# Machine Learning Model Specification for Cataloging Spatio-Temporal Models

### Making Geospatial Models Searchable and Usable

Francis Charette-Migneault, Ryan Avery, Brian Pondi, Joses Omojola, Simone Vaccari, Parham Membari, Devis Peressutti, Jia Yu, and Jed Sundwall. 2024. "Machine Learning Model Specification for Cataloging" Spatio-Temporal Models (Demo Paper)". In 3rd ACM SIGSPATIAL International Workshop on Searching and Mining Large Collections of Geospatial Data (GeoSearch'24), October 29-November 1 2024, Atlanta, GA, USA. ACM, New York, NY, USA, 4 pages. https://doi.org/10.1145/3681769.3698586











### Challenge of Cataloging Spatio-Temporal Models

### **Discovering Assets**

SpatioTemporal Asset
Catalog (STAC) has a
robust ecosystem for
describing, querying, and
fetching geospatial data.

### More and More Models

Lots of models are being trained on different kinds of geospatial data to solve various problems.

### **Problem Statement**

These models are being published without the necessary information to find and run them on the right data.

### Challenges deep-dive

# Data dependencies and constraints

Models trained on geospatial data are typically constrained to a sensor domain, a geographic domain and temporal domain.

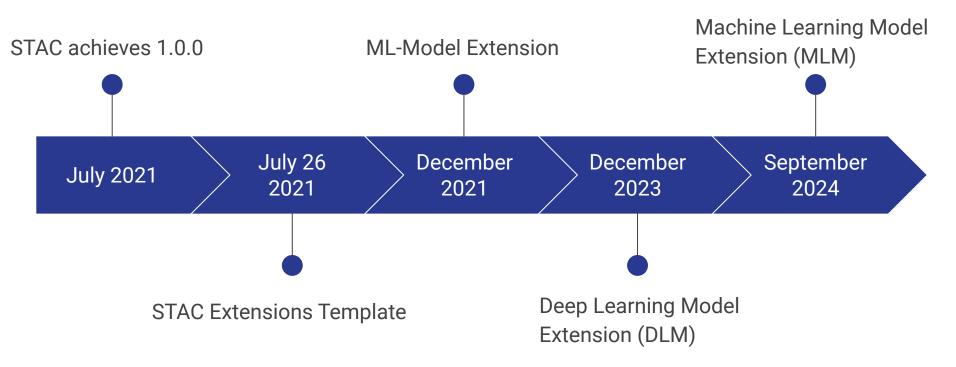
# Complex input requirements

Models have different requirements for the input shape, data type, and data preparation steps to reproduce model inference.

# Complex runtime requirements

Models have diverse dependencies on higher level language libraries, lower level libraries, and hardware-specific tooling.

### Timeline of ML standards with STAC



From 2021 - 2024, 76 STAC Extensions are indexed at <a href="https://stac-extensions.github.io/">https://stac-extensions.github.io/</a>

## Solution

A STAC Extension for Machine Learning Models

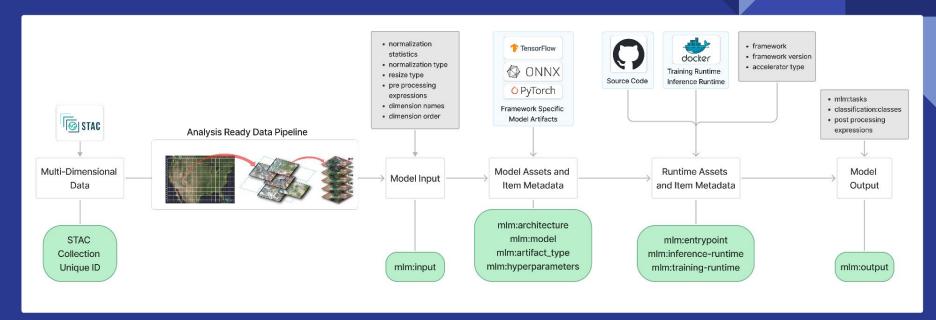
stac-extensions/mlm



### A Model Card for Geospatial

- Interoperable with STAC
   Core and Extension schemas
- ML Framework agnostic
- Multiple implementation examples from different organizations

