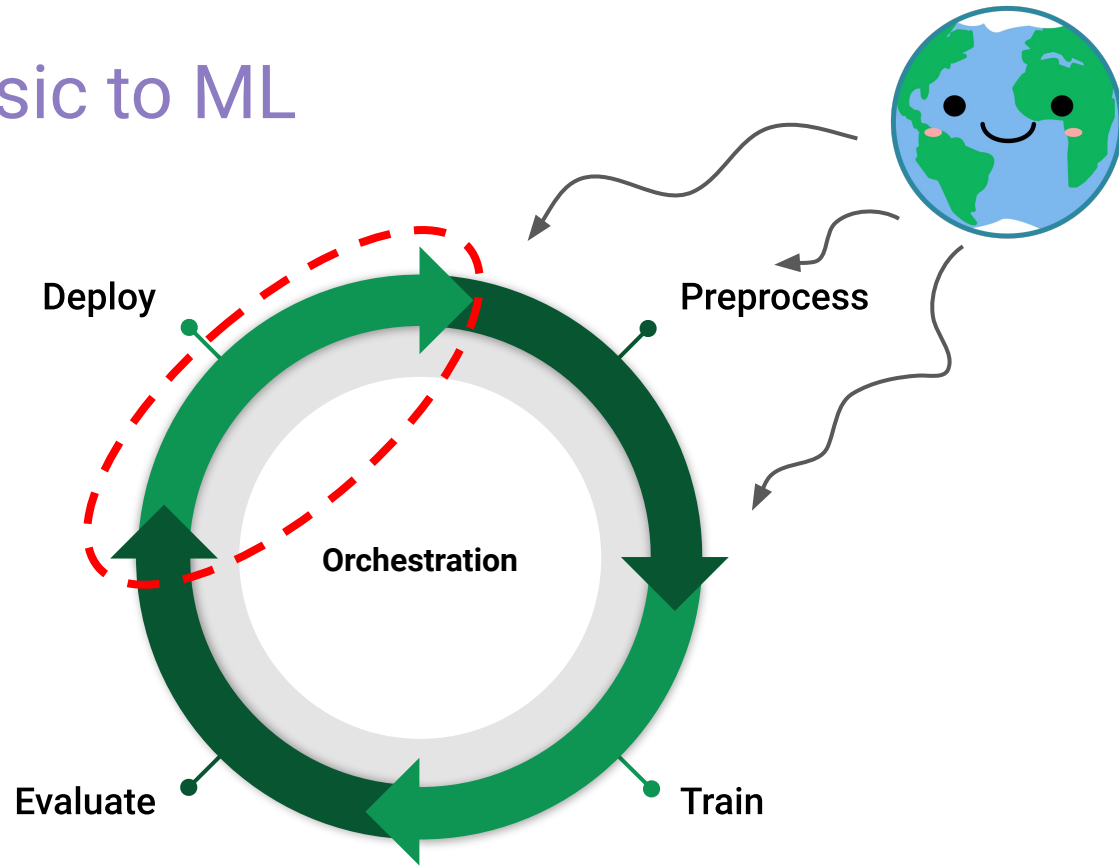
I. Ops is intrinsic to ML



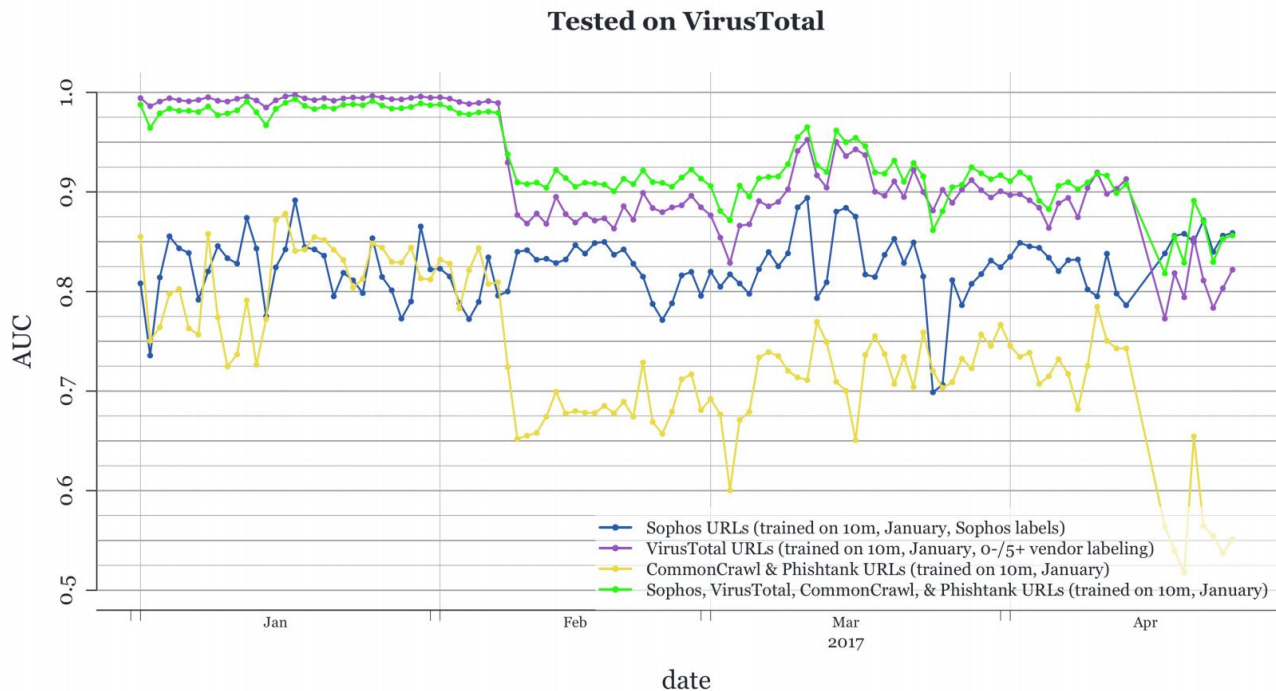
I. Ops is intrinsic to ML



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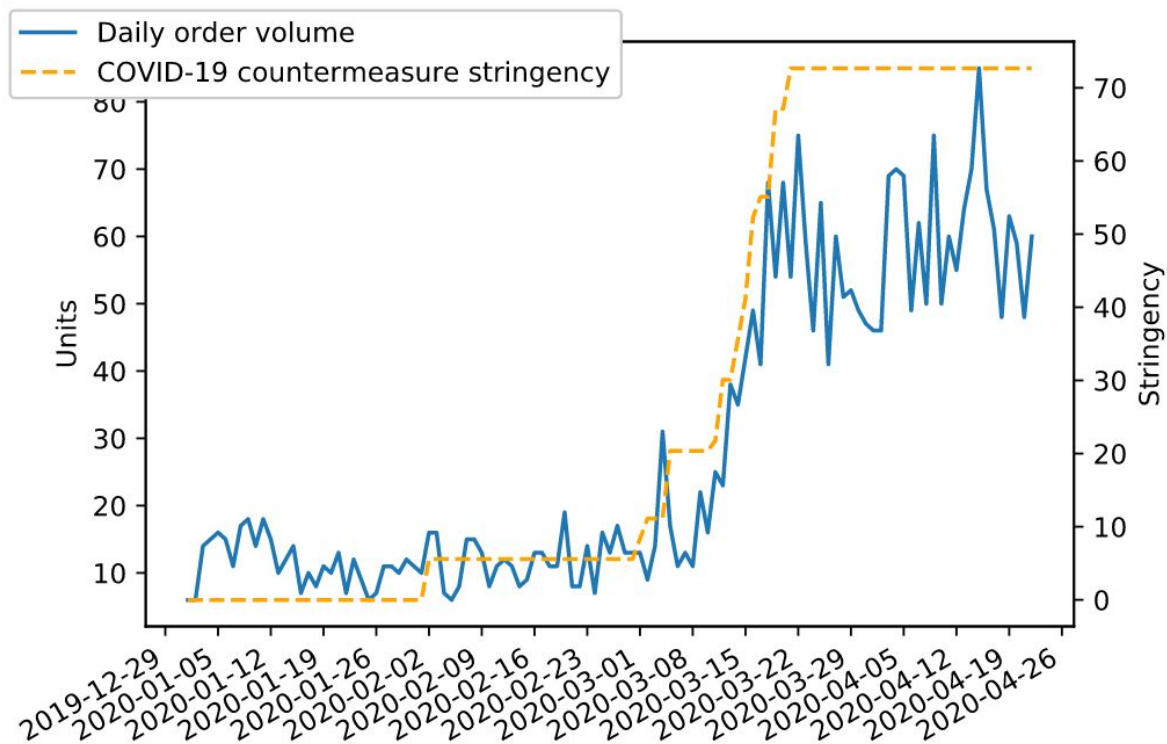


