



# FORMAT OCEAN TRACKING DATA AND REPORT

Problem#2

Srikrishnan Sengottai Kasi  
sr284605@dal.ca

## Problem 2:

### Datasets and Attributes:

Inferred from the website:

[Acoustic Receivers]<sup>1</sup> OTN Receivers is considered a Data set with a unique id, location and information as its attributes.

Marine Scientists is a dataset with attributes like professional qualification, start Date, end date, designation id, address, email address and mobile.

Funding is a dataset to keep track of all the funds given for the growth of the project. Attributes will be like fund amount, fund provider, purpose of the fund, is fund is a loan.

Recorded data is a generic data set. This is classified based on the species. Attributes like species name, age, gender, origin, current location are tracked to serve the purpose of OTN.

OTN<sup>1</sup> council is a dataset. This gives importance to the historical values. The council has name, council id, member count. Alternatively, the scientist dataset can be generalized to employees and we can tag the council id if exists in that dataset as an attribute.

ISAC<sup>1</sup> members is also a dataset to store the various designations. It should store the designation id, designation name, voting status. This id can be reference in the employee's table.

Similar goes for SAC<sup>1</sup> and IDMC<sup>1</sup>.

[Acoustic Tags]<sup>1</sup> Taggers is a dataset to keep track of their current whereabouts. A unique id mapped to the taggers can be fixed so that when we track other species movements because of this tagger, then we can have an attribute like source received from tagger id.

Listener line is another dataset. It also carries the location, latitude, longitude and the server address to which the data has to be uploaded when an aquatic animal crosses the line.

Vemco Mobile Transceivers [VMT]<sup>1</sup> is another dataset. The thing about that is it act as both receiver and sender. This is another key part in the Ocean Tracking Network. This consists of the location details, unique id, aquatic species id to which it is associated as an attribute.

Wave Gliders is another key entity in the OTN. This stores the weather data and oceanographic data as attributes. Its location is persisted as an attribute.

### Observations from dataset:

The dataset is classified to three main segments, transmitter, receiver and deployer. Other than that, a separate dataset to the aquatic species is kept and the attributes associated with those objects are scientificname, vernacularname, origin, length, weight, gender and age.

In order to detect the aquatic species detectors are used. Each detection is recorded, and each detection are corresponded to unique datacentre. There are 4 datacentres across all the datasets and datacentre reference is almost used in all the datasets provided. The transmitter has its own properties like the id of the equipment and the type of encoding associated with it. The taggers also constitute of same attributes. It has the manufacture and the type of transmitter associated with it. Also, retrieval and recovery are another keen aspect in this process. Those are also considered as separate dataset and the time at which the recovery happened is logged. Along with that location data is stored for all the datasets.

Data cleaning:

otnunit\_aat\_animals\_8dc3\_4d15\_c278.csv:

- 1) Taxon rank attribute is not recorded for any of the entries in the dataset and it doesn't derive any value in the dataset. So, it is removed completely.
- 2) Animal\_guid seems to be a composite attribute. It can be subdivided to datacenter\_reference + animal\_project\_reference + animal\_reference\_id.
- 3) animal\_reference\_id and animal\_guid are the unique attribute across the 1309 records.
- 4) Since the guid can be constructed from the dataset we can remove that from the dataset to save some space. It will be easy to query the data with simple attributes than composite attribute. So, we can remove the animal\_guid.
- 5) Even though it looks feasible to break the table based on the scientific name, the aphaid and animal\_project\_reference is differing for one scientific name(CCSALMON and CCS) and the variation is based on the animal\_origin(H and UNKNOWN) place. So, we cannot break the dataset. Animal\_origin is a critical data from the receivers which is the main focus of the research study. So, it will be bad to assume it.
- 6) Based on the inference, when animal\_project\_reference, tsn, aphaid, datacenter\_reference and scientificname are same the origin remains the same. So, we can fill the missing origin based on that in the dataset. Even though there are some outliers like CCSALMON, we can take mean in the empty columns in few continuous records and it will not reduce the effectiveness.
- 7) Since most of the entries associated with the blue shark remains the same, we can assume that the stock is reported from the same region NWAtlantic for those records. The length\_type and life\_stage of most of the records is "FORK" and "JUVENILE" respectively. This can give us an idea that those species are moving in cluster across same listening lines and reported from the same region.
- 8) There are lot of unknown factors associated with GEFT animal project. So, the data will be corrupted if we make mean or remove the null values. We can however mark them as unknown on safer side to prevent false analysis in the future.
- 9) There are two values to denote unknown in the stock attribute. We will keep UNK to mark that. So, it is replaced accordingly in the dataset.
- 10) We cannot have NAN as a length value. This value is interpreted because of improper data type sent from the acoustic receivers and taggers. We can keep length as empty fields rather than NAN to avoid confusion while entering data to a relational table. It should be a number, but it will be absurd to take length as 0. Because no device can spot a species of length 0. Length Type is filled based on the length of the species. So, we cannot determine the length\_type of those entries too. So, we can leave those blank.
- 11) The same logic goes for the weight and age attribute too. These attributes are critical to support the vision of the project. We will keep the values empty and will not add NOT\_NULL constraint in the relational tables.
- 12) Unfortunately, age of almost 3658 records out of 3808 is unknown. So, it is better to remove that attribute as it will not support strongly in the analysis.
- 13) Out of the 3808 records, the value of sex is present for only 486 records. Even in that only 145 records have their sex determined. It is easy to mark all the blank one as unknown. But without determining sex in this dataset their population expansion and survival metrics become clueless. This will not contribute to their survival metrics in the changing climate condition in the waters. Due to very less information that can be derived from this attribute, removing this column from the dataset will not affect our analysis in the future.

