

## Cloud Data Management with NetApp File-Object Duality and AWS SageMaker

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# Cloud Data Management with NetApp File-Object Duality and AWS SageMaker

## TR-4967: Cloud Data Management with NetApp File-Object Duality and AWS SageMaker

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Data scientists and engineers often need to access data stored in the NFS format, but accessing this data directly from the S3 protocol in AWS SageMaker can be challenging because AWS only supports S3 bucket access. However, NetApp ONTAP provides a solution by enabling dual-protocol access for NFS and S3. With this solution, data scientists and engineers can access NFS data from AWS SageMaker notebooks via S3 buckets from NetApp Cloud Volumes ONTAP. This approach enables easy access and sharing of the same data from both NFS and S3 without the need for additional software.

Next: Solution technology.

## Solution technology

Previous: Solution overview.

This solution utilizes the following technologies:

- AWS SageMaker Notebook. Offers machine learning capabilities to developers and data scientists to create, train, and deploy high-quality ML models efficiently.
- **NetApp BlueXP.** Enables the discovery, deployment, and operation of storage on premises as well as on AWS, Azure, and Google Cloud. It provides data protection against data loss, cyber threats, and unplanned outages and optimizes data storage and infrastructure.
- NetApp Cloud Volumes ONTAP. Provides enterprise-grade storage volumes with NFS, SMB/CIFS, iSCSI, and S3 protocols on AWS, Azure, and Google Cloud, giving users greater flexibility in accessing and managing their data in the cloud.

NetApp Cloud Volumes ONTAP created from BlueXP to store ML data.

The following figure shows the technical components of the solution.



#### **Use case summary**

A potential use case for dual protocol access of NFS and S3 is in the fields of machine learning and data science. For example, a team of data scientists might be working on a machine learning project using AWS SageMaker, which requires access to data stored in the NFS format. However, the data might also need to be accessed and shared via S3 buckets to collaborate with other team members or to integrate with other applications that use S3.

By utilizing NetApp Cloud Volumes ONTAP, the team can store their data in a single location and have it accessible with both NFS and S3 protocols. The data scientists can access the data in NFS format directly from AWS SageMaker, while other team members or applications can access the same data via S3 buckets.

This approach enables the data to be accessed and shared easily and efficiently without the need for additional software or data migration between different storage solutions. It also allows for a more streamlined workflow and collaboration among team members, resulting in faster and more effective development of machine learning models.

Next: Data duality for data scientists and other applications.

## Data duality for data scientists and other applications

Previous: Solution technology.

Data is available in NFS and accessed from S3 from AWS SageMaker.

### **Technology requirements**

You need NetApp BlueXP, NetApp Cloud Volumes ONTAP, and AWS SageMaker Notebooks for the data-duality use case.

#### Software requirements

The following table lists the software components that are required to implement the use case.

Software	Quantity
BlueXP	1

Software	Quantity
NetApp Cloud Volumes ONTAP	1
AWS SageMaker Notebook	1

#### **Deployment procedures**

Deploying the data-duality solution involves the following tasks:

- BlueXP Connector
- NetApp Cloud Volumes ONTAP
- · Data for machine learning
- · AWS SageMaker
- Validated machine learning from Jupyter Notebooks

#### BlueXP connector

In this validation, we used AWS. It's also applicable for Azure and Google Cloud. To create a BlueXP Connector in AWS, complete the following steps:

- 1. We used the credentials based on the mcarl-marketplace-subscription in BlueXP.
- 2. Choose the region suitable for your environment (for example, us-east-1 [N. Virginia]), and select the authentication method (for example, Assume Role or AWS keys). In this validation, we use AWS keys.
- 3. Provide the name of the connector and create a role.
- 4. Provide the network details such as the VPC, subnet, or keypair, depending on whether you need a public IP or not.
- 5. Provide the details for the security group, such as HTTP, HTTPS, or SSH access from the source type, such as anywhere and IP range information.
- 6. Review and create the BlueXP Connector.
- Verify that the BlueXP EC2 instance state is running in the AWS console, and check the IP address from the **Networking** tab.
- Log into the connector user interface from the BlueXP portal, or you can use the IP address for access from the browser.

#### **NetApp Cloud Volumes ONTAP**

To create a Cloud Volumes ONTAP instance in BlueXP, complete the following steps:

- 1. Create a new working environment, select the cloud provider, and select the type of Cloud Volumes ONTAP instance, (such as single-CVO, HA, or Amazon FSxN for ONTAP).
- 2. Provide details such as the Cloud Volumes ONTAP cluster name and credentials. In this validation, we created a Cloud Volumes ONTAP instance called svm sagemaker cvo sn1.
- 3. Select the services needed for Cloud Volumes ONTAP. In this validation, we choose to only monitor, so we disabled **Data Sense & Compliance** and **Backup to Cloud Services**.
- 4. In the **Location & Connectivity** section, select the AWS region, VPC, subnet, security group, SSH authentication method, and either a password or a key pair.