I. Ops is intrinsic to ML



Sanders, H., & Saxe, J. (2017). Garbage in, garbage out: how purportedly great ML models can be screwed up by bad data.

I. Ops is intrinsic to ML



II. MLOps is unsustainable (in 1970)

"You couldn't even delete a mistake"

"I had to wait hours for my programs to turn around"

"Only a select few programmers were allowed in the computer lab."

"One of our finals was to design, code, punch, debug a solution - we got 4 days to do it which means finding typos, logic errors, and design errors and eliminating them all with only 4 re-runs"

"I submitted my program to the punch card crew, and got it back several days later with a rather strong note"

II. MLOps is unsustainable (in 1970)

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II. MLOps is unsustainable (in 2000)

Code → QA → Release (?)

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The Rise Of DevOps: Why Enterprise Is Moving to DevOps

Published On: August 2, 2017 by [Thomas Johnston](#)

To stay competitive in 2017 and beyond, enterprise organizations are embracing DevOps methodologies and new technologies to accelerate

II. MLOps is unsustainable (today)

“Here’s the model”

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“Here’s the model”

“This data isn’t available yet”

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“This takes too long in prod”

II. MLOps is unsustainable (today)

“Here’s the model”

“Try this instead”

“That should be corrected”

“Try again?”

“OK, delete that part”

“... Ready to try version two?”

“This data isn’t available yet”

“Wrong version of numpy”

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Developer experience

```
$ cookiecutter git@github.com:teikametrics/sagemaker-framework.git
github_username [my-github-username]: hchasestevens
project_name [my-sagemaker-model]: europython-example-model
project_slug [europython_example_model]:
model_name [europython-example-model]:
description [An ML model living on the SageMaker platform.]: An example model for
Europython 2020.
Select model_validation_metric:
1 - sklearn.metrics.mean_squared_error
2 - sklearn.metrics.r2_score
3 - sklearn.metrics.accuracy_score
4 - sklearn.metrics.log_loss
5 - sklearn.metrics.f1_score
6 - sagemaker_framework.utils.metrics.mean_absolute_percentage_error
Choose from 1, 2, 3, 4, 5, 6 (1, 2, 3, 4, 5, 6) [1]: 1
Select promotion_criterion:
1 - sagemaker_framework.utils.promotion.maximize
2 - sagemaker_framework.utils.promotion.minimize
3 - sagemaker_framework.utils.promotion.maximize_with_tol
4 - sagemaker_framework.utils.promotion.minimize_with_tol
5 - sagemaker_framework.utils.promotion.manual
6 - sagemaker_framework.utils.promotion.always_promote
Choose from 1, 2, 3, 4, 5, 6 (1, 2, 3, 4, 5, 6) [1]: 6
preprocessing_cpus [1]:
preprocessing_memory_in_gb [4]: 8
test_proportion [0.2]: 0.1
training_cpus [1]:
training_memory_in_gb [4]:
training_volume_size_in_gb [2]:
max_training_runtime_in_minutes [30]: 60
min_serving_instances [1]:
max_serving_instances [10]: 1
```

Developer experience

```
$ cookiecutter git@github.com:teikametrics/sagemaker-framework.git
github_username [my-github-username]: hchasestevens
project_name [my-sagemaker-model]: europython-example-model
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```
$ tree -a europython-example-model/
europython-example-model/
├── .bellybutton.yml
├── bin
│   ├── build-docker-image
│   └── deploy.sh
├── .circleci
│   └── config.yml
├── docker-compose.yml
├── Dockerfile
├── europython_example_model
│   ├── config.py
│   ├── __init__.py
│   └── model.py
├── .github
│   ├── CODEOWNERS
│   └── PULL_REQUEST_TEMPLATE.md
├── .gitignore
├── README.md
├── requirements.txt
├── sagemaker-config.yml
├── setup.py
├── tests
│   ├── test_config.py
│   ├── test_model.py
│   └── test-model.txt
```

5 directories, 19 files

Developer experience

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5 directories, 19 files

Developer experience

```
1 def preprocess_data(seed=None) -> PreprocessingResult:
    """Preprocess data for training."""
    fetch_adgroup_performances_query = """
        SELECT
            ad_group_id,
            SUM(lkr.conversions_7d_attr) AS conversions,
            SUM(lkr.sales_7d_attr) AS sales
        FROM main.transforms.latest_keyword_reports lkr
        WHERE lkr.conversions_7d_attr > 0
            AND lkr.sales_7d_attr > 0
            AND lkr.keyword_report_local_date >= current_date() - 30
        GROUP BY ad_group_id
    """
    return PreprocessingResult(
        training={
            'performances.msgpack': adgroup_performances[
                ~adgroup_performances.test
            ].apply(pd.to_numeric).to_msgpack(),
        }.items(),
        validation=(),
        testing=test_cases
    )
```

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```

```
2 def train_model(training_path: Path, validation_path: Path) -> Artifacts:
    training_dfs = load_zipped_data(
        training_path,
        fnames=MSGPACK_FNAMES,
        deserializer=pd.read_msgpack
    )
    all_adgroup_prices = training_dfs['prices.msgpack']
    performances = training_dfs['performances.msgpack']
    results = {
        marketplace_id: train_marketplace_model(
            marketplace_id=marketplace_id,
            market_adgroup_prices=market_df,
            performances=performances,
        )._asdict()
        for marketplace_id, market_df in all_adgroup_prices.groupby('marketplace_id')
    }
    return Artifacts({MODEL_FNAME: json.dumps(results).encode('utf-8')})
```

Developer experience

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        for marketplace_id, market_df in all_adgroup_prices.groupby('marketplace_id')
    }
    return Artifacts({MODEL_FNAME: json.dumps(results).encode('utf-8')})
```

```
3 def load_model(path: Path) -> Model:
    with (path / MODEL_FNAME).open('r', encoding='utf-8') as f:
        parameters = {k: Parameters(**v) for k, v in json.load(f).items()}
    def model(configuration, instances) -> List[Optional[float]]:
        return [
            estimate_sales_per_conversion(...)
            for price, conversions, sales in instances
        ]
    return model
```

Developer experience

```
request_schema: !jsonschema {
  type: 'object',
  properties: {
    configuration: {
      type: 'object',
      properties: {
        marketplaceId: {type: 'string'}
      },
    },
  },
  instances: {
    type: 'array',
    items: {
      type: 'array',
      items: [
        {type: 'number', description: "Price", exclusiveMinimum: 0},
        {type: 'number', description: "Conversions", exclusiveMinimum: 0},
        {type: 'number', description: "Sales", exclusiveMinimum: 0}
      ],
    },
  },
  requesterId: {type: 'string'}
},
required: ['instances', 'configuration', 'requesterId'],
}
```

