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D%2236%22%20y%3D%2224.1%22%20style%3D%22fill%3Aurl%28%23SVGID 00000106844
134518210696150000001682923129671618217 %29%3B%22%20width%3D%2250.5%22%20
height%3D%2251.9%22%2F%3E%0A%09%09%09%09%09%09%3C%2Fg%3E%0A%09%09%09%0
9%09%09%3C%2Fq%3E%0A%09%09%09%09%09%3C%2Fq%3E%0A%09%09%09%09%3C%2Fq%3E%0A
%09%09%09%3C%2Fq%3E%0A%09%09%3C%2Fq%3E%0A%09%3C%2Fq%3E%0A%3C%2Fq%3E%0A%3C
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.4%22%2F%3E%0A%3Ccircle%20class%3D%22st6%22%20cx%3D%2248.3%22%20cy%3D%226
5%22%20r%3D%221.6%22%2F%3E%0A%3Ccircle%20class%3D%22st6%22%20cx%3D%2264.8
%22%20cy%3D%2258.2%22%20r%3D%221.6%22%2F%3E%0A%3ClinearGradient%20id%3D%2
2SVGID 00000121260344813135191390000015347433539899256216 %22%20gradientU
nits%3D%22userSpaceOnUse%22%20x1%3D%22773.8%22%20y1%3D%22-
926%22%20x2%3D%221796%22%20y2%3D%22-
926%22%20gradientTransform%3D%22matrix%281%200%200%20-1%200%20-
876%29%22%3E%0A%09%3Cstop%20%20offset%3D%220%22%20style%3D%22stop-
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22stop-
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73.8%22%20style%3D%22opacity%3A0.2%3Bfill%3Aurl%28%23SVGID 00000121260344
813135191390000015347433539899256216 %29%3Benable-
background%3Anew%20%20%20%20%3B%22%20width%3D%221022.2%22%20height%3D%221
00%22%2F%3E%0A%3Ctext%20transform%3D%22matrix%281%200%200%201%201309.5752
%2059.46%29%22%20class%3D%22st8%20st9%20st10%22%3EIncremental%20learning%
20notebook%3C%2Ftext%3E%0A%3Ctext%20transform%3D%22matrix%281%200%200%201
%20101.02%2059.33%29%22%20class%3D%22st8%20st11%20st10%22%3EAutoAI%3C%2Ft
ext%3E%0A%3Crect%20x%3D%22231.1%22%20y%3D%2234%22%20class%3D%22st12%22%20
width%3D%221%22%20height%3D%2232%22%2F%3E%0A%3Ctext%20transform%3D%22matr
ix%281%200%200%201%20256.29%2059.66%29%22%20class%3D%22st13%20st11%20st10
$22$3EPart$20of$20IBM$20Watson$C2$AE$20Studio$3C$2Ftext$3E$0A$3C$2Fsvq$3E
)\n# Pipeline 9 Notebook for training continuation - AutoAI Notebook
```

v2.1.7\n\nConsider these tips for working with an auto-generated

```
notebook:\n- Notebook code generated using AutoAI will execute
successfully. If you modify the notebook, we cannot guarantee it will run
successfully. \n- This pipeline is optimized for the original data set.
The pipeline might fail or produce sub-optimal results if used with
different data. If you want to use a different data set, consider
retraining the AutoAI experiment to generate a new pipeline. For more
information, see <a
href=\"https://dataplatform.cloud.ibm.com/docs/content/wsj/analyze-
data/autoai-notebook.html\">Cloud Platform</a>. \n- Before modifying the
pipeline or trying to re-fit the pipeline, consider that the code
converts dataframes to numpy arrays before fitting the pipeline (a
current restriction of the preprocessor pipeline).\n"}, {"cell_type":
"markdown", "metadata": {}, "source": "<a id=\"content\"></a>\n##
Notebook content\n\nThis notebook contains code to resume and continue
training an AutoAI pipeline partially trained in an AutoAI experiment. If
there is additional training data, the notebook retrieves the data in
batches and incrementally trains the model, then tests the model.\n\nSome
familiarity with Python is helpful. This notebook uses python 3.11 and
scikit-learn 1.3."}, {"cell type": "markdown", "metadata": {"pycharm":
{"name": "#%% md\n"}}, "source": "## Notebook goals\n\nThis notebook
introduces new commands and demonstrates techniques to support
incremental learning, including: \n\n- Data reader (read data in
batches) \n- Incremental learning (`partial fit`) \n- Pipeline
evaluation\n\n## Contents\n\nThis notebook contains the following
parts:\n\n**[Setup] (#Setup) **<br>\n&nbsp; &nbsp; [Package
installation](#Package-installation) < br > \n&nbsp; &nbsp; [AutoAI experiment
metadata] (#AutoAI-experiment-metadata) <br/> \n&nbsp; &nbsp; [watsonx.ai
connection] (#watsonx.ai-connection) <br>\n**[Incremental
learning] (#Incremental-learning) ** <br>\n&nbsp; &nbsp; [Get pipeline] (#Get-
pipeline) <br/> \n&nbsp; &nbsp; [Read training data (DataLoader)] (#Data-
loader) <br/>br>\n&nbsp; &nbsp; [Incrementally train pipeline model] (#Continue-
model-training) <br/>%nbsp; &nbsp; [Test pipeline model] (#Test-pipeline-
model) <br > \n**[Store the model] (#Store-the-model) ** <br > \n**[Create online
deployment] (#Create-online-deployment) ** < br > \n&nbsp; &nbsp; [Working with
spaces](#Working-with-spaces) < br > \n** [Summary and next steps] (#Summary-
and-next-steps) ** <br/>h** [Copyrights] (#Copyrights) **"}, {"cell type":
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id=\"install\"></a>\n## Package installation\nBefore you use the sample
code in this notebook, install the following packages: \n - ibm-watsonx-
ai, \n - autoai-libs, \n - scikit-learn, \n - snapml\n"}, {"cell type":
