Scalar Encoder with Buckets

Sahith Kumar Singari  
[sahith.singari@stud.fra-uas.de](mailto:sahith.singari@stud.fra-uas.de)

Anil Kumar Gadiraju  
[anil.gadiraju@stud.fra-uas.de](mailto:anil.gadiraju@stud.fra-uas.d)

Vinay Kumar Bandaru

[vinay.bandaru@stud.fra-uas.de](mailto:vinay.bandaru@stud.fra-uas.de)

*Abstract*— In real-life situations, Scalar encoders with buckets are useful in a variety of real-world scenarios where continuous quantities must be represented as discrete values. This is frequently the case in machine learning and data analysis, when you may want to convert a continuous characteristic such as age, temperature, or height into a categorical feature that may be used in a model or algorithm. Sensor data analysis is one common application for scalar encoders with buckets. Encoders convert various data types into sparse distributed representations. They accept external inputs and convert them into a binary representation known by the CLA, similar to how the retina or cochlea convert external information into binary neuronal representations. This paper evaluates Investigate the implementation and performance of scalar encoding using buckets in AI and machine learning applications. We will examine the literature on best practices for determining the number and breadth of buckets, compare scalar encoding using buckets to other encoding approaches.

*Keywords- Hierarchical Temporal Memory (HTM), Sparse Distributed representations (SDRs), neocortex, numeric, array*

# **Introduction**

Using a scalar encoder with buckets, continuous data can be transformed into a set of discrete values that can then be utilized for analysis, modeling, or machine learning. With this method, the range of continuous values is divided into a number of discrete intervals, or "buckets," and each value is then assigned to the bucket that it belongs in. The encoding procedure can be accomplished in a variety of ways and customized to the unique requirements of the application, for as by employing a binary or multi-level encoding scheme. In applications including sensor data analysis, natural language processing, and picture classification, scalar encoder with buckets are frequently employed and have shown to be effective tools for transforming continuous data into discrete data.

Scientists have gained insights by working on the cortex that sequence learning has large invariant changing series of inputs. The exact neural mechanism of sequence memory is still unknown, but models that give a reading of the neurons are used to study. These models show significant capabilities to recollect and recognize the sequence of inputs using rules. These ML models do not match the real-world issues

Hierarchical Temporal Memory (HTM) is a Biomimetics model based on the principles of memory predictions developed by scientists to capture the architectural and algorithmic features of the neocortex [4] [5]. HTM has given promising results in pattern recognition, and This can learn the temporal sequences and spatial flow of sensory inputs as data.

# **LITERATURE** **SURVEY**

## **SDRs**

Sparse Distributed representations (SDRs) of input patterns are used in HTM's language. With a set amount of active bits, it produces SDRs internally. These bits have semantic value. As a result, two inputs with equivalent semantic meaning must have equal active bit representation in SDR, which plays an important role in HTM learning.

Hierarchical Temporal Memory (HTM) technique is based on the concept of SDRs, which are high-dimensional binary vectors with only a small fraction of the bits set to 1. SDRs are a natural way for the brain to represent patterns because they allow for the efficient storage and processing of large amounts of information.

SDR can be used in a wide range of applications using HTM systems. Consequently, the first step of using an HTM system is to convert a data source into an SDR using what we call an encoder. The encoder converts the native format of the data into an SDR that can be fed into an HTM system. The encoder is responsible for determining which output bits should be ones, and which should be zeros, for a given input value in such a way as to capture the important semantic characteristics of the data. Similar input values should produce highly overlapping SDRs.

Diagram

Description automatically generated

*Figure.1: Cochlear hair cells excite a group of neurons based on the frequency of the sound.*

## **Encoders**

Encoders should generate SDRs with a constant number of bits 'N' and a fixed number of active (1's) bits 'W', regardless of what they represent. What do you know about the ideal values for N and W?

To preserve the features of sparsity, we cannot be a large proportion of N. But, if W is too small, we lose the features of a distributed representation.

While encoding data, there are several unique aspects to consider:

1. Semantically related data can trigger SDRs with overlapping active bits.

2. The same input should always provide the same SDR output.

3. The result will have the same dimensionality as all of the inputs (total number of bits).

Deterministic encoders should produce the same result from the same input each time. Without this attribute, the sequence learnt in HTM will be redundant because there is a shift in values with encoded representations. Put an end to creating adaptive or random element encoders.

The output of an encoder must produce the exact same number of bits for each of its inputs. SDRs are compared and handled so that a bit with a specific "value" is always at the same location using a bit-by-bit assumption. If the encoders offered various SDR bit lengths, comparisons and other operations would not be possible.

## **Hierarchical Temporal Memory (HTM)**

The HTM model learns the procedure that occurs in one layer of the brain. HTM works on continuous streams of input patterns, attempting to construct rare and constant representations of input sequences based on the input stream's recurrent pattern.

HTM's capacity to forecast future patterns based on previously trained data patterns. After a few cycles, HTM receives a unique pattern that compares the prior patterns to the current pattern. Input patterns should not repeat, and the uniqueness should be maintained.