- 5. Choose the charging method. We used **Professional** for this validation.
- 6. You can choose a preconfigured package, such as **POC and Small Workloads**, **Database and Application Data Production Workloads**, **Cost Effective DR**, or **Highest Performance Production Workloads**. In this validation, we choose **Poc and Small Workloads**.
- 7. Create a volume with a specific size, allowed protocols, and export options. In this validation, we created a volume called vol1.
- 8. Choose a profile disk type and tiering policy. In this validation, we disabled **Storage Efficiency** and **General-Purpose SSD Dynamic Performance**.
- 9. Finally, review and create the Cloud Volumes ONTAP instance. Then wait for 15-20 minutes for BlueXP to create the Cloud Volumes ONTAP working environment.
- 10. Configure the following parameters to enable the Duality protocol. The Duality protocol (NFS/S3) is supported from ONTAP 9. 12.1 and later.
  - a. In this validation, we created an SVM called svm\_sagemaker\_cvo\_sn1 and volume vol1.
  - b. Verify that the SVM has the protocol support for NFS and S3. If not, modify the SVM to support them.

```
sagemaker cvo sn1::> vserver show -vserver svm sagemaker cvo sn1
                                    Vserver: svm sagemaker cvo sn1
                               Vserver Type: data
                            Vserver Subtype: default
                               Vserver UUID: 911065dd-a8bc-11ed-bc24-
e1c0f00ad86b
                                Root Volume:
svm sagemaker cvo sn1 root
                                  Aggregate: aggr1
                                 NIS Domain: -
                 Root Volume Security Style: unix
                                LDAP Client: -
               Default Volume Language Code: C.UTF-8
                            Snapshot Policy: default
                              Data Services: data-cifs, data-
flexcache,
                                             data-iscsi, data-nfs,
                                             data-nvme-tcp
                                    Comment:
                               Quota Policy: default
                List of Aggregates Assigned: aggr1
Limit on Maximum Number of Volumes allowed: unlimited
                        Vserver Admin State: running
                  Vserver Operational State: running
  Vserver Operational State Stopped Reason: -
                          Allowed Protocols: nfs, cifs, fcp, iscsi,
ndmp, s3
                       Disallowed Protocols: nyme
            Is Vserver with Infinite Volume: false
                           QoS Policy Group: -
                        Caching Policy Name: -
                                Confiq Lock: false
                               IPspace Name: Default
                         Foreground Process: -
                    Logical Space Reporting: true
                  Logical Space Enforcement: false
Default Anti ransomware State of the Vserver's Volumes: disabled
            Enable Analytics on New Volumes: false
   Enable Activity Tracking on New Volumes: false
sagemaker cvo sn1::>
```

- 11. Create and install a CA certificate if required.
- 12. Create a service data policy.

```
sagemaker cvo sn1::*> network interface service-policy create -vserver
svm sagemaker cvo sn1 -policy sagemaker s3 nfs policy -services data-
core, data-s3-server, data-nfs, data-flexcache
sagemaker cvo sn1::*> network interface create -vserver
svm sagemaker cvo sn1 -lif svm sagemaker cvo sn1 s3 lif -service-policy
sagemaker s3 nfs policy -home-node sagemaker cvo sn1-01 -address
172.30.10.41 -netmask 255.255.255.192
Warning: The configured failover-group has no valid failover targets for
the LIF's failover-policy. To view the failover targets for a LIF, use
        the "network interface show -failover" command.
sagemaker cvo sn1::*>
sagemaker cvo sn1::*> network interface show
Logical Status Network
                                      Current
                                                  Current Is
         Interface Admin/Oper Address/Mask Node Port
Vserver
Home
----
sagemaker cvo sn1
           cluster-mgmt up/up 172.30.10.40/26 sagemaker cvo sn1-
01
                                                                e0a
true
           intercluster up/up 172.30.10.48/26 sagemaker cvo sn1-
01
                                                                e0a
true
           sagemaker cvo sn1-01 mgmt1
                       up/up 172.30.10.58/26 sagemaker cvo sn1-
01
                                                                e0a
true
svm sagemaker cvo sn1
           svm sagemaker cvo sn1 data lif
                       up/up 172.30.10.23/26 sagemaker cvo sn1-
01
                                                                e0a
true
           svm sagemaker cvo sn1 mgmt lif
                       up/up 172.30.10.32/26 sagemaker cvo sn1-
01
                                                                e0a
true
           svm sagemaker cvo sn1 s3 lif
                                172.30.10.41/26
                        up/up
                                                  sagemaker cvo sn1-
```

```
01
                                                                    e0a
true
6 entries were displayed.
sagemaker cvo sn1::*>
sagemaker cvo sn1::*> vserver object-store-server create -vserver
svm sagemaker cvo snl -is-http-enabled true -object-store-server
svm_sagemaker_cvo_s3_sn1 -is-https-enabled false
sagemaker_cvo_sn1::*> vserver object-store-server show
Vserver: svm sagemaker cvo sn1
           Object Store Server Name: svm sagemaker cvo s3 sn1
               Administrative State: up
                       HTTP Enabled: true
             Listener Port For HTTP: 80
                      HTTPS Enabled: false
     Secure Listener Port For HTTPS: 443
  Certificate for HTTPS Connections: -
                  Default UNIX User: pcuser
               Default Windows User: -
                            Comment:
sagemaker_cvo_sn1::*>
```

#### 13. Check the aggregate details.

```
sagemaker cvo sn1::*> aggr show
Aggregate Size Available Used% State #Vols Nodes RAID
Status
_____ _____
-----
aggr0 sagemaker cvo sn1 01
       124.0GB 50.88GB 59% online 1 sagemaker cvo
raid0,
                                     sn1-01
normal
aggr1 907.1GB 904.9GB 0% online 2 sagemaker_cvo_
raid0,
                                     sn1-01
normal
2 entries were displayed.
sagemaker cvo sn1::*>
```