## MLM Features



Sets norms for geospatial and temporal ML model publishing

Relates models back to geospatial collections (training/validation/test datasets)

Objective: make searching, finding and reusing geospatial models near-zero effort

### MLM Specification Structure







#### stac-model:

python library to create and validate MI M metadata

#### **Machine Learning Model Metadata Form:**

no-code form to fill and generate the STAC definition

#### MI M Item Describes metadata for a single MI M Assets model that may have one or more Metadata for source code, model associated assets, Includes STAC files, containers, and pipelines Core fields for spatial footprint and temporal range Model Asset Source Code Asset Container Asset Model file, artifact type, URI to source code. URI to published and STAC File entrypoint for inference container including the Extension metadata registry, image name, and tag Runtime fields mlm:tasks Hyperparameters Object scene-classification. mlm:framework. mlm:accelerator. object-detection. A flexible key:value object for mlm:accelerator count semantic-segmentation. parameters used in model mlm:hardware summary, etc. regression, etc. training. Model Output Contains shape, dtype, distribution information of the

#### Seach and Discovery Fields

Top level metadata for search and discovery, compatible with STAC APIS

"derived from" links pointing to reference datasets combined with ml-aoi:splits field to describe the dataset role

Data Provenance

#### Model Input

Contains shape, dtype, distribution information of the input to the machine learning model's predict function

output from the machine learning model's predict function

# STAC Extensions + MLM

Related STAC extensions	
Label Extension	Classes (name/index only), Annotations, ML Tasks, Methods
ML-AOI Extension	Area of Interest, Splits (Train, Validate, Test)
Classification	Classes (detailed) for supervised ML models
Processing	Level of processing (L4 = Model Output), Data Lineage, Expression/Software
Raster / EO / STAC 1.1 Bands	Relevant for which Bands to use as Model input
Stats	Statistics about Item counts, relevant for reference datasets and collections
Scientific	Reference scientific work, Publication Paper
File & Storage	Checksum and location of Model weights / checkpoints
Version	Model revisions (e.g.: experiment runs)

## STAC MLM Example

```
STAC and Extension Metadata
"stac extensions":
 "https://stac-extensions.github.io/mlm/v1.3.0/schema.json"
 "https://stac-extensions.github.io/raster/v1.1.0/schema.json"
 "https://stac-extensions.github.io/file/v1.0.0/schema.json"
 "https://stac-extensions.github.io/ml-aoi/v0.2.0/schema.json"
"type": "Feature", "id": "model-multi-input", "collection": <u>"ml-model-examples</u>",
                                                                                               Spatio-Temporal Metadata
 "coordinates":
"bbox": [ -7.882190080512502, 37.13739173208318, 27.911651652899925, 58.21798141355221 ],
 "description": "Example model that employs multiple input sources with different combination of bands and some inputs without any band at all.",
 "datetime": null. "start datetime": "1900-01-01T00:00:00Z". "end datetime": "9999-12-31T23:59:59Z"
 "mlm:name": "Resnet-18 Sentinel-2 ALL MOCO"
                                                                                                                       Model Metadata
 "mlm:tasks": [ "classification", "semantic-segmentation" ]
 "mlm:architecture": "ResNet"
 "mlm:framework": "pytorch"
 "mlm:framework version": "2.1.2+cu121".
 "file:size": 43000000
 "mlm:memory_size": 1
 "mlm:total_parameters": 11700000,
 "mlm:pretrained_source": "EuroSat Sentinel-2"
 "mlm:accelerator": "cuda"
 "mlm:accelerator_constrained": false,
 "mlm:accelerator_summary": "Unknown"
 "mlm:batch_size_suggestion": 256
     "norm by channel": false, "norm type": null, "resize type": null
    "name": "NDVI", "bands": [ "B04", "B08" ], "input": { "shape": [ -1, 1, 64, 64 ], "dim order": [ "batch", "ndvi", "height", "width" ], "data type": "uint16" },
    "name": "DEM", "description": "Digital elevation model. Comes from another source than the Sentinel bands. Therefore, no 'bands' associated to it.",
                                                                                                                                                  Inputs
       { "value": 0, "name": "NON VEGETATION", "description": "background pixels", "color hint": null }
        "value": 0, "name": "NON VEGETATION", "description": "background pixels", "color hint": [ 255, 255, 255 ] },
                                                                                                                                             Outputs
```