# **Methodology**

The project Scalar Encoder with Buckets developed using C# .Net Core in Microsoft Visual Studio 2022 IDE (Integrated Development Environment) is used as a reference model to understand the functioning of Scalar Encoder which uses

## **Spatial Pooler**

The active columns' cells are mapped during the creation of SDR input by Spatial Pooler. Each column connects to the following section of input bits via a network of synapses. While many columns would have the same appearance, these columns are distinct from one another. Varying patterns result in varied levels of activation, and stronger activation limits weaker activation of the columns. Columns may cover a little portion of the space or the entire surface. Implementing the inhibitory mechanism results in a constrained representation of the input. Similar patterns result in similar activation columns. HTM learns from the input and breaks down cell connections. Learning results by updating synapse persistence. Active bits in the active columns enhance the persistence value, while the other columns make it smaller. Inactive columns do not pick up new information. To make sure that every column takes part in the training, the inactive columns are boosted. The spatial pooler implies collections or clusters of spatially related data. During the spatial pooler's learning process, every pattern that manifests at the input is compared to the database of other patterns. The other columns make it smaller. Inactive columns do not pick up new information. To make sure that every column takes part in the training, the inactive columns are boosted. The spatial pooler implies collections or clusters of spatially related data. During the spatial pooler's learning process, every pattern that manifests at the input is compared to the database of other patterns.

*Figure. 3: HTM Algorithm Flow*

## **Sparse Distributed Representation**

In the HTM, SDR is an effective information organization system. Sparse means that only a tiny percentage of the big, interconnected cells are active at any given time. "Distributed" denotes that active cell are dispersed throughout the region and will be used to depict the region's activity. Because the binary representation is more biologically reasonable and highly computationally efficient, HTM considers the binary SDR converted from a specific encoder. Even though the number of possible inputs exceeds the number of possible representations, the binary SDR does not result in a functional loss of information due to the following critical features of the SDR.

## **Bucket Concept**

Buckets, a predetermined number of uniformly sized intervals or "buckets" are used to partition the range of continuous values. The width of the interval, which is equal to the difference between the maximum and minimum values divided by the number of buckets, establishes the size of each bucket.

For instance, if we wanted to generate 10 fixed-width buckets and we had a range of continuous values from 0 to 100, we would first determine the width of each bucket:

Bucket width = (100 - 0) / 10 = 10 copy the code

This implies that each bucket will have a width of 10 units. Here is how the 10 buckets would be described [0, 10], [10, 20], [20], [30], [40], [40], [50], [60], [60], [70], [80], [80, 90], [90, 100]

The interval boundaries are inclusive at the lower end and exclusive at the top end, as you can see. A square bracket [indicates that the endpoint is included while a parenthesis] indicates that the endpoint is excluded. This is a widely used convention in interval notation.

Each continuous value is assigned to the appropriate bucket based on its value after the buckets have been defined. For instance, as it lies between 40 and 50, the bucket [40, 50] would be given a value of 42.

Fixed-width bucketing is a straightforward and easy-to-use scalar encoding technique that works well with a variety of data formats. However, given that some buckets may include data that is not evenly distributed, it might not be the best option for greater than others in terms of values. Adaptive or variable-width bucketing may be a better option in some circumstance.

# **implementation**

To build an effective encoder that can result in similarity, you must comprehend the components of your data.

In the afore mentioned hearing frequency example, the encoder was designed to have noises with a comparable pitch but did not account for how noisy the sounds were, which would call for a different strategy. The first step in designing an encoder is to choose whatever data element you want to capture.

Pitch and amplitude may be two of a sound's primary characteristics, whereas the weekend status of a date may be one. When two inputs have the same values for one or more of the selected data attributes, the encoder will produce overlapping representations of those inputs.

Moreover, there must be enough one-bits to account for subsampling and noise. Having one bit of at least 20 to 25 is a good general rule of thumb. Because of the noise and non-determinism in HTM systems, encoders with representations less than 20 one-bits will perform poorly and be more prone to errors.

We first divide the range of values into buckets when building an encoder implementation, and then we map the buckets into a group of active cells.

1) Decide on the value range (MinVal and MaxVal ).

Range = MaxVal - MinVal in 2.

3) The "width" of the output signal "W"—the amount of bits that are set to encode a single value—should be determined.

4) The output's total number of bits, "N," should be chosen as the representational bit count.

5) A periodic or non-periodic parameter should be used to calculate the resolution.

6) Starting with N unset bits and setting them one at a time results in the encoded representation.

## **Scalar Encoder Implementation**

A scalar encoder converts a numeric (floating point) value into a bit array. Except for a contiguous block of 1's, the output is all 0's. The placement of this contiguous block changes in real time as the input value changes.

Linear encoding is used. If you want a nonlinear encoding, simply alter the scalar before encoding (e.g., with a logarithm function).

Binding the data as a pre-processing step, such as "1" = $0 - $.20, "2" = $.21-$0.80, "3" = $.81-$1.20, etc., is not advised because it removes a lot of information and prevents neighboring values from overlapping in the result. Instead, employ a continuous transformation that scales the data (a piecewise transformation is fine)...otherwise, "MaxVal " is a valid upper bound.

**param w**: The number of bits assigned to encode a single value - the "width" of the output signal restriction: w must be odd to avoid centering difficulties. **MaxVal :** The maximum value of the input signal. (If "periodic Equals= True," input is strictly smaller) **periodic**: If true, the input value "wraps around" so that "MinVal" = "MaxVal ". The input for a periodic value must be strictly less than "MaxVal ," otherwise "MaxVal " is a valid upper bound.