#### 14. Create a user and group.

```
sagemaker_cvo_sn1::*> vserver object-store-server user create -vserver
svm sagemaker cvo sn1 -user s3user
sagemaker cvo sn1::*> vserver object-store-server user show
Vserver User ID Access Key Secret Key
______
-----
svm sagemaker cvo sn1
          root
                      0 –
 Comment: Root User
svm sagemaker cvo sn1
          s3user 1
                               0ZNAX21JW5Q8AP80CQ2E
PpLs4gA9K0 2gPhuykkp014gBjcC9Rbi3QDX 6rr
2 entries were displayed.
sagemaker cvo sn1::*>
sagemaker cvo sn1::*> vserver object-store-server group create -name
s3group -users s3user -comment ""
sagemaker cvo sn1::*>
sagemaker cvo sn1::*> vserver object-store-server group delete -gid 1
-vserver svm sagemaker cvo snl
sagemaker cvo sn1::*> vserver object-store-server group create -name
s3group -users s3user -comment "" -policies FullAccess
sagemaker cvo sn1::*>
```

15. Create a bucket on the NFS volume.

#### AWS SageMaker

To create an AWS Notebook from AWS SageMaker, complete the following steps:

- Make sure the user who is creating Notebook instance has an AmazonSageMakerFullAccess IAM policy or is part of an existing group that has AmazonSageMakerFullAccess rights. In this validation, the user is part of an existing group.
- Provide the following information:
  - Notebook instance name.
  - Instance type.
  - · Platform identifier.
  - Select the IAM role that has AmazonSageMakerFullAccess rights.
  - Root access enable.
  - Encryption key Select no custom encryption.
  - Keep the remaining default options.
- 3. In this validation, the SageMaker instance details are as follows:





4. Start the AWS Notebook.



5. Open the Jupyter lab.



6. Log into the terminal and mount the Cloud Volumes ONTAP volume.

```
sh-4.2$ sudo mkdir /vol1; sudo mount -t nfs 172.30.10.41:/vol1 /vol1
sh-4.2$ df -h
                           Used Avail Use% Mounted on
Filesystem
                     Size
devtmpfs
                     2.0G
                              0
                                 2.0G
                                         0% /dev
tmpfs
                     2.0G
                              0
                                 2.0G
                                         0% /dev/shm
tmpfs
                                 2.0G
                                        1% /run
                     2.0G
                           624K
tmpfs
                     2.0G
                              0
                                 2.0G
                                         0% /sys/fs/cgroup
/dev/xvda1
                     140G
                           114G
                                  27G
                                        82% /
/dev/xvdf
                     4.8G
                            72K
                                 4.6G
                                        1% /home/ec2-user/SageMaker
tmpfs
                              0
                                 393M
                                        0% /run/user/1001
                     393M
                                       0% /run/user/1002
tmpfs
                     393M
                              0
                                 393M
                              0
                                 393M
                                        0% /run/user/1000
tmpfs
                     393M
172.30.10.41:/vol1
                                       20% /vol1
                     973M
                          189M
                                 785M
sh-4.2$
```

7. Check the bucket created on the Cloud Volumes ONTAP volume using the AWS CLI commands.

```
sh-4.2$ aws configure --profile netapp

AWS Access Key ID [None]: 0ZNAX21JW5Q8AP80CQ2E

AWS Secret Access Key [None]: PpLs4gA9K0_2gPhuykkp014gBjcC9Rbi3QDX_6rr

Default region name [None]: us-east-1

Default output format [None]:

sh-4.2$

sh-4.2$ aws s3 ls --profile netapp --endpoint-url

2023-02-10 17:59:48 ontapbucket1

sh-4.2$ aws s3 ls --profile netapp --endpoint-url s3://ontapbucket1/

2023-02-10 18:46:44 4747 1

2023-02-10 18:48:32 96 setup.cfg

sh-4.2$
```

#### Data for machine learning

In this validation, we used a dataset from DBpedia, a crowd-sourced community effort, to extract structured content from the information created in various Wikimedia projects.

1. Download the data from the DBpedia GitHub location and extract it. Use the same terminal used in the previous section.

```
sh-4.2$ wget
--2023-02-14 23:12:11--
Resolving github.com (github.com)... 140.82.113.3
Connecting to github.com (github.com) |140.82.113.3|:443... connected.
HTTP request sent, awaiting response... 302 Found
Location: [following]
--2023-02-14 23:12:11--
Resolving raw.githubusercontent.com (raw.githubusercontent.com)...
185.199.109.133, 185.199.110.133, 185.199.111.133, ...
Connecting to raw.githubusercontent.com
(raw.githubusercontent.com) | 185.199.109.133 | :443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 68431223 (65M) [application/octet-stream]
Saving to: 'dbpedia csv.tar.gz'
______
2023-02-14 23:12:13 (56.2 MB/s) - 'dbpedia csv.tar.gz' saved
[68431223/68431223]
sh-4.2$ tar -zxvf dbpedia csv.tar.gz
dbpedia csv/
dbpedia csv/test.csv
dbpedia csv/classes.txt
dbpedia csv/train.csv
dbpedia csv/readme.txt
sh-4.2$
```