- 14) Without proper information in the age, it will be difficult to point out its life stage. But life stage is present for considerably a greater number of species so we can fill empty columns with UNK as a value.

otnunit\_aat\_datacenter\_attributes\_8a94\_cefd\_f8a3.csv:

- 1) In this a lot of data are repeated. Vast information like abstract and license are repeated and stored in the records.
- 2) Additionally, we have empty columns which are critical but because of its empty values provides no value to the present dataset can be removed. The columns in that category are, date\_modified, datacenter\_distribution\_statement, datacenter\_date\_modified, time\_coverage\_start, time\_coverage\_end.
- 3) NAN is not applicable to fields that takes number as input. So, they are made blank and optional while creating relational database. Those fields are datacenter\_geospatial\_lat\_min, datacenter\_geospatial\_lat\_max, datacenter\_geospatial\_lon\_min. It will be bad to assume that information based on Mean or deviation because of the drastic variation in the location of the datacenter.
- 4) Fields like datacenter\_pi\_organization, datacenter\_abstract, datacenter\_pi, datacenter\_pi\_contact, datacenter\_infourl, datacenter\_keywords, datacenter\_keywords\_vocabulary, datacenter\_doi, datacenter\_license, remains the same for all four records. Those fields can be linked to the organization in a separate table. The organization can later be used to access those details.

otnunit\_aat\_detections\_9062\_5923\_1394.csv

- 1) detection\_guid is a composite attribute comprising of datacenter\_reference + detection\_id + detection\_project\_reference. So, select queries can use and operator to retrieve those records and the column can be deleted to eliminate repetition and save some space.
- 2) detection\_reference\_type stores only ANIMAL as field value. So that can be deleted from the spreadsheet and can be set as a default value while creating database. It is evident that OTN is tracking only ANIMAL in the ecosystem. So that column can be deleted as no predictions can be made from that field.
- 3) detection\_transmittername is the combination of transmitter\_codespace and transmitter\_id. So, its again storing a composite attribute in the dataset which can be removed to reduce the redundancy and increase the storage.
- 4) receiver\_log\_id is full of NULL values and doesn't add any value to the collected dataset. So, we can remove the complete column from the dataset. This is a critical attribute to trace back in case of spotting future failures. But the system failed to record it. So removing the column for now from the provided dataset.
- 5) depth is another column that is filled with NaN fields. So, for the cleaning process we can remove that column from the dataset. But that was an important attribute to find the location of the species.
- 6) position\_data\_source entity carries only one value across the dataset. The value is "Receiver Metadata". The storage is required only when there are multiple values for a given field type. But there are more than 2 lakhs records carrying the same value. This reduces the fields importance in the dataset and can be removed. These values can be kept constant in the application layer itself. May be if there are different types of values in the future, we can add that in the dataset.

- 7) `uncertainty_in_latitude`, `uncertainty_in_depth` and `uncertainty_in_longitude` are another three disappointing fields in the dataset. Even from the field name we can understand that it is present to calculate any error that may have occurred from the readings sent by transmitter it doesn't add any significant value as an overall. In fact, it dilutes the precision of the information stored in the dataset. So, we can remove these two columns from the dataset to keep it clean. The same goes for `depth_data_source`, `other_position_data`, `dataset_quality`. They don't contribute anything for decision making in future for the recorded data. So, wiping them too.
- 8) `detection_quality` is another interesting attribute showcasing only one field value. "Found Receiver". But by the name of the field, it symbolises a quantitative measure than a string value. Because of the only one value recorded by that field in the dataset it loses its significance. In real time it plays a major role in deciding the authenticity of the data. Filling the empty values as "Not Available" will be more appropriate in this case.
- 9) `sensor_data` is empty for many rows. Based on the attribute name this data is recorded by the sensors deployed in the OTN. If no data is sent from the sensor it is safe to assume that as 0.0. As most of the fields are integers in this column.
- 10) Based on the `detection_serial_number` the `latitude_degrees_north` and `longitude_degrees_east` is detected. This shows the position of the detector which detects the species and send to the receiver. The table can be divided to store a mapping of `detection_serial_number` list and the latitude and longitude values associated with it.