The number of bits in the output is specified by the parameter n. "**w**" must be more than or equal to the representations of two inputs separated by more than the radius are non-overlapping. In general, two inputs separated by less than the radius will overlap in at least some of their bits. This can be thought of as the radius of the input. param resolution: Two inputs separated by larger than or equal to the resolution will always have different representations. name: an optional string that will be included in the description: If clip Input is set to true, non-periodic inputs less than MinVal or greater than MaxVal are trimmed to MinVal/MaxVal .

If true, skip several safety tests (for compatibility reasons), otherwise false. Please keep in mind that "radius" and "resolution" are specified in relation to the input, not the output. "w" is given in relation to the output.

**InitEncoder:** (Assistant function) There are three ways to think about the representation.

**GetFirstOnBit:** The bit offset of the first bit to be set in the encoder output is returned.

As the encoded output wraps around in periodic encoders, this can be a negative value.

**GenerateRangeDescription**: create a description from a range's text description.

**GetBucketIndex**: Subclasses must override this.

returns a list of things, one for every bucket that this encoder has specified.

Each item represents the value assigned to that bucket; it has the same structure as the input that would be returned by the GetBucketInfo method.

If all you need are the bucket data, this call is quicker than calling:meth:'.getBucketInfo' on each bucket separately.

return: a list of things, each item corresponding to a bucket's value.

**SetBucketValue:** Set the value of the bucket at the given index to the given value.

**GetBucketInfo:** The information about the associated bucket in the scalar encoder is included in an int array that is returned by the GetBucketInfo method, which accepts a double input value.

By determining if the input value is more than the maximum value or less than the minimum value, the input value is first clipped to the range of the scalar encoder.

After dividing the encoder's range into N buckets, the algorithm determines which bucket the input value belongs to before calculating the index of that bucket. As the bucket index is determined by dividing by integers, it will always be an integer.

The algorithm then determines the center of the bucket by multiplying the bucket's initial value by half its width. Moreover, it determines the bucket's beginning and ending values.

In the case of a periodic encoder, the approach wraps the bucket index if it exceeds N buckets. Then, accounting for the periodicity of the encoder, it determines the distances to the closest bucket edges. The center of the bucket is determined by the method using the nearest edge.

A rounded bucket center value, rounded bucket starting value, rounded bucket ending value, and an int array with the bucket index are all included in the method's final output. These values can be used to represent the input value in the scalar encoder as an array of binary digits.

## **EncoderBase:**

An encoder is a program that converts a value to a sparse distributed representation.

This is the foundation class for encoders that are OPF compatible. For use in locations like the SDR Classifier, the OPF requires that values be represented as a scalar value.

**EncodeIntoArray**: encodes input Data and writes the encoded value to the 1-D array of length .

param output: 1-D array of the same length.

getEncodedValues: gives the input back in the same format as the method ".topDownCompute". This is the same as the input data for the majority of encoder types. For instance, this corresponds to the string and numeric values from the inputs for the scalar and category types, respectively. This gives the list of scalars for each of the sub-fields for datetime encoders (timeOfDay, dayOfWeek, etc.)

return: a list of values with the same structure and arrangement as the values produced by the "topDownCompute" method.

returns an array containing the input Data’s input sub-fields' bucket indices. To acquire the related field names for each of the buckets. param input Data: The data from the source. Usually, this is a member of an object. return: array of bucket indices. encodeIntoArray"

Given that a new array is allocated with each call, this might be less efficient.

inputData: The input data that needs to be encoded.

return: an array containing the inputData's encoded form.

**ClosenessScores:** Calculate ratings of proximity between the expected and actual scalar values.

For each value in expValues (or actValues, which must be the same length), this method provides a single closeness score. The proximity score runs from 0 to 1.0, with 1.0 representing the best possible match and 0 representing the poorest one.

If this encoder is a straightforward one-field encoder, it will only accept one item in each of the "actValues" and "expValues" arrays.

One item per sub-encoder is what multi-encoders anticipate.

Each type of encoder has its own proximity metric that can be defined. A category encoder, for instance, might return either 1 or 0.

## **Testcases Methods of Scalar Encoder**

**ScalarEncoderTests:** This is a unit test for the ScalarEncoder class' Encode function. It makes clear that this is a unit test that the testing framework may run by using the [TestMethod] attribute. When performing smaller sets of tests, it can be useful to categorize the test using the [TestCategory("UnitTest")] attribute.

Test data for the test method is provided by the [DataRow] attribute. The test data in this instance consists of a double value and an array of integers that represents the predicted outcome. The width, MinVal, MaxVal , and radius parameters should be used by the ScalarEncoder class to encrypt the double value, and it should then return an array of integers that contains the resulting value. The test then compares the outcome to the anticipated outcome to see if the test has passed or failed.

**ScalarEncoderUnitTestNonPeriodic:** A scalar input value and an array of integer values (the sparsedistributed representation) are the arguments for the "ScalarEncoderUnitTestNonPeriodic" function. A new instance of the "CortexNetworkContext" class is then created, and various encoder options are then set using a dictionary. A new instance of the "ScalarEncoder" class is then created and used to encode the input value. With the "Assert.IsTrue" method, the obtained sparse distributed representation is contrasted with the anticipated outcome.