Copy the data to the Cloud Volumes ONTAP location and check it from the S3 bucket using the AWS CLI.

```
sh-4.2$ df -h
Filesystem
                   Size Used Avail Use% Mounted on
                           0 2.0G 0% /dev
devtmpfs
                   2.0G
                           0 2.0G 0% /dev/shm
                   2.0G
tmpfs
                   2.0G 628K 2.0G 1% /run
tmpfs
tmpfs
                   2.0G
                           0 2.0G 0% /sys/fs/cgroup
/dev/xvda1
                   140G 114G 27G 82% /
/dev/xvdf
                   4.8G 52K 4.6G 1% /home/ec2-user/SageMaker
tmpfs
                   393M
                           0 393M 0% /run/user/1002
tmpfs
                   393M
                           0 393M 0% /run/user/1001
                           0 393M 0% /run/user/1000
tmpfs
                   393M
172.30.10.41:/vol1 973M 384K 973M 1% /vol1
sh-4.2$ pwd
/home/ec2-user
sh-4.2$ cp -ra dbpedia csv /vol1
sh-4.2$ aws s3 ls --profile netapp --endpoint-url s3://ontapbucket1/
                         PRE dbpedia csv/
2023-02-10 18:46:44
                         4747 1
2023-02-10 18:48:32
                          96 setup.cfg
sh-4.2$
```

3. Perform basic validation to make sure that read/write functionality works on the S3 bucket.

```
sh-4.2$ aws s3 cp --profile netapp --endpoint-url /usr/share/doc/util-
linux-2.30.2 s3://ontapbucket1/ --recursive
upload: ../../usr/share/doc/util-linux-2.30.2/deprecated.txt to
s3://ontapbucket1/deprecated.txt
upload: ../../usr/share/doc/util-linux-2.30.2/getopt-parse.bash to
s3://ontapbucket1/getopt-parse.bash
upload: ../../usr/share/doc/util-linux-2.30.2/README to
s3://ontapbucket1/README
upload: ../../usr/share/doc/util-linux-2.30.2/getopt-parse.tcsh to
s3://ontapbucket1/getopt-parse.tcsh
upload: ../../usr/share/doc/util-linux-2.30.2/AUTHORS to
s3://ontapbucket1/AUTHORS
upload: ../../usr/share/doc/util-linux-2.30.2/NEWS to
s3://ontapbucket1/NEWS
sh-4.2$ aws s3 ls --profile netapp --endpoint-url
s3://ontapbucket1/s3://ontapbucket1/
An error occurred (InternalError) when calling the ListObjectsV2
operation: We encountered an internal error. Please try again.
sh-4.2$ aws s3 ls --profile netapp --endpoint-url s3://ontapbucket1/
                          PRE dbpedia csv/
2023-02-16 19:19:27
                        26774 AUTHORS
```

```
2023-02-16 19:19:27
                        72727 NEWS
2023-02-16 19:19:27
                        4493 README
2023-02-16 19:19:27
                         2825 deprecated.txt
2023-02-16 19:19:27
                         1590 getopt-parse.bash
                         2245 getopt-parse.tcsh
2023-02-16 19:19:27
sh-4.2$ ls -ltr /vol1
total 132
drwxrwxr-x 2 ec2-user ec2-user 4096 Mar 29 2015 dbpedia csv
-rw-r--r- 1 nobody nobody 2245 Apr 10 17:37 getopt-parse.tcsh
-rw-r--r-- 1 nobody nobody 2825 Apr 10 17:37 deprecated.txt
-rw-r--r-- 1 nobody nobody 4493 Apr 10 17:37 README
-rw-r--r-- 1 nobody nobody
                              1590 Apr 10 17:37 getopt-parse.bash
-rw-r--r-- 1 nobody nobody 26774 Apr 10 17:37 AUTHORS
-rw-r--r-- 1 nobody nobody 72727 Apr 10 17:37 NEWS
sh-4.2$ ls -ltr /vol1/dbpedia csv/
total 192104
-rw----- 1 ec2-user ec2-user 174148970 Mar 28 2015 train.csv
-rw----- 1 ec2-user ec2-user 21775285 Mar 28 2015 test.csv
-rw----- 1 ec2-user ec2-user 146 Mar 28 2015 classes.txt
-rw-rw-r-- 1 ec2-user ec2-user 1758 Mar 29 2015 readme.txt
sh-4.2$ chmod -R 777 /vol1/dbpedia csv
sh-4.2$ ls -ltr /vol1/dbpedia csv/
total 192104
-rwxrwxrwx 1 ec2-user ec2-user 174148970 Mar 28 2015 train.csv
-rwxrwxrwx 1 ec2-user ec2-user 21775285 Mar 28 2015 test.csv
-rwxrwxrwx 1 ec2-user ec2-user 146 Mar 28 2015 classes.txt
-rwxrwxrwx 1 ec2-user ec2-user 1758 Mar 29 2015 readme.txt
sh-4.2$ aws s3 cp --profile netapp --endpoint-url http://172.30.2.248/
s3://ontapbucket1/ /tmp --recursive
download: s3://ontapbucket1/AUTHORS to ../../tmp/AUTHORS
download: s3://ontapbucket1/README to ../../tmp/README
download: s3://ontapbucket1/NEWS to ../../tmp/NEWS
download: s3://ontapbucket1/dbpedia csv/classes.txt to
../../tmp/dbpedia csv/classes.txt
download: s3://ontapbucket1/dbpedia csv/readme.txt to
../../tmp/dbpedia csv/readme.txt
download: s3://ontapbucket1/deprecated.txt to ../../tmp/deprecated.txt
download: s3://ontapbucket1/getopt-parse.bash to ../../tmp/getopt-
parse.bash
download: s3://ontapbucket1/getopt-parse.tcsh to ../../tmp/getopt-
parse.tcsh
download: s3://ontapbucket1/dbpedia csv/test.csv to
../../tmp/dbpedia csv/test.csv
download: s3://ontapbucket1/dbpedia csv/train.csv to
../../tmp/dbpedia csv/train.csv
sh-4.2$
```