`otnunit_aat_manmade_platform_0735_7c9f_329c.csv`:

- 1) `platform_guid` = `datacenter_reference` + `platform_project_reference` + `platform_reference_id`. So, this is a composite attribute and can be constructed later based on that logic. Hence, removing this column from the dataset to remove repetitive data.
- 2) `platform_reference_id` and `platform_name` is exactly one and the same. So, these records are considered duplicates in the dataset and `platform_name` can be removed. To avoid this the dataset should had different naming conventions for the `platform_name`.
- 3) NaN in `platform_depth` can be replaced with empty values. As clearly the depth can't be a string. So, we can have empty values for whatever the reading there is no depth information provided. Keeping it 0 will be wrong as it conveys a position.
- 4) This is a weak dataset as it doesn't have a single unique key to identify the record. So, bringing together latitude, longitude and `platform_reference_id` makes the values in the dataset look unique. Because of the null values present in the latitude and longitude makes it difficult to create an index. So, we can keep the `platform_reference_id` as the unique id and make the entity identifiable.

`otnunit_aat_project_attributes_f29c_fb21_23a3.csv`:

- 1) `project_references`, `project_distribution_statement`, `project_date_modified` and `project_doi` are completely null for all the records. They are project attributes associated with the `project_reference`. Moreover `project_references` and `project_reference` almost looks the same. So, removing it will not degrade the data. Same goes for `geospatial_vertical_positive`, `time_coverage_start`, `time_coverage_end`, `project_linestring` are also completely null and provides no use to the dataset.

- 2) project\_infourl has few values that are improper. <NULL>, NA and some blank fields. So making it all same by filling the value as NA which represents Not Available for that field value.
- 3) project\_pi\_contact field is present for most of the columns. So, we can keep the empty values as Not Available(NA) for that column.
- 4) project\_keywords\_vocabulary, project\_license, project\_datum haven't varied based on any records. These types of static data can be associated separately in the application level. The license is pretty common across the OTN process and huge. The license column can be removed to reduce the redundancy and increase the storage.

otnunit\_aat\_receivers\_c595\_05f4\_68b2.csv

- 1) expected\_receiver\_life is a field that is populated with NaN and blank values. This is an empty data in the dataset and provides no information about the receiver's lifespan. But this is an important attribute to be kept track. So, that we can stock our receivers accordingly in the future. Removing it in the spreadsheet to reduce the data.
- 2) deployed\_by is another key attribute which is required to make a point of contact in the future. But the entire column is empty which makes the data insignificant and it can be removed from spreadsheet. But it can be used while creating a relational schema. This can be updated in the future.
- 3) deployment\_guid is a composite attribute which comprises of datacenter\_reference, deployment\_project\_reference and deployment\_id. This creates repeated data, so it can be removed.
- 4) frequencies\_monitored, receiver\_coding\_scheme are empty throughout the records and it is removed.
- 5) Receiver\_serial\_number values are tampered. It has few "?" and values like "can't read receiver number". Those values are replaced by NA for consistency.

otnunit\_aat\_recover\_offload\_details\_4b23\_f002\_f89a.csv

- 1) recovery\_guid = datacenter\_reference + recovery\_project\_reference + recovery\_id + deployment\_id. It's a composite attribute and it can be derived from the above-mentioned simple attributes whenever needed. So, removing them to save some space.
- 2) clock\_synchronized, recovered\_by fields are completely null for records and makes it clueless to arrive at any decision based on the dataset. Thus, this can be added when sufficient information is recorded for it in the future.
- 3) There is a high resemblance in the deployment\_id and recovery\_id with few outliers. Even few deployment\_id had some timestamp appended to it. Those are dirty data. From the list of deployment\_id formats that is observed they seem to be more like a combination of the receiver\_id and recovery\_project\_reference.

otnunit\_aat\_tag\_releases\_b793\_03e7\_a230.csv:

- 1) tag\_programming\_id, tag\_frequency, and transmitter\_type is completely null for all records. So, it can be removed from the dataset. In the case of taggers, they are attached with the transmitters. So, those attributes are important attributes and can be considered for the relational schema.
- 2) release\_guid = datacenter\_reference + release\_project\_reference + tag\_device\_id. Another composite attribute that can be removed for easy retrieval purposes in the future.