**SeasonEncoderTest:** A binary vector is created by the encoder from a scalar value that represents the day of the year. The test determines whether the encoder outputs the anticipated binary vector given a specific input value.

An expected binary vector and an input value of 1.0 (indicating January 1) are specified in the test's single data row. The length of the anticipated binary vector, which is equal to the number of bits utilized to encrypt the value, is 12. The value is on the edge of the range of values that the encoder can encode because the first and last bits are both set to 1, which indicates this. The value falls into the first group of encoded values because the second bit is also set to 1, signalling this. the rest of set to 0, indicating that the value is not in any of the other categories.

The test uses the Assert.IsTrue method to check if the expected binary vector is equal to the vector returned by the encoder for the given input value. If the vectors are not equal, the test will fail.

**TimeTickEncodingTest:** scalar encoder is used. The test is configured to encode a collection of numbers and generate a bitmap output, and it currently seems to be passing.

The code creates a scalar encoder that can encode values in the range of 0 to the number of days in the current year by using a DateTime object to obtain the current time. After encoding each value in the range using a loop, it ensures that the output has the right number of bits set to 1.

The aim of this test is not quite evident from the code, but it appears to be confirming that the scalar encoder can accurately encode a variety of time values.

**ScalarEncodingGetBucketIndexNonPeriodic:** Using a scalar encoder with non-periodic settings, a scalar value is encoded, and for each encoded scalar value, the encoder's GetBucketIndex method is used to retrieve the bucket index. The encoded output is then displayed as a 2D array bitmap, with green pixels designating active bits. The bitmap is then saved to a file with a name derived from the scalar value.

The test appears to cover proper GetBucketIndex method operation as well as visual verification of the encoded bit pattern for non-periodic scalar encoder settings.

**ScalarEncodingGetBucketIndexPeriodic:** These methods appear to be testing the ScalarEncoder class's GetBucketIndex method, which returns the bucket index for a specified scalar value. With a non-periodic encoder, the first method, ScalarEncodingGetBucketIndexNonPeriodic, generates a bitmap image for a set of scalar values and their corresponding bucket indices. For a periodic encoder, the second method, ScalarEncodingGetBucketIndexPeriodic, accomplishes the same task. These techniques appear to be helpful for observing how the bucket indices alter depending on the encoder parameters and scalar values.

**TestGenerateRangeDescription:** The ScalarEncoder class's GenerateRangeDescription method, which accepts a list of tuples representing ranges of values and produces a textual description of those ranges, is tested by the TestGenerateRangeDescription method.

First, the function constructs a ScalarEncoder class instance with a range of 0 to 100 with periodicity set to true. Asserting that the created string descriptions match the anticipated values, it then generates three separate lists of value ranges.

Two ranges are specified in the first test case: 1.0 to 3.0 and 7.0 to 10.0. The two ranges are separated by a comma and are each represented by its lower and higher boundaries, formatted to two decimal places, in the expected string description of "1.00-3.00, 7.00-10.00."

In the second test scenario, a single value is specified as a range of 2.5 to 2.5. The string description that should be used is "2.50," which merely denotes a single value with two decimal places.

Two ranges are specified in the third test case: 1.0 to 1.0 and 5.0 to 6.0. The two ranges are separated by a comma, and the intended string description is "1.00, 5.00-6.00," which represents the first range as a single value and the second range as its lower and upper bounds formatted to two decimal places.

**ClosenessScorestest:** The ScalarEncoder class ClosenessScores method is being tested in this test. The ScalarEncoder is created with a set of parameters, two arrays of expected and actual values are defined, fractional is set to true, and the expected closeness score is set to 0.99. With these parameters, it then invokes the ClosenessScores function and verifies that the outcome is within 0.01 of the predicted closeness score. This test case verifies that the ScalarEncoder is used by the ClosenessScores function to determine the closeness score between the expected and actual values.

**InitTest:** The setting-based initialization of the ScalarEncoderExperimental object.

By invoking the getDefaultSettings method, it first sets the default encoder settings. The "Name" property is then changed to "hello" and a new property called "TestProp1" with the value "hello" is added.

Then, it calls the Initialize method on a newly created ScalarEncoderExperimental object to initialize it with the adjusted encoder settings. Also, it uses the indexer property to set a new property "abc" with the value "1".

Finally, it claims that the encoder object's "TestProp1," "Name," and "abc" attributes are set to the required values.

Overall, the ScalarEncoderExperimental object can be initialized appropriately with the given values, and properties may be set and accessed correctly, according to this test method.

**InitializeAllEncodersTest:** It initializes every encoder listed in the CortexNetworkContext object and verifies that the initialized encoder's properties match the encoderSettings dictionary's default values.

The CortexNetworkContext instance is first created, and the method confirms that it has at least one encoder and is not null. It then generates a fresh instance of the ScalarEncoderExperimental class using the encoderSettings dictionary for each encoder in the context. It then repeatedly checks that each setting's value corresponds to the appropriate value of the initialized encoder by iterating over the encoderSettings dictionary.