```
response_schema: !jsonschema {
  type: 'array',
  items: {type: 'number'},
  description: "Estimated sales per conversion, in order corresponding to request order"
}
```


Developer experience

- Test suite
- Linting (pylint, mypy, bellybutton)
- Dockerization
- CI/CD
- Airflow DAG generation
- Training orchestration
- Automated model evaluation and promotion
- Gradual rollout

Developer experience

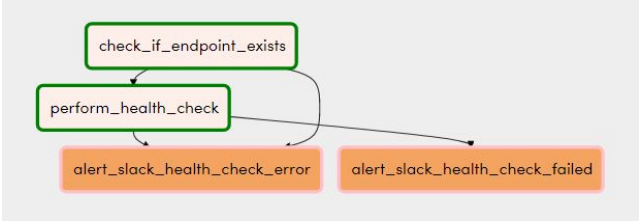
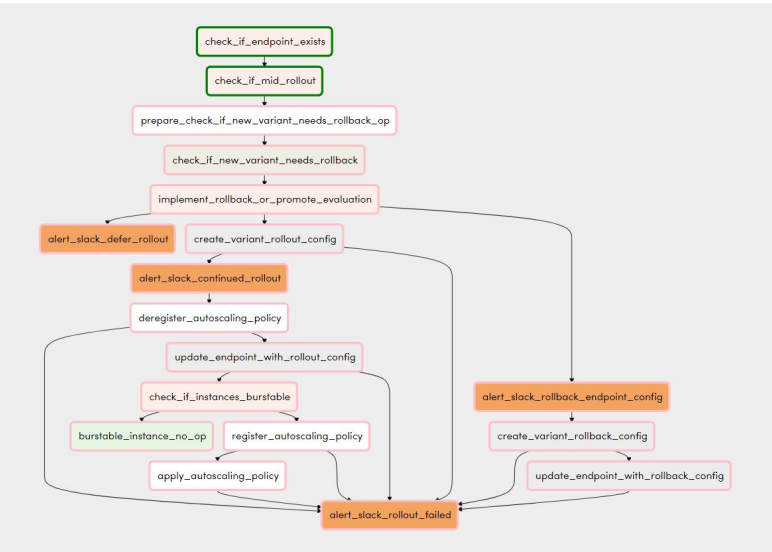
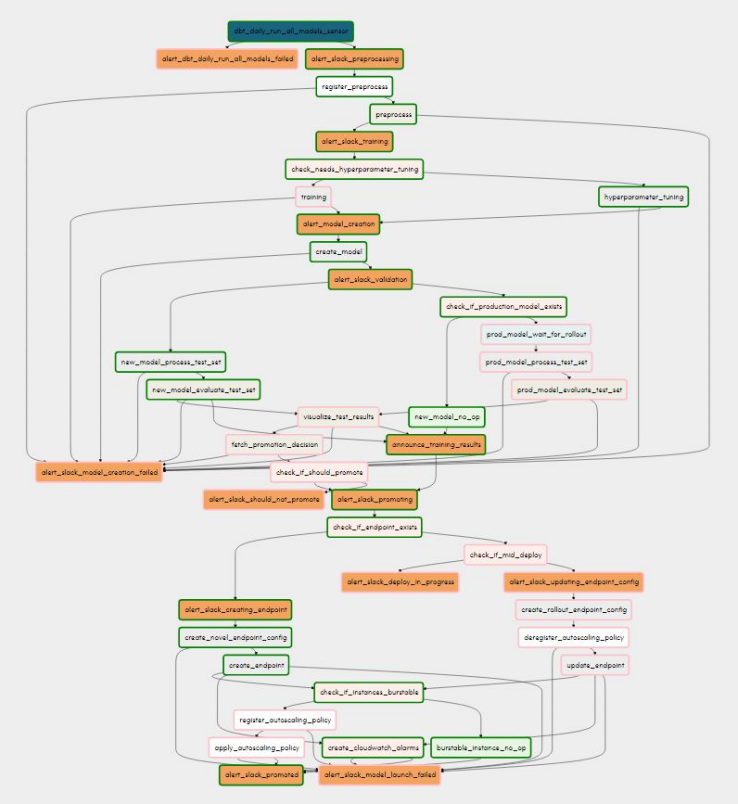
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- Autoscaling
- Schema validation
- Data capture
- Healthchecks
- Cost monitoring

Developer experience

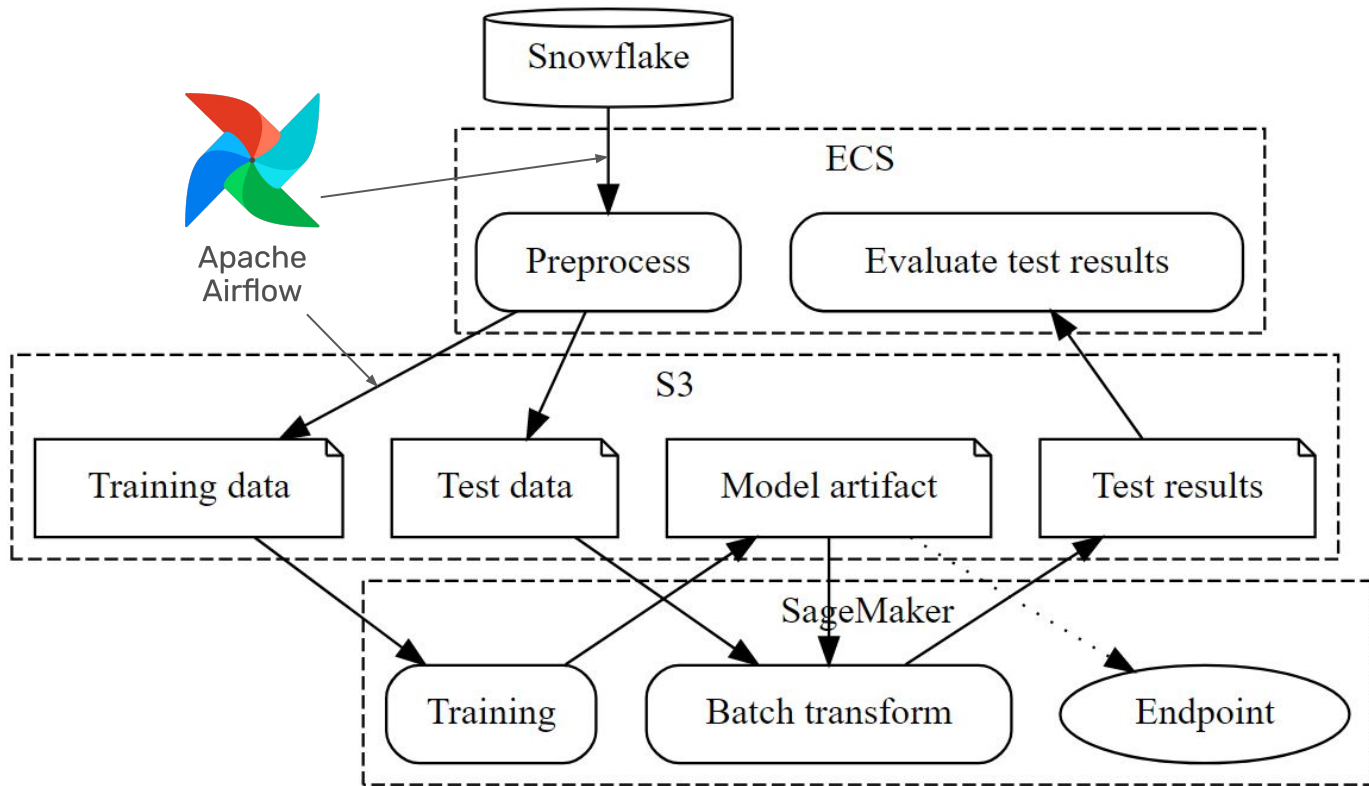
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III. Data scientists want to do data science

Developer experience



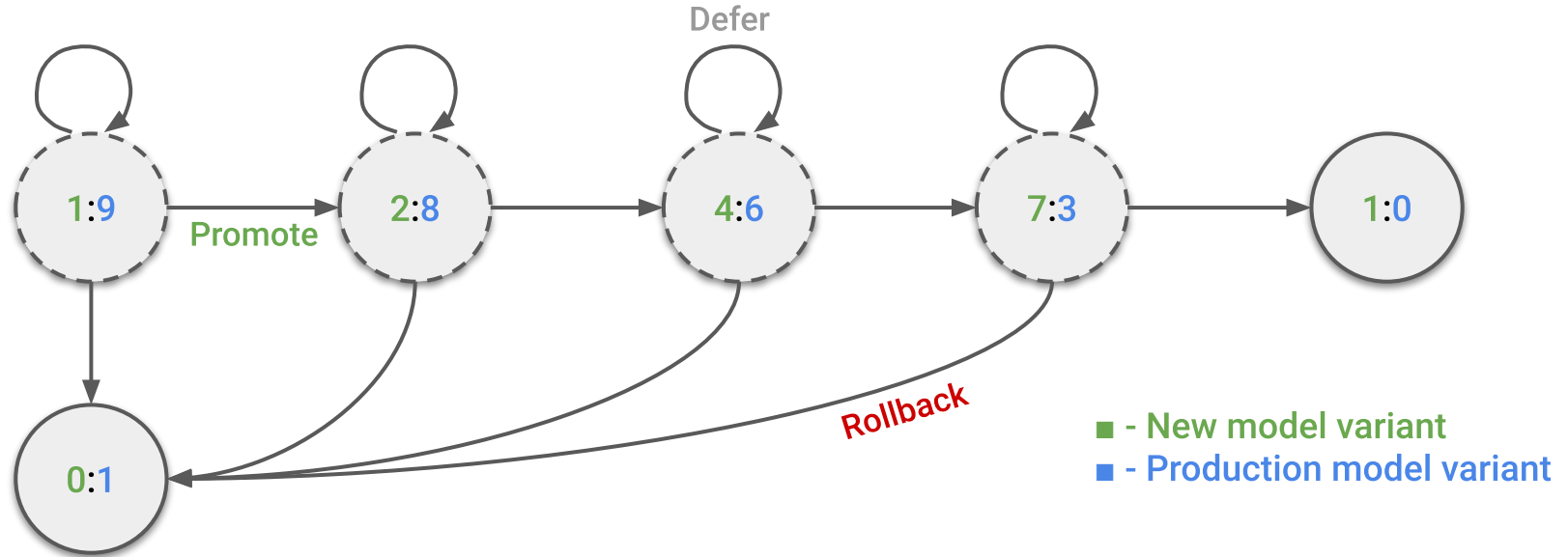
Our stack



Our stack

Technology	Application
AWS SageMaker	Model training, hosting; provenance info
Airflow (Astronomer.io)	Model lifecycle orchestration
Docker	Model packaging
Cookiecutter	Model repo templating
Jsonschema	Schema definition; PBT
Flask, gunicorn	Model server
DBT	Scalable data processing (in-warehouse)
Slack	Notifications, diagnostics
Pylint, mypy, bellybutton	Linting
Pytest, hypothesis, hypothesis-jsonschema	Test suite

Our stack



Lessons learned



SageMaker: sales-per-conversion-estimator [CREATE] APP 8:15 PM

Creating model.

Validation started.

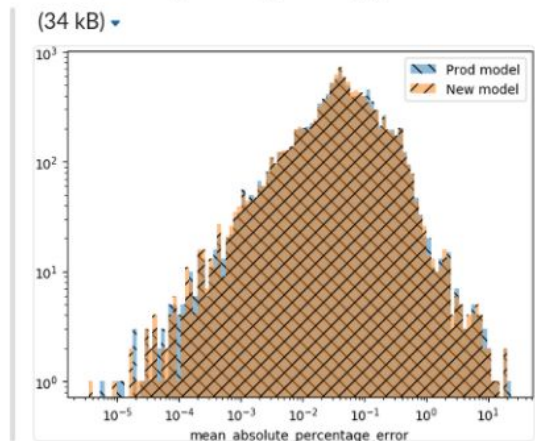


SageMaker: sales-per-conversion-estimator [CREATE] APP 8:24 PM

i Model ready for promotion. **i**

New model mean_absolute_percentage_error: 0.125717809901924

Production mean_absolute_percentage_error: 0.12795042672835955



Promote to endpoint

View new model job

View prod model job

Lessons learned



SageMaker: sales-per-conversion-estimator [CREATE] APP 8:15 PM

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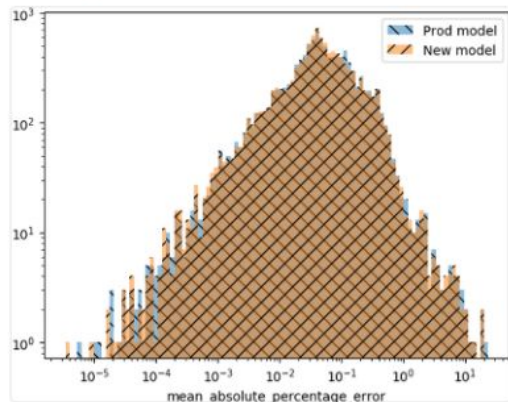
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(34 kB) ▾



Promote to endpoint

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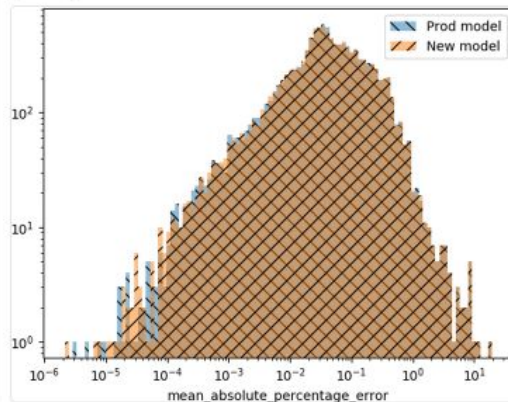
SageMaker: sales-per-conversion-estimator APP 8:25 PM

i Model ready for promotion. **i**

New model mean_absolute_percentage_error: 0.12105149215450292

Production mean_absolute_percentage_error: 0.12096298151072199

(34 kB) ▾



View new model job

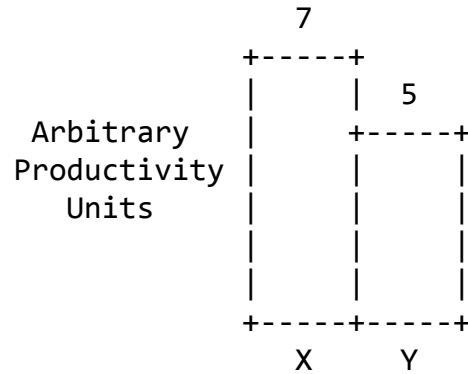
View prod model job

Promoting to endpoint.

Updating existing endpoint configuration.

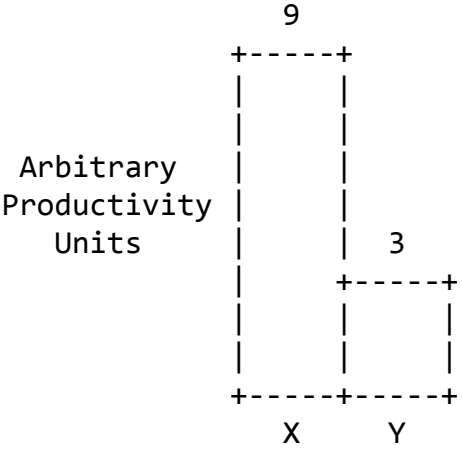
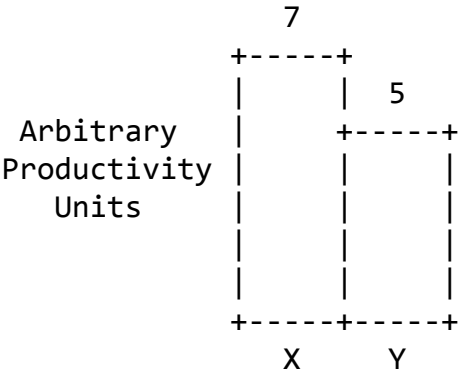
Lessons learned

"Best Practices" (whatever that means):



Lessons learned

"Best Practices" (whatever that means): "Worst Practices" (whatever that means):



Gonzales, G. (2016). Worst practices should be hard. <http://www.haskellforall.com/2016/04/worst-practices-should-be-hard.html>

Lessons learned

Instance type	vCPU	GPU	Mem (GiB)	GPU Mem (GiB)	Network Performance
Standard – Current Generation					
ml.t2.medium	2	-	4	-	Low to Moderate
ml.t2.large	2	-	8	-	Low to Moderate
ml.t2.xlarge	4	-	16	-	Moderate
ml.t2.2xlarge	8	-	32	-	Moderate
ml.t3.medium	2	-	4	-	Low to Moderate
ml.t3.large	2	-	8	-	Low to Moderate

Lessons learned

Instance type	VCPU	GPU	Mem (GB)	GPU Mem (GB)	Remark Performance