#### Validate machine learning from Jupyter Notebooks

The following validation provides the machine-learning build, train, and deploy models through text classification by using the SageMaker BlazingText example below:

1. Install the boto3 and SageMaker packages.

```
In [1]: pip install --upgrade boto3 sagemaker
```

#### Output:

```
Looking in indexes: https://pypi.org/simple,
https://pip.repos.neuron.amazo naws.com
Requirement already satisfied: boto3 in /home/ec2-
user/anaconda3/envs/pytho n3/lib/python3.10/site-packages (1.26.44)
Collecting boto3
 Downloading boto3-1.26.72-py3-none-any.whl (132 kB)
    0: 00:00
Requirement already satisfied: sagemaker in /home/ec2-
user/anaconda3/envs/p ython3/lib/python3.10/site-packages (2.127.0)
Collecting sagemaker
 Downloading sagemaker-2.132.0.tar.gz (668 kB)
    0:
00:0000:01
 Preparing metadata (setup.py) ... done
Collecting botocore<1.30.0,>=1.29.72
 Downloading botocore-1.29.72-py3-none-any.whl (10.4 MB)
    0: 00:0000:010:01
Requirement already satisfied: s3transfer<0.7.0,>=0.6.0 in /home/ec2-
user/a naconda3/envs/python3/lib/python3.10/site-packages (from boto3)
(0.6.0)
Requirement already satisfied: jmespath<2.0.0,>=0.7.1 in /home/ec2-
```

```
user/ana conda3/envs/python3/lib/python3.10/site-packages (from boto3)
(0.10.0)
Requirement already satisfied: attrs<23,>=20.3.0 in /home/ec2-
user/anaconda
3/envs/python3/lib/python3.10/site-packages (from sagemaker) (22.1.0)
Requirement already satisfied: google-pasta in /home/ec2-
user/anaconda3/env s/python3/lib/python3.10/site-packages (from
sagemaker) (0.2.0)
Requirement already satisfied: numpy<2.0,>=1.9.0 in /home/ec2-
user/anaconda
3/envs/python3/lib/python3.10/site-packages (from sagemaker) (1.22.4)
Requirement already satisfied: protobuf<4.0,>=3.1 in /home/ec2-
user/anacond a3/envs/python3/lib/python3.10/site-packages (from
sagemaker) (3.20.3)
Requirement already satisfied: protobuf3-to-dict<1.0,>=0.1.5 in
/home/ec2-u ser/anaconda3/envs/python3/lib/python3.10/site-packages
(from sagemaker)
(0.1.5)
Requirement already satisfied: smdebug rulesconfig==1.0.1 in /home/ec2-
use r/anaconda3/envs/python3/lib/python3.10/site-packages (from
sagemaker) (1.
0.1) Requirement already satisfied: importlib-metadata<5.0,>=1.4.0 in
/home/ec2user/anaconda3/envs/python3/lib/python3.10/site-packages (from
sagemaker)
(4.13.0)
Requirement already satisfied: packaging>=20.0 in /home/ec2-
user/anaconda3/ envs/python3/lib/python3.10/site-packages (from
sagemaker) (21.3)
Requirement already satisfied: pandas in /home/ec2-
user/anaconda3/envs/pyth on3/lib/python3.10/site-packages (from
sagemaker) (1.5.1)
Requirement already satisfied: pathos in /home/ec2-
user/anaconda3/envs/pyth on3/lib/python3.10/site-packages (from
sagemaker) (0.3.0)
Requirement already satisfied: schema in /home/ec2-
user/anaconda3/envs/pyth on3/lib/python3.10/site-packages (from
sagemaker) (0.7.5) Requirement already satisfied: python-
dateutil<3.0.0,>=2.1 in /home/ec2-use
r/anaconda3/envs/python3/lib/python3.10/site-packages (from
botocore<1.30.
0, >=1.29.72 -> boto3) (2.8.2)
Requirement already satisfied: urllib3<1.27,>=1.25.4 in /home/ec2-
user/anac onda3/envs/python3/lib/python3.10/site-packages (from
botocore<1.30.0,>=1.2
9.72->boto3) (1.26.8) Requirement already satisfied: zipp>=0.5 in
/home/ec2-user/anaconda3/envs/p ython3/lib/python3.10/site-packages
```

```
(from importlib-metadata<5.0,>=1.4.0->s agemaker) (3.10.0)
Requirement already satisfied: pyparsing!=3.0.5,>=2.0.2 in /home/ec2-
user/a naconda3/envs/python3/lib/python3.10/site-packages (from
packaging>=20.0->s agemaker) (3.0.9)
Requirement already satisfied: six in /home/ec2-
user/anaconda3/envs/python
3/lib/python3.10/site-packages (from protobuf3-to-dict<1.0,>=0.1.5-
>sagemak er) (1.16.0)
Requirement already satisfied: pytz>=2020.1 in /home/ec2-
user/anaconda3/env s/python3/lib/python3.10/site-packages (from pandas-
>sagemaker) (2022.5)
Requirement already satisfied: ppft>=1.7.6.6 in /home/ec2-
user/anaconda3/en vs/python3/lib/python3.10/site-packages (from pathos-
>sagemaker) (1.7.6.6) Requirement already satisfied:
multiprocess>=0.70.14 in /home/ec2-user/anac
onda3/envs/python3/lib/python3.10/site-packages (from pathos->sagemaker)
(0.70.14)
Requirement already satisfied: dill>=0.3.6 in /home/ec2-
user/anaconda3/env s/python3/lib/python3.10/site-packages (from pathos-
>sagemaker) (0.3.6)
Requirement already satisfied: pox>=0.3.2 in /home/ec2-
user/anaconda3/envs/ python3/lib/python3.10/site-packages (from pathos-
>sagemaker) (0.3.2) Requirement already satisfied: contextlib2>=0.5.5 in
/home/ec2-user/anacond a3/envs/python3/lib/python3.10/site-packages
(from schema->sagemaker) (21.
6.0) Building wheels for collected packages: sagemaker
  Building wheel for sagemaker (setup.py) ... done
  Created wheel for sagemaker: filename=sagemaker-2.132.0-py2.py3-none-
any. whl size=905449
sha256=f6100a5dc95627f2e2a49824e38f0481459a27805ee19b5a06ec
83db0252fd41
  Stored in directory: /home/ec2-
user/.cache/pip/wheels/60/41/b6/482e7ab096
520df034fbf2dddd244a1d7ba0681b27ef45aa61
Successfully built sagemaker
Installing collected packages: botocore, boto3, sagemaker
 Attempting uninstall: botocore Found existing installation:
botocore 1.24.19
    Uninstalling botocore-1.24.19:
                                         Successfully uninstalled
botocore-1.24.19
  Attempting uninstall: boto3 Found existing installation: boto3
1.26.44
    Uninstalling boto3-1.26.44:
      Successfully uninstalled boto3-1.26.44
  Attempting uninstall: sagemaker Found existing installation:
sagemaker 2.127.0
```