Overall, this function verifies that all encoders are appropriately initialized with the specified settings by the CortexNetworkContext's CreateEncoder method.

static Dictionary: A dictionary containing the default settings for a scalar encoder is the result of the getDefaultSettings function. These conditions consist of:

W: The "width" of the output signal, or the number of bits that are set to represent a single value. The default value for this integer is 11. Keep in mind that W must be odd to prevent centering issues.

N: how many bits are in the output. The default value for this integer is 0. N will be determined as the smallest power of 2 bigger than or equal to W if it is not specified.

MinVal: the input signal's lowest value. The default value for this double value is 1.

MaxVal : the input signal's maximum value. This value has two parts.

MaxVal : the input signal's maximum value. The default value for this double number is 100.

Radius: Non-overlapping representations exist when two inputs are separated by greater than the radius. In general, there will be at least some bit overlap between two inputs that are closer together than the radius. The default value for this double value is 0.

Resolution: It is assured that two inputs that are separated by a distance larger than or equal to the resolution will have different representations. The default value for this double number is 0.15.

Periodic: If this condition is true, the input value "wraps around" so that MinVal = MaxVal. If the input for a periodic value is not strictly less than MaxVal , MaxVal becomes a valid upper bound. A boolean value, true is the default for this one.

Non-periodic inputs that are smaller than MinVal or greater than MaxVal will be trimmed to MinVal/MaxVal if ClipInput is true. A boolean value, true is the default for this one. Name: a string that identifies the encoder. It is "TestScalarEncoder" by default.

The boolean flag IsRealCortexModel indicates whether the encoder is a component of a real cortex model. True is the default.

**ScalarEncodingEncodeIntoArray:** This test method examines the ScalarEncoder class EncodeIntoArray function. The method requires a boolean indication indicating whether or not the encoder should learn, an output array, the length of the output array, and an integer input value. The test method calls the EncodeIntoArray method for a range of input values after creating an instance of the ScalarEncoder class with certain default settings.

The method checks each index of the array for each input value to ensure that the encoded input value has been successfully populated into the output array. Moreover, a bitmap of the encoded value is created and saved to disk. The method checks sure the EncodeIntoArray function returns zero before concluding.

This test method effectively covers the EncodeIntoArray method and validates that input values are appropriately encoded into an output array. Additionally, it confirms that the procedure produces the desired outcome. To make sure that the approach is effective for a variety of inputs, extra edge cases and input values could be tested as part of the test procedure.

**ScalarEncodingDecode:** This unit test is for the ScalarEncoder class decode function. The decode method extracts the original scalar value from an array of integers that represent a scalar value that has been encoded with the encoder.

The test cases use the decode method to obtain the original scalar value and contain 8 different sets of encoded values (output1 through output8). The scalar's lowest and maximum values, the number of bits utilized for encoding, the size of each bit, and whether or not the encoding is periodic are all test parameters.

Each of the eight test scenarios is iterated through in the foreach loop, which prints the encoded output and the decoded input for each case. This makes it possible to manually check that the decode method is working correctly.

ScalarEncoder\_EncodeIntoArray\_RangeOfInputValues\_ReturnsCorrectArrays: This code looks to be testing the ScalarEncoder class EncodeIntoArray function. With an increment of 0.1, it loops through a set of input values from MinValue to MaxVal ue. It uses the EncodeIntoArray function to retrieve the intended array for each input value, then compares it to the actual array it gets using the same method.

It makes a bitmap image with the expected array in green and the actual array in gray (or red if they are not equal) and saves it to a folder if the expected and actual arrays are not identical. The intended and actual arrays, as well as the input value, are all indicated in the image's accompanying text. This test checks that the ScalarEncoder class's EncodeIntoArray function returns the right arrays for a variety of input values.

**TestGetBucketInfo:** The code is a unit test for the Scalar Encoder class GetBucketInfo function. Using a number of binary bits, the Scalar Encoder is used to encode scalar values into a sparse distributed representation (SDR). A scalar value is passed into the GetBucketInfo method, which returns details about the bucket into which the data belongs.

Using a set of parameters that specify the encoding process, the first test, TestGetBucketInfoNonPeriodic, produces an instance of the Scalar Encoder. The test then looks for a set of scalar values in the bucket information given by the GetBucketInfo function. For values close to the bucket boundaries, outside of the range of valid values, and in the middle of the range of valid values, the test verifies that the bucket information is as expected.

The second test, TestGetBucketInfoPeriodic, is similar to the first test but with the Periodic parameter set to true. This means that the encoding process treats the range of scalar values as a periodic range, where values at the upper and lower bounds of the range are considered adjacent. The test checks the bucket information returned by the GetBucketInfo method for a set of scalar values, including values outside of the range of valid values.

The Console.WriteLine statements are used to print the bucket information for each value to the console for debugging purposes. The CollectionAssert.AreEqual statements are used to compare the expected bucket information with the actual bucket information returned by the GetBucketInfo method, and throw an exception if they do not match.

# TestCase **with Results**

## **Testcase-ScalarEncoderTests**

It Tests the operation of a Scalar Encoder class, which is in charge of converting numerical data into a format that a machine learning algorithm can use.

Four arguments are passed to the Scalar Encoder:

The number of bits used to represent the encoded item is called the width. The width is 25 in this instance.

The smallest value that can be encoded is called MinVal. The minimum value is set to 1 in this instance.