Standard - Current Generation					
m3.2xlarge	2	-	4	-	Low to Moderate
m3.xlarge	2	-	8	-	Low to Moderate
m3.large	4	-	16	-	Moderate
m3.2xlarge	8	-	32	-	Moderate
m3.xlarge	2	-	4	-	Low to Moderate
m3.large	2	-	8	-	Low to Moderate
m3.xlarge	4	-	16	-	Low to Moderate
m3.2xlarge	8	-	32	-	Low to Moderate
m4.xlarge	2	-	8	-	High
m4.xlarge	4	-	16	-	High
m4.xlarge	8	-	32	-	High
m4.xlarge	16	-	64	-	High
m4.xlarge	48	-	192	-	100+ ECU
m4.xlarge	96	-	384	-	25 ECU
m4.xlarge	4	-	16	-	High
m4.xlarge	16	-	64	-	High
m4.xlarge	48	-	192	-	10 ECU
m4.xlarge	144	-	456	-	25 ECU
m4.xlarge	2	-	4	-	High to 10 ECU
m4.xlarge	4	-	8	-	High to 10 ECU
m4.xlarge	8	-	16	-	High to 10 ECU
m4.xlarge	8	-	32	-	High to 10 ECU
m4.xlarge	16	-	64	-	High to 10 ECU
m4.xlarge	32	-	128	-	10 ECU
m4.xlarge	48	-	192	-	10 ECU
m4.xlarge	96	-	384	-	25 ECU
Memory Optimized - Current Generation					
m5.xlarge	2	-	16	-	High to 10 ECU
m5.xlarge	4	-	32	-	High to 10 ECU
m5.2xlarge	8	-	64	-	High to 10 ECU
m5.xlarge	16	-	128	-	High to 10 ECU
m5.2xlarge	48	-	384	-	10 ECU
m5.2xlarge	96	-	768	-	25 ECU
m5.xlarge	2	-	16	-	High to 10 ECU
m5.xlarge	4	-	32	-	High to 10 ECU
m5.2xlarge	8	-	64	-	High to 10 ECU
m5.xlarge	16	-	128	-	High to 10 ECU
m5.2xlarge	32	-	256	-	10 ECU
m5.2xlarge	48	-	384	-	10 ECU
m5.2xlarge	64	-	512	-	20 ECU
m5.2xlarge	96	-	768	-	25 ECU
Compute Optimized - Current Generation					
m6.xlarge	2	-	4	-	High to 10 ECU
m6.xlarge	4	-	8	-	High to 10 ECU
m6.2xlarge	8	-	16	-	High to 10 ECU
m6.xlarge	16	-	32	-	High to 10 ECU
m6.xlarge	32	-	64	-	10 ECU
m6.2xlarge	72	-	144	-	25 ECU
m6.xlarge	4	-	8	-	High to 10 ECU
m6.xlarge	8	-	16	-	High to 10 ECU
m6.xlarge	16	-	32	-	High to 10 ECU
m6.xlarge	36	-	72	-	10 ECU
m6.xlarge	72	-	144	-	25 ECU
m6.xlarge	2	-	4	-	Moderate
m6.xlarge	4	-	8	-	High
m6.xlarge	8	-	16	-	High
m6.xlarge	16	-	32	-	High
m6.xlarge	36	-	72	-	10 ECU
-	-	-	-	-	-
Accelerated Computing - Current Generation					
m7.xlarge	8	1x1280	61	16	High to 10 ECU
m7.xlarge	32	4x1280	244	64	100 ECU
m7.xlarge	64	8x1280	488	128	200 ECU
m7.xlarge	96	12x1280	732	204	100 ECU
m7.xlarge	4	1x640	31	72	High
m7.xlarge	12	4x640	124	16	10 ECU
m7.xlarge	34	10x640	372	52	25 ECU
m7.xlarge	4	1x714	16	16	High to 25 ECU
m7.xlarge	8	1x143	32	16	High to 25 ECU
m7.xlarge	16	1x714	64	16	High to 25 ECU
m7.xlarge	32	1x714	128	16	30 ECU
m7.xlarge	48	4x714	192	64	30 ECU
m7.xlarge	64	1x714	256	16	30 ECU

Lessons learned

Instance type	CPU	GPU	Mem (GiB)	GPU Mem (GiB)	Network Performance
Standard - Current Generation					
m5.2xlarge	2	-	4	-	Link to Marketplace
m5.xlarge	2	-	8	-	Link to Marketplace
m5.large	2	-	16	-	Marketplace
m5.2xlarge	8	-	32	-	Marketplace
m5.xlarge	2	-	4	-	Link to Marketplace
m5.large	2	-	8	-	Link to Marketplace
m5.xlarge	4	-	16	-	Link to Marketplace
m5.2xlarge	8	-	32	-	Link to Marketplace
m5n.xlarge	2	-	8	-	High
m5n.xlarge	4	-	16	-	High
m5n.xlarge	8	-	32	-	High
m5n.xlarge	16	-	64	-	High
m5n.2xlarge	48	-	192	-	100 Gbps
m5n.xlarge	96	-	384	-	25 Gbps
m5n.xlarge	4	-	16	-	High
m5n.xlarge	16	-	64	-	High
m5n.xlarge	40	-	160	-	10 Gbps
m5n.xlarge	140	-	560	-	25 Gbps
m5n.xlarge	2	-	8	-	100 to 10 Gbps
m5n.xlarge	4	-	16	-	10 to 10 Gbps
m5n.xlarge	8	-	32	-	100 to 10 Gbps
m5n.xlarge	16	-	64	-	10 to 10 Gbps
m5n.xlarge	32	-	128	-	10 Gbps
m5n.xlarge	48	-	192	-	10 Gbps
m5n.xlarge	96	-	384	-	25 Gbps