Uninstalling sagemaker-2.127.0:

Successfully uninstalled sagemaker-2.127.0

ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source of the following dependency conflicts.

awscli 1.27.44 requires botocore==1.29.44, but you have botocore 1.29.72 wh ich is incompatible.

aiobotocore 2.0.1 requires botocore<1.22.9,>=1.22.8, but you have botocore 1.29.72 which is incompatible. Successfully installed boto3-

1.26.72 botocore-1.29.72 sagemaker-2.132.0 Note: you may need to restart the kernel to use updated packages.

2. In the following step, the data (dbpedia\_csv) is downloaded from the s3 bucket ontapbucket1 to a Jupyter Notebook instance used in machine learning.

```
In [2]: import sagemaker
In [3]: from sagemaker import get execution role
In [4]:
import json
import boto3
sess = sagemaker.Session()
role = get execution role()
print(role)
bucket = "ontapbucket1"
print(bucket)
sess.s3 client = boto3.client('s3',region name='',aws access key id =
'0ZNAX21JW5Q8AP80CQ2E', aws secret access key =
'PpLs4qA9K0 2qPhuykkp014qBjcC9Rbi3QDX 6rr',
                              use ssl = False, endpoint url =
'http://172.30.10.41',
config=boto3.session.Config(signature version='s3v4',
s3={'addressing style':'path'}) )
sess.s3 resource = boto3.resource('s3',region name='',aws access key id
= '0ZNAX21JW5Q8AP80CQ2E', aws secret access key =
'PpLs4gA9K0 2gPhuykkp014gBjcC9Rbi3QDX 6rr',
                              use ssl = False, endpoint url =
'http://172.30.10.41',
config=boto3.session.Config(signature version='s3v4',
s3={'addressing style':'path'}) )
prefix = "blazingtext/supervised"
import os
my bucket = sess.s3 resource.Bucket(bucket)
my bucket = sess.s3 resource.Bucket(bucket)
#os.mkdir('dbpedia csv')
for s3_object in my bucket.objects.all():
    filename = s3 object.key
#
   print(filename)
    print(s3 object.key)
    my bucket.download file(s3 object.key, filename)
```

3. The following code creates the mapping from integer indices to class labels that are used to retrieve the actual class name during inference.

```
index_to_label = {}
with open("dbpedia_csv/classes.txt") as f:
   for i,label in enumerate(f.readlines()):
      index_to_label[str(i + 1)] = label.strip()
```

The output lists the files and folders in the ontapbucket1 bucket that are used as data for the AWS SageMaker machine-learning validation.

```
arn:aws:iam::210811600188:role/SageMakerFullRole ontapbucket1
AUTHORS
AUTHORS
NEWS
NEWS
README README
dbpedia csv/classes.txt dbpedia csv/classes.txt dbpedia csv/readme.txt
dbpedia csv/readme.txt dbpedia csv/test.csv dbpedia csv/test.csv
dbpedia csv/train.csv dbpedia csv/train.csv deprecated.txt
deprecated.txt getopt-parse.bash getopt-parse.tcsh
getopt-parse.tcsh
In [5]: ls
AUTHORS
              deprecated.txt getopt-parse.tcsh NEWS
Untitled.ipynb dbpedia csv/ getopt-parse.bash lost+found/
README
In [6]: ls -l dbpedia_csv
total 191344
-rw-rw-r-- 1 ec2-user ec2-user 146 Feb 16 19:43 classes.txt
-rw-rw-r-- 1 ec2-user ec2-user 1758 Feb 16 19:43 readme.txt
-rw-rw-r-- 1 ec2-user ec2-user 21775285 Feb 16 19:43 test.csv
-rw-rw-r-- 1 ec2-user ec2-user 174148970 Feb 16 19:43 train.csv
```