The most significant value that can be encoded is MaxVal . The highest value in this instance is 50.

Radius: The radius of the "smoothing" function that transforms numbers into binary vectors. The radius is set to 2.5 in this instance.

The Data Row attribute is used in the unit test to pass various inputs and anticipated results to the Scalar Encoder class's Encode function.

The Data Row’s first input and anticipated result sets are for an input value of 1.0. An integer array of 25 elements is what is anticipated as the result, with the first 11 elements set to 1 and the rest elements set to 0.

Encoding functions by projecting the input value onto a group of binary bits. The lowest and maximum values define the range of input values that can be encoded, while the width parameter sets the number of bits required for encoding.

The input value of 1.0 is mapped onto a binary vector of 25 bits because it can be encoded in this particular test case (between 1 and 50). The first 11 bits are set to 1 and the remaining bits are set to 0, after the value of 1.0 is "smoothed" using a Gaussian distribution with a standard deviation of 2.5.

## **Testcase -ScalarEncoderUnitTestNonPeriodic**

In order to create a sparse distributed representation (SDR) that can be used as input to a neural network, scalar data (such as temperature, speed, etc.) must first be encoded using the ScalarEncoder. How the scalar value is transferred to an SDR is determined by the encoder's parameters.

The parameters used in this unit test are described in the following concise manner:

input: The intended scalar value to be encoded

Results: The encoder's anticipated SDR output for the specified input value encoder.

Setting: A dictionary that includes the encoder's settings

W: The size of the SDR's bits

N: The SDR's total amount of active bits (0 means that N will be automatically determined based on W and the radius) The largest scalar value that can be encoded is called MaxVal ..The radius of each Gaussian curve used to convert scalar values to SDRs is its width.

Resolution: The encoder's level of resolution (i.e. the smallest difference between two encoded values)

Periodic: a boolean value indicating whether or not the scalar values circle back on themselves (much like a clock)

Whether input values outside the range [MinVal, MaxVal ] should be clipped to the range or not is determined by the boolean value "clipInput."

Name: The encoder is known as IsRealCortex. Model: A boolean value indicating whether or not an algorithm for real-time cortical learning is using the encoder.

The ScalarEncoderUnitTestNonPeriodic method examines if the ScalarEncoder class's Encode method returns the expected SDR when given a scalar input value and its expected SDR output. If the SDR output matches the expected SDR, the test passes. If not, the test fails.

## **Testcase-ScalarEncoderUnitTestPeriodic**

Encoder Settings: A dictionary object containing the encoder settings. These settings include a number of different parameters, including the size of the output array, the number of bits to be used for encoding, and the minimum and maximum values for the input space. These options control how the input data is converted into a binary representation that the HTM network can use. the input value that needs to be encoded. An array of scalar values or a single scalar value can be used here.

When to clip the input value to the range specified by the encoder settings is determined by the boolean value clipInput. Input values that are outside of the minimum and maximum range will be clipped to the closest boundary value if the value is set to true. If set to false, if the input value is outside of range, the encoder will throw an exception.

ReturnArray: A boolean value that indicates whether the encoded value should be returned as a sparse array of (index, value) pairs or as an array of bits. The encoder will return the bits as a sparse array, with each element being a bit, if true is set. where each component is a tuple of the form (index, value), where index is the index of the active bit and value is that bit's value (usually 1). The bits will be returned by the encoder as a dense binary array, where each member denotes the value of a single bit, if the value is set to false.

## **Testcase-SeasonEncoderTest:**

This method serves as a unit test for the SeasonEncoder class. It requires two parameters: expectedResult, an array of integers representing the expected output of the encoder for the given input, and input, an integer value representing the day of the year (1-366).

The method first generates a new ScalarEncoder object with settings particular to the SeasonEncoder, including the lowest and maximum values that the encoder can handle as well as the width of each bucket (W) and the number of buckets (N) (MinVal and MaxVal ).

The result is saved in the result variable after the input value has been encoded using the ScalarEncoder object's Encode method. Additionally, the procedure prints the value of the input, the result vector, and the expected Result vector for debugging purposes. The expectedResult and result vectors are then compared using the Assert.IsTrue method. The test fails if they are not equal.

## **TimeTickEncodingTest:**

"TimeTickEncodingTest" is a unit test method that examines the encoding of time ticks using a predetermined input value and anticipated output vector.

The procedure accepts two inputs:

The input value to be encoded is represented by a double input anticipated. After encoding the input value, the result represents the anticipated result vector.

The CortexNetworkContext class, which is used to specify the context for a neural network, is initially created by the method as a new instance.

The ScalarEncoder class is then built with the following parameters:

W: 21,N: 1024 ,MinVal: 0.0, Radius: -1.0,MaxVal : The number of days in a year as a whole (as determined by subtracting the current date and time from the same date and time one year ago). Periodical: true, Name: "season, “clip input:True.  The procedure then iterates through a range of values up to the encoder's maximum value (specified by the MaxVal parameter above), encoding the input using the encoder for each value and outputting the encoded vector to the debug console.

The approach concludes by claiming that the amount of "on" bits—that is, bits with a value of 1—in the resulting vector corresponds to the value of the W parameter mentioned earlier.