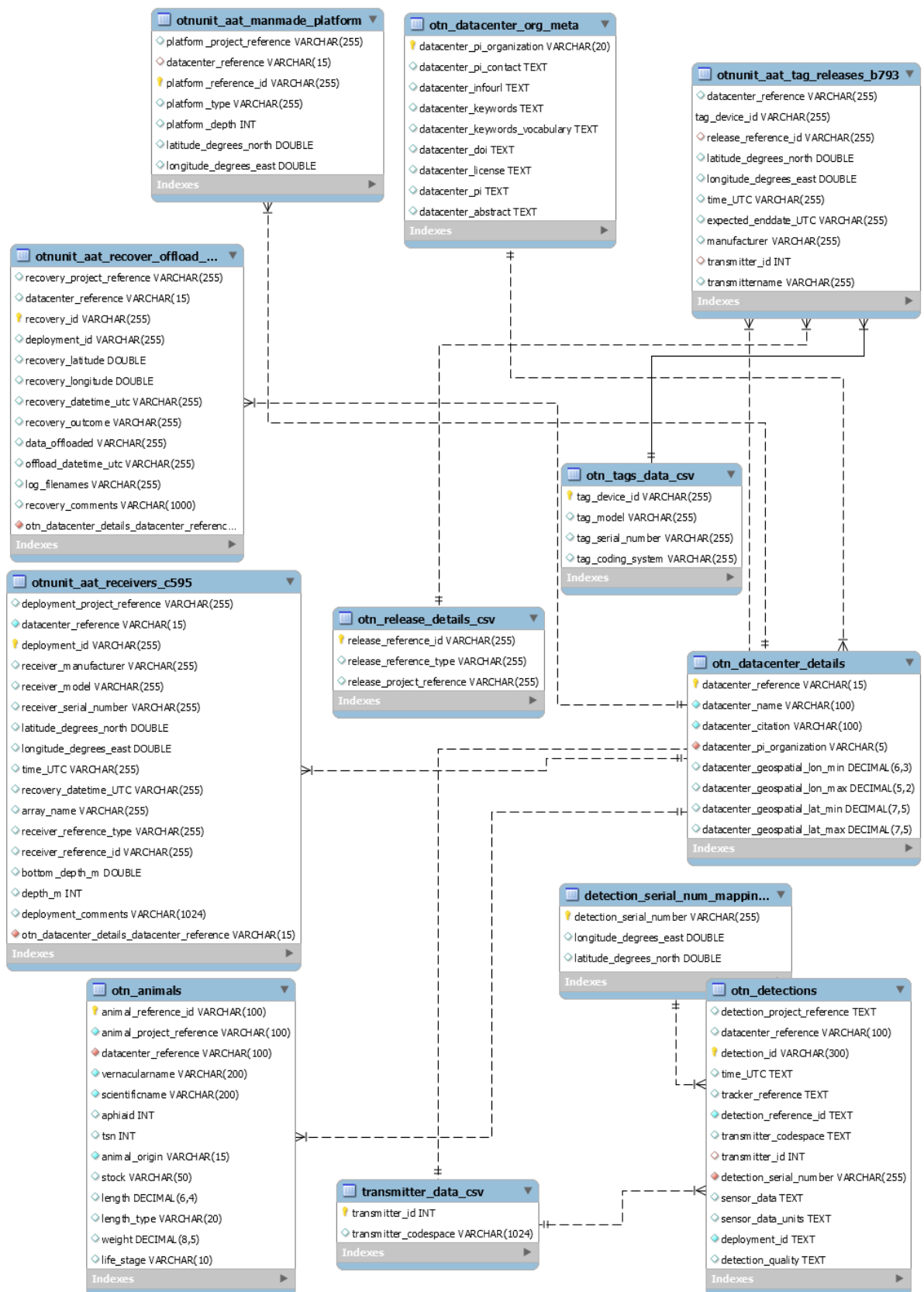
- 3) The table contains detailed information about the release and taggers. Hence, they are subdivided from the main dataset. We can keep the reference alone in the main dataset.
- 4) One table derived from it is release table. It will have release\_reference\_id as a primary id. It will carry release\_reference\_type and release\_project\_reference as other two fields. The release\_reference\_id alone can be stored in the main dataset.
- 5) Another part that can be divided is the tags. The tag\_device\_id can be the primary key of this table. It takes tag\_model, tag\_serial\_number, tag\_coding\_system along with it.

In common the transmitter\_id has been linked in two of the above datasets.

- 1) detectors
- 2) taggers

So, a separate decomposed dataset is made to map transmitter\_id and the encoding pattern associated with it.

Reverse engineered ERD:





Citations and references:

[1] <https://oceantrackingnetwork.org/about/#oceanmonitoring>