https://github.com/ stac-extensions/mlm/tree/main/examples

```
"weights"
 "href": "https://huggingface.co/torchgeo/resnet50 sentinel2 rgb moco/blob/main/resnet50 sentinel2 rgb moco.pth"
 "title": "Pytorch weights checkpoint"
 description": "A Resnet-50 classification model trained on Sentinel-2 RGB imagery with torchgeo.",
 "type": "application/octet-stream; application=pytorch",
   "mlm:weights"
 "raster:bands":
     "name": "B02 - blue", "nodata": 0, "data_type": "uint16", "bits_per_sample": 15,
     "spatial resolution": 10. "scale": 0.0001. "offset": 0. "unit": "m"
                          Model Weights and Input Bands
 "href": "https://earth-search.aws.element84.com/v1/collections/sentinel-2-12a",
 "type": "application/json"
 "ml-aoi:split": "train"
                       Reference Dataset (training data)
```

### Discussion

- Initial focus of MLM is to support FAIR principles
  - Findable
  - Accessible
  - o <u>I</u>nteroperable
  - Reusable
- Adaptable to multiple applications, data formats, variable input/output structures
- Leverages existing STAC extensions to maximize reusability and cross-compatibility with existing datasets, source data retrieval, citing related work references, etc.
- Improves integrity, provenance tracking and trust involving ML processing pipelines and spatio-temporal data for decision making

### Contact

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## Annex: STAC MLM Example

```
"stac version": "1.0.0",
                                                                                                                        STAC and Extension Metadata
"stac extensions": [
 "https://stac-extensions.github.io/mlm/v1.3.0/schema.json",
 "https://stac-extensions.github.io/raster/v1.1.0/schema.json",
  "https://stac-extensions.github.io/file/v1.0.0/schema.json".
  "https://stac-extensions.github.io/ml-aoi/v0.2.0/schema.json"
"type": "Feature",    "id": "model-multi-input",    "collection": "ml-model-examples",
                                                                                                                              Spatio-Temporal Metadata
"geometry": {
  "type": "Polygon",
  "coordinates": [ [
    [-7.882190080512502, 37.13739173208318], [-7.882190080512502, 58.21798141355221],
    [ 27.911651652899925, 58.21798141355221 ], [ 27.911651652899925, 37.13739173208318 ], [ -7.882190080512502, 37.13739173208318 ]
"bbox": [ -7.882190080512502, 37.13739173208318, 27.911651652899925, 58.21798141355221 ],
"properties": {
  "description": "Example model that employs multiple input sources with different combination of bands and some inputs without any band at all.",
  "datetime": null, "start_datetime": "1900-01-01T00:00:00Z", "end_datetime": "9999-12-31T23:59:59Z",
  "mlm:name": "Resnet-18 Sentinel-2 ALL MOCO",
                                                                                                                                              Model Metadata
  "mlm:tasks": [ "classification", "semantic-segmentation" ],
  "mlm:architecture": "ResNet",
  "mlm:framework": "pytorch",
  "mlm:framework_version": "2.1.2+cu121",
  "file:size": 43000000.
  "mlm:memory size": 1,
  "mlm:total_parameters": 11700000,
  "mlm:pretrained_source": "EuroSat Sentinel-2",
  "mlm:accelerator": "cuda",
  "mlm:accelerator constrained": false,
  "mlm:accelerator summary": "Unknown",
  "mlm:batch_size_suggestion": 256,
```