## **Testcase-ScalarEncodingGetBucketIndexNonPeriodic:**

This code is an experiment to encode a set of numbers and produce bitmap output. It encodes a decimal number using a ScalarEncoder object with non-periodic settings. The encoder has the following settings:

"W": 21 - The width of the output vector.

"N": 1024 - The number of bits to use to encode the input range. “Radius": -1.0 - The radius of the output vector.

"MinVal": 0.0 - The minimum value of the input range.

"MaxVal ": 100.0 - The maximum value of the input range.

"Periodic": false - Whether the encoder should wrap values around the ends of the input range."ClipInput": false - Whether the input values outside the input range should be clipped to the minimum or maximum value. The code then loops through a range of decimal numbers, encodes each number using the encoder, and gets the bucket index of the number using the GetBucketIndex() method of the encoder. The result is then displayed as a bitmap image, with the value of the input and its bucket index as text.The bitmap image is saved in a folder named "ScalarEncodingGetBucketIndexNonPeriodic".

## **TestCase- ScalarEncodingGetBucketIndexPeriodic**

This code is an experiment to encode a set of numbers and produce bitmap output. It encodes a decimal number using a ScalarEncoder object with periodic settings. The encoder has the following settings:

"W": 21 - The width of the output vector.

"N": 1024 - The number of bits to use to encode the input range.

"Radius": -1.0 - The radius of the output vector.

"MinVal": 0.0 - The minimum value of the input range.

"MaxVal ": 100.0 - The maximum value of the input range.

"Periodic": true - Whether the encoder should wrap values around the ends of the input range.

"ClipInput": false - Whether the input values outside the input range should be clipped to the minimum or maximum value. The code then loops through a range of decimal numbers, encodes each number using the encoder, and gets the bucket index of the number using the GetBucketIndex() method of the encoder. The result is then displayed as a bitmap image, with the value of the input and its bucket index as text. The bitmap image is saved in a folder named "ScalarEncodingGetBucketIndexPeriodic".

## **Testcase-GenerateRangeDescription:**

The ScalarEncoder class GenerateRangeDescription () function. The method takes a list of ranges, where each range is represented by a tuple of two doubles and produces a string with a human-readable description of those ranges.

In the test, the GenerateRangeDescription () method receives three different sets of ranges as input. The Assert.AreEqual() method compares the method's actual output to the predicted output for each combination of ranges.

The tuples (1.0, 3.0) and are included in the first set of ranges (7.0, 10.0). The desired results are "1.00-3.00, 7.00-10.00."

The tuple is the single item in the second set of ranges (2.5, 2.5). The anticipated result is "2.50."

The tuples (1.0, 1.0) and are included in the third set of ranges (5.0, 6.0) if all the expected outputs match the actual outputs, then the test passes.

## **Testcases- ClosenessScores:**

This is a unit test method that tests the ClosenessScores method of the ScalarEncoder class.

The test arranges a new instance of ScalarEncoder with specific parameters. Then it sets up expected and actual value arrays with one value each. It then sets a boolean flag to true to indicate that the values are fractional. The expected closeness score is set to 0.99.

Finally, the test asserts that the closeness score of the expected and actual value arrays is equal to the expected value with a tolerance of 0.01.

Params:

“expValues”: an array of expected values, set to { 50 }

“actValues”: an array of actual values, set to { 51 }

fractional: a boolean flag indicating whether the values are fractional, set to true expected. Closeness: the expected closeness score, set to 0.99

The four test methods are testing the ClosenessScores method of the ScalarEncoder class with different input values and encoder parameters.

In ClosenessScorestest1, a periodic encoder is used with a target value of 50 and an actual value of 51. The expected closeness is 0.99, meaning that the actual value is very close to the target value.

In ClosenessScorestest2, a non-periodic encoder is used with a target value of 50 and an actual value of 50.3. The expected closeness is also 0.99 with a maximum difference of 0.01.

In ClosenessScorestest3, a periodic encoder is used with a target value of 50 and an actual value of 70. The expected closeness is 0.8, meaning that the actual value is farther from the target value compared to the previous two tests.

In ClosenessScoresTest4, a non-periodic encoder is used with a target value of 25 and an actual value of 75. The expected closeness is 0.5, meaning that the actual value is very far from the target value.

The main difference between the tests is the encoder parameters used, which affect the encoding of input values and the closeness scores calculation.

## **Testcases- initialization:**

This code is a unit test for the ScalarEncoderExperimental class, which tests the initialization and property setting of the encoder.

The test initializes the encoder with default settings obtained from getDefaultSettingsmethod, then changes the value of the Name property and adds a new property TestProp1. The code then sets a property abc using the indexer syntax of the encoder.

The test then checks that the values of the TestProp1, Name, and abc properties are as expected.

Overall, this test verifies that the encoder can be initialized with custom settings and properties can be set using the indexer syntax.

The three methods are similar in that they create an instance of the ScalarEncoderExperimental class and initialize its properties with a dictionary of settings. However, they differ in the way they create and initialize the encoder instance.

InitTest2() creates an instance of the CortexNetworkContext class and then uses it to create the encoder instance using ctx.CreateEncoder(). It then checks that each key-value pair in the encoderSettings dictionary is set correctly in the encoder instance.