Memory Optimized - Current Generation					
m5.xlarge	2	-	16	-	Up to 10 Gbps
m5.xlarge	4	-	32	-	Up to 10 Gbps
m5.2xlarge	8	-	64	-	Up to 10 Gbps
m5.xlarge	16	-	128	-	Up to 10 Gbps
m5.2xlarge	48	-	384	-	10 Gbps
m5.4xlarge	96	-	768	-	25 Gbps
m5.xlarge	2	-	16	-	Up to 10 Gbps
m5.xlarge	4	-	32	-	Up to 10 Gbps
m5.2xlarge	8	-	64	-	Up to 10 Gbps
m5.xlarge	16	-	128	-	Up to 10 Gbps
m5.2xlarge	52	-	208	-	10 Gbps
m5n.xlarge	48	-	192	-	10 Gbps
m5n.xlarge	96	-	372	-	20 Gbps
m5n.xlarge	96	-	768	-	25 Gbps
Compute Optimized - Current Generation					
m5.xlarge	2	-	4	-	Up to 10 Gbps
m5.xlarge	4	-	8	-	Up to 10 Gbps
m5.2xlarge	8	-	16	-	Up to 10 Gbps
m5.xlarge	16	-	32	-	Up to 10 Gbps
m5n.xlarge	96	-	192	-	10 Gbps
m5.2xlarge	72	-	144	-	25 Gbps
m5n.xlarge	4	-	8	-	Up to 10 Gbps
m5n.xlarge	8	-	16	-	Up to 10 Gbps
m5n.xlarge	16	-	32	-	Up to 10 Gbps
m5n.xlarge	32	-	64	-	10 Gbps
m5n.xlarge	64	-	128	-	10 Gbps
m5n.xlarge	72	-	144	-	25 Gbps
m5n.xlarge	2	-	4	-	Marketplace
m5n.xlarge	4	-	7.5	-	High
m5n.xlarge	8	-	15	-	High
m5n.xlarge	16	-	30	-	High
m5n.xlarge	36	-	60	-	10 Gbps
Accelerated Computing - Current Generation					
m5g.xlarge	8	1xRTX 6000	61	16	Up to 10 Gbps
m5g.xlarge	16	4xRTX 6000	124	64	10 Gbps
m5g.xlarge	32	8xRTX 6000	248	128	25 Gbps
m5g.xlarge	96	4xRTX 6000	768	204	100 Gbps
m5g.xlarge	8	1xRTX 6000	61	16	High
m5g.xlarge	16	4xRTX 6000	124	64	10 Gbps
m5g.xlarge	32	8xRTX 6000	248	128	25 Gbps
m5g.xlarge	96	4xRTX 6000	768	204	100 Gbps
m5g.xlarge	8	1xRTX 6000	61	16	Up to 10 Gbps
m5g.xlarge	16	4xRTX 6000	124	64	Up to 25 Gbps
m5g.xlarge	32	8xRTX 6000	248	128	Up to 25 Gbps
m5g.xlarge	96	4xRTX 6000	768	204	Up to 25 Gbps
m5g.xlarge	16	4xRTX 6000	124	64	Up to 25 Gbps
m5g.xlarge	32	8xRTX 6000	248	128	Up to 25 Gbps
m5g.xlarge	96	4xRTX 6000	768	204	Up to 25 Gbps
m5g.xlarge	144	6xRTX 6000	1116	288	Up to 25 Gbps
m5g.xlarge	288	12xRTX 6000	2232	576	Up to 25 Gbps
m5g.xlarge	576	24xRTX 6000	4464	1152	Up to 25 Gbps
m5g.xlarge	1152	48xRTX 6000	8928	2304	Up to 25 Gbps
m5g.xlarge	2304	96xRTX 6000	17856	4608	Up to 25 Gbps
m5g.xlarge	4608	192xRTX 6000	35712	9216	Up to 25 Gbps
m5g.xlarge	9216	384xRTX 6000	71424	18432	Up to 25 Gbps
m5g.xlarge	18432	768xRTX 6000	142848	36864	Up to 25 Gbps

On-Demand ML Notebook Instances

Standard Instance - Current Generation	Price per Hour
m5.2xlarge	\$0.0664
m5.xlarge	\$0.1209
m5.2xlarge	\$0.2308
m5.xlarge	\$0.3187
m5.2xlarge	\$0.0562
m5.xlarge	\$0.1165
m5.2xlarge	\$0.215
m5.xlarge	\$0.4659
m5n.xlarge	\$0.28
m5n.2xlarge	\$0.56
m5n.4xlarge	\$1.12
m5n.8xlarge	\$2.80
m5n.16xlarge	\$4.48
m5n.32xlarge	\$8.96
m5n.64xlarge	\$17.92
m5n.128xlarge	\$35.84
m5n.256xlarge	\$71.68

Compute Optimized - Current Generation	Price per Hour
m5.xlarge	\$0.279
m5.2xlarge	\$0.617
m5.4xlarge	\$1.114
m5.8xlarge	\$2.227
m5.16xlarge	\$0.276
m5.32xlarge	\$0.476
m5.64xlarge	\$0.952
m5.128xlarge	\$1.742
m5.256xlarge	\$4.284
m5.512xlarge	\$8.285
m5.1024xlarge	\$16.56
m5.2048xlarge	\$33.12
m5.4096xlarge	\$66.24
m5.8192xlarge	\$132.48

Accelerated Computing - Current Generation	Price per Hour
m5g.xlarge	\$1.26
m5g.2xlarge	\$10.08
m5g.4xlarge	\$20.16
m5g.8xlarge	\$40.32
m5g.16xlarge	\$80.64
m5g.32xlarge	\$161.28
m5g.64xlarge	\$322.56

Amazon SageMaker Studio Notebook Instances

Standard Instance - Current Generation	Price per Hour
m5.xlarge	\$0.0554
m5.2xlarge	\$0.0746
m5.xlarge	\$0.0201