4. Start the data preprocessing phase to preprocess the training data into a space-separated, tokenized text format that can be consumed by the BlazingText algorithm and the nltk library to tokenize the input sentences from the DBPedia dataset. Download the nltk tokenizer and other libraries. The transform instance applied to each data instance in parallel uses the Python multiprocessing module.

```
ln [7]: from random import shuffle
import multiprocessing
from multiprocessing import Pool
import csv
import nltk
nltk.download("punkt")
def transform_instance(row):
    cur_row = []
    label = "__label__" + index_to_label [row[0]] # Prefix the index-ed
label with __label__
    cur_row.append (label)
    cur_row.extend(nltk.word_tokenize(row[1].lower ()))
    cur_row.extend(nltk.word_tokenize(row[2].lower ()))
    return cur_row
def preprocess(input_file, output_file, keep=1):
```

```
all rows = []
    with open(input file, "r") as csvinfile:
        csv reader = csv.reader(csvinfile, delimiter=",")
        for row in csv reader:
            all rows.append(row)
    shuffle(all rows)
    all rows = all rows[: int(keep * len(all rows))]
    pool = Pool(processes=multiprocessing.cpu count())
    transformed rows = pool.map(transform instance, all rows)
    pool.close()
    pool. join()
    with open(output file, "w") as csvoutfile:
        csv writer = csv.writer (csvoutfile, delimiter=" ",
lineterminator="\n")
        csv writer.writerows (transformed rows)
# Preparing the training dataset
# since preprocessing the whole dataset might take a couple of minutes,
# we keep 20% of the training dataset for this demo.
# Set keep to 1 if you want to use the complete dataset
preprocess("dbpedia csv/train.csv", "dbpedia.train", keep=0.2)
# Preparing the validation dataset
preprocess("dbpedia csv/test.csv", "dbpedia.validation")
sess = sagemaker.Session()
role = get execution role()
print (role) # This is the role that sageMaker would use to leverage Aws
resources (S3, Cloudwatch) on your behalf
bucket = sess.default bucket() # Replace with your own bucket name if
needed
print("default Bucket::: ")
print(bucket)
```

#### Output:

```
[nltk_data] Downloading package punkt to /home/ec2-user/nltk_data...
[nltk_data] Package punkt is already up-to-date!
arn:aws:iam::210811600188:role/SageMakerFullRole default Bucket:::
sagemaker-us-east-1-210811600188
```

5. Upload the formatted and training dataset to S3 so that it can be used by SageMaker to execute training jobs. Then upload two files to the bucket and prefix location using the Python SDK.

```
ln [8]: %%time
train_channel = prefix + "/train"
validation_channel = prefix + "/validation"
sess.upload_data(path="dbpedia.train", bucket=bucket,
key_prefix=train_channel)
sess.upload_data(path="dbpedia.validation", bucket=bucket,
key_prefix=validation_channel)
s3_train_data = "s3://{}/{}".format(bucket, train_channel)
s3_validation_data = "s3://{}/{}".format(bucket, validation_channel)
```

#### Output:

```
CPU times: user 546 ms, sys: 163 ms, total: 709 ms
Wall time: 1.32 s
```

6. Set up an output location at S3 where the model artifact is loaded so that artifacts can be the output of the algorithm's training job. Create a sageMaker.estimator.Estimator object to launch the training job.

```
In [9]: s3_output_location = "s3://{}/{}/output".format(bucket, prefix)
In [10]: region_name = boto3.Session().region_name
In [11]: container =
sagemaker.amazon.amazon_estimator.get_image_uri(region_name,
"blazingtext","latest")
print("Using SageMaker BlazingText container: {} ({})".format(container, region_name))
```

#### Output:

```
The method get_image_uri has been renamed in sagemaker>=2.
See: https://sagemaker.readthedocs.io/en/stable/v2.html for details.
Defaulting to the only supported framework/algorithm version: 1.
Ignoring f ramework/algorithm version: latest.
Using SageMaker BlazingText container: 811284229777.dkr.ecr.us-east-1.amazo naws.com/blazingtext:1 (us-east-1)
```

7. Define the SageMaker Estrimator with resource configurations and hyperparameters to train text classification on the DBPedia dataset using the supervised mode on a c4.4xlarge instance.

```
In [12]: bt model = sagemaker.estimator.Estimator(
container,
role,
instance count=1,
instance type="ml.c4.4xlarge",
volume size=30,
max run=360000,
input mode="File",
output path=s3 output location,
hyperparameters={
        "mode": "supervised",
        "epochs": 1,
        "min count": 2,
        "learning rate": 0.05,
        "vector dim": 10,
        "early stopping": True,
        "patience": 4,
        "min epochs": 5,
        "word ngrams": 2,
 },
```

8. Prepare a handshake between the data channels and the algorithm. To do this, create the sagemaker.session.s3\_input objects from the data channels, and keep them in a dictionary for the algorithm to consume.

```
In [13]: train_data = sagemaker.inputs.TrainingInput(
    s3_train_data,
    distribution="FullyReplicated",
    content_type="text/plain",
    s3_data_type="S3Prefix",
)

validation_data = sagemaker.inputs.TrainingInput(
    s3_validation_data,
    distribution="FullyReplicated",
    content_type="text/plain",
    s3_data_type="S3Prefix",
)
data_channels = {"train": train_data, "validation": validation_data}
```