## Annex: STAC MLM Example

```
"mlm:input":
  { "name": "RGB", "bands": [ "B04", "B03", "B02" ], "input": { "shape": [ -1, 3, 64, 64 ], "dim_order": [ "batch", "channel", "height", "width" ], "data type": "uint16"
    "norm by channel": false, "norm type": null, "resize type": null },
  { "name": "NDVI", "bands": [ "B04", "B08" ],                                 "input": { "shape": [ -1, 1, 64, 64 ], "dim order": [ "batch", "ndvi", "height", "width" ], "data type": "uint16" },
    "pre processing function": { "format": "gdal-calc", "expression": "(A - B) / (A + B)" }
  { "name": "DEM", "description": "Digital elevation model. Comes from another source than the Sentinel bands. Therefore, no 'bands' associated to it.",
    Inputs
"mlm:output":
 { "name": "vegetation-segmentation", "tasks": [ "semantic-segmentation" ], "result": { "shape": [ -1, 1 ], "dim order": [ "batch", "class" ], "data type": "uint8" },
   "classification classes": [
     { "value": 0, "name": "NON_VEGETATION", "description": "background pixels", "color_hint": null },
      "value": 1, "name": "VEGETATION", "description": "pixels where vegetation was detected", "color hint": [ 0, 255, 0 ] }
   "post processing function": null
   "name": "inverse-mask", "tasks": [ "semantic-segmentation" ], "result": { "shape": [ -1, 1 ], "dim order": [ "batch", "class" ], "data type": "uint8" },
   "classification classes": [
     { "value": 0, "name": "NON VEGETATION", "description": "background pixels", "color hint": [ 255, 255, 255 ] },
     { "value": 1, "name": "VEGETATION", "description": "pixels where vegetation was detected", "color hint": [ 0, 0, 0 ] }
   "post processing function": { "format": "gdal-calc", "expression": "logical not(A)" }
                                                                                                                                                     Outputs
```

## Annex: STAC MLM Example

```
"assets":
 "weights":
   "href": "https://huggingface.co/torchgeo/resnet50 sentinel2 rgb moco/blob/main/resnet50 sentinel2 rgb moco.pth"
   "title": "Pytorch weights checkpoint".
   "description": "A Resnet-50 classification model trained on Sentinel-2 RGB imagery with torchgeo.",
   "type": "application/octet-stream; application=pytorch",
   "roles":
     "mlm:model".
     "mlm:weights"
   "raster:bands":
       "name": "B02 - blue", "nodata": 0, "data_type": "uint16", "bits_per_sample": 15,
       "spatial_resolution": 10, "scale": 0.0001, "offset": 0, "unit": "m"
       "name": "B03 - green", "nodata": 0, "data type": "uint16", "bits per sample": 15,
       "spatial resolution": 10, "scale": 0.0001, "offset": 0, "unit": "m"
       "name": "B04 - red", "nodata": 0, "data type": "uint16", "bits per sample": 15,
       "spatial resolution": 10, "scale": 0.0001, "offset": 0, "unit": "m"
       "name": "B08 - nir", "nodata": 0, "data_type": "uint16", "bits_per_sample": 15,
       "spatial resolution": 10, "scale": 0.0001, "offset": 0, "unit": "m"
                                                             Model Weights and Input Bands
   "rel": "derived from",
   "href": "https://earth-search.aws.element84.com/v1/collections/sentinel-2-12a",
   "type": "application/json",
   "ml-aoi:split": "train"
                                                           Reference Dataset (training data)
```