m5.2xlarge	\$0.0268
m5.xlarge	\$0.1165
m5.2xlarge	\$0.215
m5.4xlarge	\$0.4659
m5.8xlarge	\$0.154
m5.16xlarge	\$0.289
m5.32xlarge	\$0.558
m5.64xlarge	\$1.075
m5.128xlarge	\$2.150
m5.256xlarge	\$4.301
m5.512xlarge	\$8.601

Compute Optimized - Current Generation	Price per Hour
m5.xlarge	\$0.179
m5.2xlarge	\$0.236
m5.4xlarge	\$0.472
m5.8xlarge	\$0.944
m5.16xlarge	\$1.742
m5.32xlarge	\$3.484

Lessons learned

Instance type	CPU	GPU	Mem (GiB)	GPU Mem (GiB)	Network Performance
Standard - Current Generation					
m3.2xlarge	2	-	4	-	Low to Moderate
m3.xlarge	2	-	6	-	Low to Moderate
m3.large	2	-	10	-	Moderate
m3.xlarge	6	-	16	-	Moderate
m3.2xlarge	8	-	24	-	Moderate
m3.xlarge	2	-	4	-	Low to Moderate
m3.large	2	-	6	-	Low to Moderate
m3.xlarge	2	-	10	-	Low to Moderate
m3.2xlarge	2	-	12	-	Low to Moderate
m3.xlarge	2	-	8	-	High
m3.xlarge	2	-	10	-	High
m3.xlarge	4	-	16	-	High
m3.xlarge	6	-	24	-	High
m3.xlarge	10	-	34	-	High
m3.xlarge	16	-	48	-	High
m3.xlarge	16	-	64	-	High
m3.xlarge	40	-	100	-	High
m3.xlarge	48	-	200	-	High
m3.xlarge	2	-	8	-	Up to 10 Gbps
m3.xlarge	4	-	16	-	Up to 10 Gbps
m3.xlarge	8	-	32	-	Up to 10 Gbps
m3.xlarge	16	-	64	-	Up to 10 Gbps
m3.xlarge	32	-	128	-	Up to 10 Gbps
m3.xlarge	48	-	192	-	Up to 10 Gbps
m3.xlarge	96	-	384	-	Up to 10 Gbps
Memory Optimized - Current Generation					
m3.xlarge	2	-	16	-	Up to 10 Gbps
m3.xlarge	4	-	32	-	Up to 10 Gbps
m3.xlarge	8	-	64	-	Up to 10 Gbps
m3.xlarge	16	-	128	-	Up to 10 Gbps
m3.xlarge	48	-	384	-	Up to 10 Gbps
m3.xlarge	96	-	768	-	Up to 10 Gbps
m3.xlarge	2	-	16	-	Up to 10 Gbps
m3.xlarge	4	-	32	-	Up to 10 Gbps
m3.xlarge	8	-	64	-	Up to 10 Gbps
m3.xlarge	16	-	128	-	Up to 10 Gbps
m3.xlarge	48	-	384	-	Up to 10 Gbps
m3.xlarge	96	-	768	-	Up to 10 Gbps
Compute Optimized - Current Generation					
m3.xlarge	2	-	4	-	Up to 10 Gbps
m3.xlarge	4	-	8	-	Up to 10 Gbps
m3.xlarge	8	-	16	-	Up to 10 Gbps
m3.xlarge	16	-	32	-	Up to 10 Gbps
m3.xlarge	32	-	64	-	Up to 10 Gbps
m3.xlarge	72	-	144	-	Up to 10 Gbps
m3.xlarge	8	-	6	-	Up to 10 Gbps
m3.xlarge	16	-	12	-	Up to 10 Gbps
m3.xlarge	32	-	24	-	Up to 10 Gbps
m3.xlarge	48	-	36	-	Up to 10 Gbps
m3.xlarge	64	-	54	-	Up to 10 Gbps
m3.xlarge	96	-	81	-	Up to 10 Gbps
Accelerated Computing - Current Generation					
m3.xlarge	2	-	4	-	Up to 10 Gbps
m3.xlarge	4	-	8	-	Up to 10 Gbps
m3.xlarge	8	-	16	-	Up to 10 Gbps
m3.xlarge	16	-	32	-	Up to 10 Gbps
m3.xlarge	32	-	64	-	Up to 10 Gbps
m3.xlarge	72	-	144	-	Up to 10 Gbps
m3.xlarge	8	-	6	-	Up to 10 Gbps
m3.xlarge	16	-	12	-	Up to 10 Gbps
m3.xlarge	32	-	24	-	Up to 10 Gbps
m3.xlarge	48	-	36	-	Up to 10 Gbps
m3.xlarge	64	-	54	-	Up to 10 Gbps
m3.xlarge	96	-	81	-	Up to 10 Gbps
Standard Instances - Previous Generation					
m3.xlarge	2	-	4	-	Up to 10 Gbps
m3.xlarge	4	-	8	-	Up to 10 Gbps
m3.xlarge	8	-	16	-	Up to 10 Gbps
m3.xlarge	16	-	32	-	Up to 10 Gbps
m3.xlarge	32	-	64	-	Up to 10 Gbps
m3.xlarge	72	-	144	-	Up to 10 Gbps
m3.xlarge	8	-	6	-	Up to 10 Gbps
m3.xlarge	16	-	12	-	Up to 10 Gbps
m3.xlarge	32	-	24	-	Up to 10 Gbps
m3.xlarge	48	-	36	-	Up to 10 Gbps
m3.xlarge	64	-	54	-	Up to 10 Gbps
m3.xlarge	96	-	81	-	Up to 10 Gbps
Accelerated Computing - Previous Generation					
m3.xlarge	2	-	4	-	Up to 10 Gbps
m3.xlarge	4	-	8	-	Up to 10 Gbps
m3.xlarge	8	-	16	-	Up to 10 Gbps
m3.xlarge	16	-	32	-	Up to 10 Gbps
m3.xlarge	32	-	64	-	Up to 10 Gbps
m3.xlarge	72	-	144	-	Up to 10 Gbps