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install autoai-libs~=2.0 | tail -n 1\n!pip install scikit-learn==1.3.* |
tail -n 1\n!pip install -U lale~=0.8.3 | tail -n 1\n!pip install
snapml==1.14.* | tail -n 1"}, {"cell type": "markdown", "metadata": {},
"source": "<a id=\"variables definition\"></a>\n## AutoAI experiment
metadata\nThe following cell contains the training data connection
details. \n**Note**: The connection might contain authorization
credentials, so be careful when sharing the notebook."}, {"cell type":
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```

```
12T14:00:57.183132Z"}, "pycharm": {"is executing": true}}, "outputs": [],
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ibm watsonx ai.helpers import
ContainerLocation\n\ntraining data references = [\n
                                                        DataConnection(\n
data asset id='a2ee2f7a-afa6-\overline{4306-8ed6-8b741b2f3842'}n
), \n]\ntraining result reference = DataConnection(\n
location=ContainerLocation(\n
                                     path='auto ml/023b00e1-0a46-4441-
89bb-28fa4dd88a65/wml data/94f47c52-82df-4cec-b71e-
3e13c17bddf2/data/automl',\n
                                    model location='auto ml/023b00e1-
0a46-4441-89bb-28fa4dd88a65/wml data/94f47c52-82df-4cec-b71e-
3e13c17bddf2/data/automl/model.zip',\n
training status='auto ml/023b00e1-0a46-4441-89bb-
28fa4dd88a65/wml data/94f47c52-82df-4cec-b71e-3e13c17bddf2/training-
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the AutoAI experiment in Watson Studio."}, {"cell type": "code",
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prediction type='multiclass',\n prediction column='Fault Type',\n
holdout size=0.1,\n
                      scoring='accuracy',\n csv separator=',',\n
                    max_number_of_estimators=2,\n
random state=33,\n
training data references=training data references, \n
training_result_reference=training_result_reference, \n
deployment url='https://au-syd.ml.cloud.ibm.com',\n
project id='091af446-afc8-42a1-a29d-8a4f8e20845b',\n
drop duplicates=True, \n
include batched ensemble estimators=['BatchedTreeEnsembleClassifier(Extra
TreesClassifier)', 'BatchedTreeEnsembleClassifier(LGBMClassifier)',
'BatchedTreeEnsembleClassifier(RandomForestClassifier)',
'BatchedTreeEnsembleClassifier(SnapBoostingMachineClassifier)',
'BatchedTreeEnsembleClassifier(SnapRandomForestClassifier)',
'BatchedTreeEnsembleClassifier(XGBClassifier)'], \n
Breakage', 'Overheating', 'Transformer Failure'], \n
feature selector mode='auto'\n)"}, {"cell type": "markdown", "metadata":
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'RUNTIME HARDWARE SPEC' in os.environ:\n
                                            CPU NUMBER =
int(ast.literal eval(os.environ['RUNTIME HARDWARE SPEC'])['num cpu'])"},
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id=\"connection\"></a>\n\#\# watsonx.ai connection\n\nThis cell defines the
credentials required to work with the watsonx.ai Runtime.\n\n**Action**:
Provide the IBM Cloud apikey, For details, see
[documentation] (https://cloud.ibm.com/docs/account?topic=account-
userapikey)."}, {"cell_type": "code", "execution_count": null,
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url=experiment metadata['deployment url']\n)"}, {"cell type": "code",
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ibm watsonx ai import APIClient\n\nclient = APIClient(credentials)\n\nif
'space id' in experiment metadata:\n
```

```
client.set.default space(experiment metadata['space id']) \nelse:\n
client.set.default_project(experiment_metadata['project_id']) \n\ntraining
_data_references[0].set_client(client)"}, {"cell_type": "markdown",
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Incremental learning"}, {"cell type": "markdown", "metadata": {},
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pipeline\n\nDownload and save a pipeline model object from the AutoAI
training job (`lale` pipeline type is used for inspection and
`partial fit` capabilities)."}, {"cell type": "code", "execution count":
null, "metadata": {}, "outputs": [], "source": "from
ibm watsonx ai.experiment import AutoAI\n\npipeline optimizer =
AutoAI (credentials,
project id=experiment metadata['project id']).runs.get optimizer(metadata
=experiment metadata) \npipeline model =
pipeline optimizer.get pipeline(pipeline name='Pipeline 9',
astype='lale')"}, {"cell type": "markdown", "metadata": {}, "source": "<a</pre>
id=\''data\ loader\''></a> \ n## Data\ loader\'n\'nCreate\ DataLoader\ iterator\ to
retrieve training dataset in batches. DataLoader is `Torch` compatible
(`torch.utils.data`), returning Pandas DataFrames.\n\n**Note**: If
reading data results in an error, provide data as iterable reader (e.g.