9. After the job has finished, a Job Complete message appears. The trained model can be found in the S3 bucket that was set up as the output\_path in the estimator.

```
ln [14]: bt model.fit(inputs=data_channels, logs=True)
```

#### Output:

```
INFO: sagemaker: Creating training-job with name: blazingtext-2023-02-16-
20-3
7-30-748
2023-02-16 20:37:30 Starting - Starting the training job.....
2023-02-16 20:38:09 Starting - Preparing the instances for
training.....
2023-02-16 20:39:24 Downloading - Downloading input data
2023-02-16 20:39:24 Training - Training image download completed.
Training in progress... Arguments: train
[02/16/2023 20:39:41 WARNING 140279908747072] Loggers have already been
set up. [02/16/2023 20:39:41 WARNING 140279908747072] Loggers have
already been set up.
[02/16/2023 20:39:41 INFO 140279908747072] nvidia-smi took:
0.0251793861389
16016 secs to identify 0 gpus
[02/16/2023 20:39:41 INFO 140279908747072] Running single machine CPU
Blazi ngText training using supervised mode.
Number of CPU sockets found in instance is 1
[02/16/2023 20:39:41 INFO 140279908747072] Processing
/opt/ml/input/data/tr ain/dbpedia.train . File size: 35.0693244934082 MB
[02/16/2023 20:39:41 INFO 140279908747072] Processing
/opt/ml/input/data/va lidation/dbpedia.validation . File size:
21.887572288513184 MB
Read 6M words
Number of words: 149301
Loading validation data from
/opt/ml/input/data/validation/dbpedia.validati on
Loaded validation data.
----- End of epoch: 1 ##### Alpha: 0.0000 Progress: 100.00%
Million Words/sec: 10.39 ##### Training finished.
Average throughput in Million words/sec: 10.39
Total training time in seconds: 0.60
#train accuracy: 0.7223
Number of train examples: 112000
#validation accuracy: 0.7205
Number of validation examples: 70000
2023-02-16 20:39:55 Uploading - Uploading generated training model
2023-02-16 20:40:11 Completed - Training job completed
Training seconds: 68
Billable seconds: 68
```

10. After training is complete, deploy the trained model as an Amazon SageMaker real-time hosted endpoint to make predictions.

```
In [15]: from sagemaker.serializers import JSONSerializer
  text_classifier = bt_model.deploy(
        initial_instance_count=1, instance_type="ml.m4.xlarge",
        serializer=JSONS
)
```

#### Output:

```
INFO:sagemaker:Creating model with name: blazingtext-2023-02-16-20-41-33-10

INFO:sagemaker:Creating endpoint-config with name blazingtext-2023-02-16-20
-41-33-100
INFO:sagemaker:Creating endpoint with name blazingtext-2023-02-16-20-41-33-
100
-----!
```

```
In [16]: sentences = [
    "Convair was an american aircraft manufacturing company which later
expanded into rockets and spacecraft.",
    "Berwick secondary college is situated in the outer melbourne
metropolitan suburb of berwick .",
]
# using the same nltk tokenizer that we used during data preparation for
training
tokenized_sentences = [" ".join(nltk.word_tokenize(sent)) for sent in
sentences]
payload = {"instances": tokenized_sentences} response =
text_classifier.predict(payload)
predictions = json.loads(response)
print(json.dumps(predictions, indent=2))
```

```
[
    "label": [
        "_label_Artist"
    ],
    "prob": [
        0.4090951681137085
    ]
},
{
    "label": [
        "_label_EducationalInstitution"
    ],
    "prob": [
        0.49466073513031006
    ]
}
```

11. By default, the model returns one prediction with the highest probability. To retrieve the top k predictions, set k in the configuration file.

```
In [17]: payload = {"instances": tokenized_sentences, "configuration":
{"k": 2}}
response = text_classifier.predict(payload)

predictions = json.loads(response)
print(json.dumps(predictions, indent=2))
```

```
"label": [
   " label Artist",
    " label MeanOfTransportation"
 ],
  "prob": [
   0.4090951681137085,
    0.26930734515190125
 ]
},
  "label": [
    " label EducationalInstitution",
    " label Building"
 ],
  "prob": [
   0.49466073513031006,
    0.15817692875862122
}
```

12. Delete the endpoint before closing the notebook.

```
In [18]: sess.delete_endpoint(text_classifier.endpoint)
WARNING:sagemaker.deprecations:The endpoint attribute has been renamed
in s agemaker>=2.
See: https://sagemaker.readthedocs.io/en/stable/v2.html for details.
INFO:sagemaker:Deleting endpoint with name: blazingtext-2023-02-16-20-
41-33
-100
```

Next: Conclusion.

## Conclusion

Previous: Data duality for data scientists and other applications.

Based on this validation, Data scientists and engineers can access NFS data from AWS SageMaker Jupyter Notebooks via S3 buckets from NetApp Cloud Volumes ONTAP. This approach enables easy access and sharing of the same data from both NFS and S3 without the need for additional software.

#### Where to find additional information

To learn more about the information that is described in this document, review the following documents and/or websites:

· Text classification using SageMaker BlazingText

https://sagemaker-examples.readthedocs.io/en/latest/introduction\_to\_amazon\_algorithms/blazingtext\_text\_classification\_dbpedia.html

· ONTAP version support for S3 object storage

https://docs.netapp.com/us-en/ontap/s3-config/ontap-version-support-s3-concept.html

#### **Version history**

Version	Date	Document version history
Version 1.0	April 2023	Initial release

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