m3.xlarge	8	-	6	-	Up to 10 Gbps
m3.xlarge	16	-	12	-	Up to 10 Gbps
m3.xlarge	32	-	24	-	Up to 10 Gbps
m3.xlarge	48	-	36	-	Up to 10 Gbps
m3.xlarge	64	-	54	-	Up to 10 Gbps
m3.xlarge	96	-	81	-	Up to 10 Gbps

Standard Instances - Current Generation	Price per Hour
m3.2xlarge	\$0.0664
m3.xlarge	\$0.1299
m3.xlarge	\$0.2308
m3.2xlarge	\$0.3187
m3.xlarge	\$0.0582
m3.xlarge	\$0.1165
m3.xlarge	\$0.215
m3.2xlarge	\$0.4839
m3.xlarge	\$0.28
m3.xlarge	\$0.36
m3.xlarge	\$1.72
m3.xlarge	\$2.80
m3.xlarge	\$4.48
m3.xlarge	\$0.209
m3.xlarge	\$0.538
m3.xlarge	\$1.075
m3.2xlarge	\$1.226
m3.2xlarge	\$6.451

Compute Optimized - Current Generation	Price per Hour
m3.xlarge	\$0.279
m3.2xlarge	\$0.657
m3.xlarge	\$1.114
m3.2xlarge	\$1.227
m3.xlarge	\$0.278
m3.2xlarge	\$0.476
m3.xlarge	\$0.952
m3.2xlarge	\$2.142
m3.xlarge	\$4.284
m3.2xlarge	\$0.285
m3.2xlarge	\$0.558
m3.2xlarge	\$1.075
m3.2xlarge	\$1.419
m3.2xlarge	\$4.838

Accelerated Computing - Current Generation	Price per Hour
m3.xlarge	\$1.28
m3.2xlarge	\$10.88
m3.2xlarge	\$20.76
m3.2xlarge	\$4.284
m3.2xlarge	\$15.96
m3.2xlarge	\$42.272

Amazon SageMaker Studio Notebook Instance	Price per Hour
m3.xlarge	\$0.0054
m3.xlarge	\$0.0746
m3.xlarge	\$0.0201
m3.xlarge	\$0.0582
m3.xlarge	\$0.1165
m3.xlarge	\$0.215
m3.2xlarge	\$0.4839
m3.xlarge	\$0.154
m3.xlarge	\$0.289
m3.2xlarge	\$0.558
m3.xlarge	\$1.075
m3.2xlarge	\$2.150
m3.2xlarge	\$4.301
m3.2xlarge	\$6.451

Compute Optimized - Previous Generation	Price per Hour
m3.xlarge	\$0.179
m3.xlarge	\$0.238
m3.xlarge	\$0.476
m3.xlarge	\$0.952
m3.2xlarge	\$2.142
m3.xlarge	\$1.484

Lessons learned

Instance type	VCPU	GPU	Mem (GB)	GPU Mem (GB)	Network Performance
Standard - Current Generation					
m5.2xlarge	2	-	4	-	Up to 10 Gbps
m5.xlarge	2	-	8	-	Up to 10 Gbps
m5.large	2	-	16	-	Up to 10 Gbps
m5.2xlarge	8	-	32	-	Up to 10 Gbps
m5.xlarge	4	-	64	-	Up to 10 Gbps
m5.large	2	-	128	-	Up to 10 Gbps
m5a.2xlarge	2	-	4	-	Up to 10 Gbps
m5a.xlarge	2	-	8	-	Up to 10 Gbps
m5a.large	2	-	16	-	Up to 10 Gbps
m5a.2xlarge	8	-	32	-	Up to 10 Gbps
m5a.xlarge	4	-	64	-	Up to 10 Gbps
m5a.large	2	-	128	-	Up to 10 Gbps
m5ad.2xlarge	2	-	4	-	Up to 10 Gbps
m5ad.xlarge	2	-	8	-	Up to 10 Gbps
m5ad.large	2	-	16	-	Up to 10 Gbps
m5ad.2xlarge	8	-	32	-	Up to 10 Gbps
m5ad.xlarge	4	-	64	-	Up to 10 Gbps
m5ad.large	2	-	128	-	Up to 10 Gbps
m5ad.2xlarge	16	-	256	-	Up to 10 Gbps
m5ad.xlarge	8	-	512	-	Up to 10 Gbps
m5ad.large	4	-	1024	-	Up to 10 Gbps
m5d.2xlarge	16	-	256	-	Up to 10 Gbps
m5d.xlarge	8	-	512	-	Up to 10 Gbps
m5d.large	4	-	1024	-	Up to 10 Gbps
m5d.2xlarge	48	-	768	-	Up to 10 Gbps
m5d.xlarge	24	-	1536	-	Up to 10 Gbps
m5d.large	12	-	3072	-	Up to 10 Gbps
Memory Optimized - Current Generation					
m5n.2xlarge	2	-	16	-	Up to 10 Gbps
m5n.xlarge	4	-	32	-	Up to 10 Gbps
m5n.large	8	-	64	-	Up to 10 Gbps
m5n.2xlarge	16	-	128	-	Up to 10 Gbps
m5n.xlarge	8	-	256	-	Up to 10 Gbps
m5n.large	4	-	512	-	Up to 10 Gbps
m5n.2xlarge	48	-	768	-	Up to 10 Gbps
m5n.xlarge	24	-	1536	-	Up to 10 Gbps
m5n.large	12	-	3072	-	Up to 10 Gbps
Compute Optimized - Current Generation					
m5c.2xlarge	2	-	4	-	Up to 10 Gbps
m5c.xlarge	4	-	8	-	Up to 10 Gbps
m5c.large	8	-	16	-	Up to 10 Gbps
m5c.2xlarge	16	-	32	-	Up to 10 Gbps
m5c.xlarge	8	-	64	-	Up to 10 Gbps
m5c.large	4	-	128	-	Up to 10 Gbps
m5c.2xlarge	72	-	144	-	Up to 10 Gbps
m5c.xlarge	36	-	288	-	Up to 10 Gbps