`read csv()` method from Pandas with chunks). It may be necessary to use
methods for initial data pre-processing like: e.g. `DataFrame.dropna()`,
`DataFrame.drop duplicates()`,
`DataFrame.sample()`.\n\n```\nreader full data = pd.read csv(DATA PATH,
chunksize=CHUNK_SIZE)\n```"}, {"cell_type": "markdown", "metadata": {},
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null, "metadata": {}, "outputs": [], "source": "from
ibm watsonx ai.data loaders import experiment as data loaders\nfrom
ibm watsonx ai.data loaders.datasets import experiment as
datasets\n\ndataset = datasets.ExperimentIterableDataset(\n
connection=training data references[0], \n
                                               enable sampling=False, \n
experiment metadata=experiment_metadata,\n
number_of_batch_rows=number_of_batch_rows\n )\n \ndata_loader =
data_loaders.ExperimentDataLoader(dataset=dataset)"}, {"cell_type":
"markdown", "metadata": {}, "source": "<a id=\"train\"></a>\n## Continue
model training\n\nIn this cell, the pipeline is incrementally fitted
using data batches (via `partial fit` calls).\n\n**Note**: If you need,
you can evaluate the pipeline using custom holdout data. Provide the
`X test`, `y test` and call `scorer` on them.\n"}, {"cell type":
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optimization metric\nThis cell constructs the cell scorer based on the
experiment metadata."}, {"cell_type": "code", "execution_count": null,
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get scorer\n\nscorer = get scorer(experiment metadata['scoring'])"},
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incremental learner\n\nFor the best training performance set:\n\n-
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plot"}, {"cell type": "code", "execution count": null, "metadata": {},
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ibm watsonx ai.utils.autoai.incremental import
plot_learning_curve\nimport time\n\npartial_fit_scores = []\nfit_times =
[]"}, {"cell type": "markdown", "metadata": {}, "source": "<a
```

```
id=\"test model\"></a>\n### Fit pipeline model in batches\n\n**Tip**: If
the data passed to `partial_fit` is highly imbalanced (>1:10), please
consider applying the `sample_weight` parameter: \n\
sklearn.utils.class weight import
compute sample weight\n\npipeline model.partial fit(X train, y train,
freeze trained prefix=True,\n
sample weight=compute sample weight('balanced', y train))\n```
"}, {"cell type": "markdown", "metadata": {}, "source": "**Note**: If you
have a holdout/test set please provide it for better pipeline evaluation
and replace X_{\text{test}} and y_{\text{test}} in the following cell.\n^{\ }\ \
import read_csv\ntest_df = read_csv('DATA_PATH')\n\nX_test =
test df.drop([experiment metadata['prediction column']],
axis=1).values\ny test =
test df[experiment metadata['prediction column']].values\n```\n\nIf
holdout set was not provided, 30% of first training batch would be used
as holdout.\n\n"}, {"cell type": "markdown", "metadata": {}, "source":
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sklearn.model selection import train test split\n\nfig, axes =
plt.subplots(1, 3, figsize=(18, 4))\ninfor i, batch df in
enumerate(data loader):\n
batch df.dropna(subset=experiment metadata[\"prediction column\"],
inplace=True) \n
                    X train =
batch df.drop([experiment metadata['prediction column']],
                    y_{train} =
axis=1).values\n
batch df[experiment metadata['prediction column']].values\n
X train, X test, y train, y test = train test split(X train, y train,
pipeline model.partial fit(X train, y train,
freeze_trained_prefix=True) \n fit times.append(time.time() -
start time) \n
                 partial fit scores.append(scorer(pipeline model, X test,
y test))\n plot learning curve(fig=fig, axes=axes,
scores=partial_fit_scores, fit_times=fit_times)"}, {"cell_type":
"markdown", "metadata": {}, "source": "<a id=\"test_model\"></a>\n## Test
pipeline model"}, {"cell type": "markdown", "metadata": {"pycharm":
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(`predict`)."}, {"cell_type": "code", "execution count": null,
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"pipeline_model.predict(X_test[:10])"}, {"cell_type": "markdown",
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model\n\nIn this section you will learn how to store the incrementally