m5c.large	18	-	576	-	Up to 10 Gbps
m5c.2xlarge	16	-	32	-	Up to 10 Gbps
m5c.xlarge	8	-	64	-	Up to 10 Gbps
m5c.large	4	-	128	-	Up to 10 Gbps
m5c.2xlarge	48	-	96	-	Up to 10 Gbps
m5c.xlarge	24	-	192	-	Up to 10 Gbps
m5c.large	12	-	384	-	Up to 10 Gbps
Accelerated Computing - Current Generation					
m5g.2xlarge	2	-	4	-	Up to 10 Gbps
m5g.xlarge	4	-	8	-	Up to 10 Gbps
m5g.large	8	-	16	-	Up to 10 Gbps
m5g.2xlarge	16	-	32	-	Up to 10 Gbps
m5g.xlarge	8	-	64	-	Up to 10 Gbps
m5g.large	4	-	128	-	Up to 10 Gbps
m5g.2xlarge	48	-	96	-	Up to 10 Gbps
m5g.xlarge	24	-	192	-	Up to 10 Gbps
m5g.large	12	-	384	-	Up to 10 Gbps
General Purpose - Current Generation					
m5g.2xlarge	2	-	4	-	Up to 10 Gbps
m5g.xlarge	4	-	8	-	Up to 10 Gbps
m5g.large	8	-	16	-	Up to 10 Gbps
m5g.2xlarge	16	-	32	-	Up to 10 Gbps
m5g.xlarge	8	-	64	-	Up to 10 Gbps
m5g.large	4	-	128	-	Up to 10 Gbps
m5g.2xlarge	72	-	144	-	Up to 10 Gbps
m5g.xlarge	36	-	288	-	Up to 10 Gbps
m5g.large	18	-	576	-	Up to 10 Gbps
m5g.2xlarge	16	-	32	-	Up to 10 Gbps
m5g.xlarge	8	-	64	-	Up to 10 Gbps
m5g.large	4	-	128	-	Up to 10 Gbps
m5g.2xlarge	48	-	96	-	Up to 10 Gbps
m5g.xlarge	24	-	192	-	Up to 10 Gbps
m5g.large	12	-	384	-	Up to 10 Gbps
Accelerated Computing - Current Generation					
m5g.2xlarge	8	-	16	-	Up to 10 Gbps
m5g.xlarge	16	-	32	-	Up to 10 Gbps
m5g.large	32	-	64	-	Up to 10 Gbps
m5g.2xlarge	64	-	128	-	Up to 10 Gbps
m5g.xlarge	32	-	256	-	Up to 10 Gbps
m5g.large	16	-	512	-	Up to 10 Gbps
m5g.2xlarge	48	-	96	-	Up to 10 Gbps
m5g.xlarge	24	-	192	-	Up to 10 Gbps
m5g.large	12	-	384	-	Up to 10 Gbps

instances: !instance
instance_count: 1
cpu: <0.25 vCPUs>
memory: <0.5 GB>
volume_size: <2 GB>

Standard Instances - Current Generation	Price per Hour
m5.2xlarge	\$0.0664
m5.xlarge	\$0.1209
m5.large	\$0.2308
m5.2xlarge	\$0.3107
m5.xlarge	\$0.0562
m5.large	\$0.1165
m5.2xlarge	\$0.215
m5.xlarge	\$0.4859
m5a.2xlarge	\$0.28
m5a.xlarge	\$0.56
m5a.2xlarge	\$1.72
m5a.xlarge	\$2.80
m5a.2xlarge	\$4.48
m5a.xlarge	\$0.209
m5a.2xlarge	\$0.558
m5a.xlarge	\$1.075
m5a.2xlarge	\$1.226
m5a.xlarge	\$6.851
Compute Optimized - Current Generation	
m5c.2xlarge	\$0.279
m5c.xlarge	\$0.557
m5c.large	\$1.114
m5c.2xlarge	\$1.227
m5c.xlarge	\$0.278
m5c.large	\$0.476
m5c.2xlarge	\$0.952
m5c.xlarge	\$1.742
m5c.large	\$4.284
m5c.2xlarge	\$0.205
m5c.xlarge	\$0.558
m5c.2xlarge	\$1.075
m5c.xlarge	\$1.819
m5c.2xlarge	\$4.838
Accelerated Computing - Current Generation	
m5g.2xlarge	\$1.26
m5g.xlarge	\$10.88
m5g.2xlarge	\$20.76
m5g.xlarge	\$4.204
m5g.2xlarge	\$17136
m5g.xlarge	\$42.272
Amazon SageMaker Studio Notebook Instance	
Standard Instances - Current Generation	
m5.xlarge	\$0.00564
m5.2xlarge	\$0.0746
m5.xlarge	\$0.0201
m5.2xlarge	\$0.0562
m5.xlarge	\$0.1165
m5.2xlarge	\$0.215
m5.2xlarge	\$0.4859
m5.xlarge	\$0.154
m5.2xlarge	\$0.269
m5.2xlarge	\$0.558
m5.2xlarge	\$1.075
m5.2xlarge	\$1.550
m5.2xlarge	\$1.226
m5.2xlarge	\$4.307
m5.2xlarge	\$6.851
Compute Optimized - Current Generation	
m5c.2xlarge	\$0.179
m5c.2xlarge	\$0.238
m5c.2xlarge	\$0.476
m5c.2xlarge	\$0.952
m5c.2xlarge	\$1.742
m5c.2xlarge	\$1.144
m5c.2xlarge	\$1.227

Lessons learned

Airflow:

- Hosting our own stack
- Deployment interruptions
- Not all contributions created equal



Questions?

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<https://www.teikametrics.com/company.html#careers>