trained model."}, {"cell_type": "code", "execution_count": null,
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client.repository.ModelMetaNames.NAME: 'P9 - Pretrained AutoAI
pipeline'\n}\n\nstored model details =
client.repository.store model(model=pipeline model,
meta_props=model_metadata, experiment_metadata=experiment_metadata)"},
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id=\"deployment\"></a>\n## Create online deployment"}, {"cell type":
"markdown", "metadata": {}, "source": "You can use the commands below to
promote the model to space and create online deployment (web
service).\n\n<a id=\"working spaces\"></a>\n### Working with spaces\n\nIn
this section you will specify a deployment space for organizing the
```

```
assets for deploying and scoring the model. If you do not have an
existing space, you can use <a href=\"https://au-
syd.dai.cloud.ibm.com/ml-runtime/dashboard?context=wx\">Deployment Spaces
Dashboard</a> to create a new space, following these steps:\n\n- Click
**New Deployment Space**.\n- Create an empty space.\n- Select Cloud
Object Storage.\n- Select watsonx.ai Runtime and press **Create**.\n-
Copy `space id` and paste it below.\n\n**Tip**: You can also use the API
to prepare the space for your work. Learn more
[here] (https://github.com/IBM/watson-machine-learning-
samples/blob/master/cloud/notebooks/python sdk/instance-
\label{lem:management.pynb).} $$\operatorname{management.ipynb}.\n\n^*: Below cells are `raw` $$
type - in order to run them, change their type to `code` and run them (no
need to restart the notebook). You may need to add some additional info
(see the **action** below).\n\n**Action**: Assign or update space ID
below.\n"}, {"cell type": "raw", "metadata": {}, "source": "space id =
\"PUT YOUR SPACE ID HERE\"\n\nmodel id =
client.spaces.promote(asset id=stored model details[\"metadata\"][\"id\"]
, source project id=experiment metadata[\"project_id\"],
target_space_id=space_id)"}, {"cell_type": "markdown", "metadata": {},
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"client.set.default space(space id) \n\ndeploy meta = {\n
client.deployments.ConfigurationMetaNames.NAME: \"Incrementally trained
AutoAI pipeline\",\n
client.deployments.ConfigurationMetaNames.ONLINE: {},\n
}\n\ndeployment details =
client.deployments.create(artifact uid=model id,
meta props=deploy meta) \ndeployment id =
client.deployments.get id(deployment details)"}, {"cell type":
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\"input data\": [{\n
                            'values': X test[:5]\n
}]\n}\n\nclient.deployments.score(deployment id, scoring payload)"},
{"cell type": "markdown", "metadata": {}, "source": "<a
id=\"cleanup\"></a>\n### Deleting deployment\nYou can delete the existing
deployment by calling the `client.deployments.delete(deployment_id)`
command.\nTo list the existing web services, use
`client.deployments.list()`."}, {"cell type": "markdown", "metadata": {},
"source": "<a id=\"summary and next steps\"></a>\n# Summary and next
steps\nYou've successfully completed this notebook!\nYou've learned how
to use AutoAI pipeline definition to train the model.\nCheck out the
official [AutoAI site] (https://www.ibm.com/cloud/watson-studio/autoai)
for more samples, tutorials, documentation, how-tos, and blog posts."},
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id=\"copyrights\"></a>\n### Copyrights\n\nLicensed Materials - Copyright
\u00a9 2025 IBM. This notebook and its source code are released under the
terms of the ILAN License. Use, duplication disclosure restricted by GSA
ADP Schedule Contract with IBM Corp.\n\n**Note:** The auto-generated
notebooks are subject to the International License Agreement for Non-
Warranted Programs (or equivalent) and License Information document for
Watson Studio Auto-generated Notebook (License Terms), such agreements
located in the link below. Specifically, the Source Components and Sample
Materials clause included in the License Information document for Watson
Studio Auto-generated Notebook applies to the auto-generated notebooks.
\n\nBy downloading, copying, accessing, or otherwise using the materials,
you agree to the <a href=\"https://www14.software.ibm.com/cgi-
bin/weblap/lap.pl?li formnum=L-AMCU-BYC7LF\">License Terms</a>\n\n "}],
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