InitTest3() also creates an instance of the CortexNetworkContext class and uses it to create the encoder instance, but it modifies the encoderSettings dictionary by changing the value of the "Name" key and adding a new key-value pair "TestProp1":"hello". It then sets another property "abc":"1" in the encoder instance and checks that all three key-value pairs are set correctly.

InitTest4() creates an instance of the ScalarEncoderExperimental class directly and then initializes its properties with the encoderSettings dictionary using the encoder.Initialize() method. It then sets the "abc":"1" property in the encoder instance and checks that all three key-value pairs are set correctly.

In terms of parameters and values used, all three methods use the getDefaultSettingsmethod to obtain a dictionary of default settings for the encoder instance. The encoderSettings dictionary is then modified in some way, either by changing the value of an existing key or adding a new key-value pair. The modified encoderSettings dictionary is then used to create and initialize the encoder instance. Finally, the methods check that the properties of the encoder instance have been set correctly.

## **Testcase-Static Dictionary:**

This code defines a private static method called getDefaultSettings which returns a Dictionary containing default settings for a scalar encoder. A scalar encoder is a type of encoder used in machine learning and neuroscience to transform continuous variables into sparse binary representations suitable for use in a neural network.

The default settings are:

W (int): the number of bits that are set to encode a single value - the "width" of the output signal. The value used here is 11.

N (int): the number of bits in the output. Must be greater than or equal to W. The value used here is 0.

MinVal (double): the minimum value of the input signal. The value used here is 1.

MaxVal (double): the upper bound of the input signal. The value used here is 100.

Radius (double): two inputs separated by more than the Radius have non-overlapping representations. Two inputs separated by less than the Radius will in general overlap in at least some of their bits. You can think of this as the radius of the input. The value used here is 0.

Resolution (double): two inputs separated by greater than, or equal to the Resolution are guaranteed to have different representations. The value used here is 0.15.

Periodic (bool): if true, then the input value "wraps around" such that MinVal = MaxVal. For a periodic value, the input must be strictly less than MaxVal, otherwise MaxVal is a true upper bound. The value used here is true.

ClipInput (bool): if true, non-periodic inputs smaller than MinVal or greater than MaxVal will be clipped to MinVal/MaxVal. The value used here is true.

Name (string): the name of the scalar encoder. The value used here is "TestScalarEncoder".

IsRealCortexModel (bool): a flag indicating whether the encoder is being used in a real cortical model. The value used here is false. InitTest4() creates an instance of the ScalarEncoderExperimental class directly and then initializes its properties with the encoderSettings dictionary using the encoder.Initialize() method. It then sets the "abc":"1" property in the encoder instance and checks that all three key-value pairs are set correctly.

In terms of parameters and values used, all three methods use the getDefaultSettings method to obtain a dictionary of default settings for the encoder instance. The encoderSettings dictionary is then modified in some way, either by changing the value of an existing key or adding a new key-value pair. The modified encoderSettings dictionary is then used to create and initialize the encoder instance. Finally, the methods check that the properties of the encoder instance have been set correctly.

## **Testcase-ScalarEncodingEncodeIntoArray:**

This program is a unit test method for the ScalarEncoder class's EncodeIntoArray method. A collection of input values are encoded into a bitmap, and the output array is checked to make sure the input value was encoded correctly.

The following variables are used in the constructor of the ScalarEncoder:

"W": 21, the "width" of the output signal, which is the number of bits set to encode a single value.

The output's "N": 1024 bit count must be larger than or equal to W.

"Radius": 1.0, non-overlapping representations between two inputs separated by greater than the radius.

The input signal's minimal value, "MinVal," is 0.0.

The input signal's upper bound, "MaxVal ," is 100.0.

If "periodic" is true, the input value is determined.

"wraps around" such that MinVal = MaxVal . For a periodic value, the input must be strictly less than MaxVal , otherwise MaxVal is a true upper bound.

"Name": "scalar\_periodic", the name of the encoder.

"ClipInput": false, if true, non-periodic inputs smaller than MinVal or greater than MaxVal will be clipped to MinVal/MaxVal ..The inputs array contains the input values to be encoded: { 1, 25, 60, 75, 99 }.

The method loops over each input value, calls EncodeIntoArray, verifies the output array, creates a bitmap of the encoded value, and saves it to disk.

The bitmap is created using the NeoCortexUtils.DrawBitmap method with the following parameters:

twoDimArray: the 2D array containing the encoded value.

1024: the width of the bitmap.

1024: the height of the bitmap.

${outFolder}\\encoded\_{input}.png: the filename of the bitmap to be saved.

Color.Gray: the color of the background.

Color.Green: the color of the foreground.

text: a string containing the input value and the encoded value.

Finally, the method verifies that EncodeIntoArray returns 0 for each input value.

## **Testcase-ScalarEncodingDecode:**

This unit test method examines the ScalarEncoder class' decode method. The method accepts some decoding-related parameters as well as encoded input values in the form of an array of integers (output1, output2, etc.).

The variables employed are:

MinVal: the input range's minimal value (0)

MaxVal: The input range's highest value (100)

The encoding's bit count is n. (14)

w: each encoding bin's width (3.0)

If the input range is periodic, a boolean value called periodic will be present (true)

The procedure goes over the test cases and applies the decode technique to each one in turn. The console is then printed with the decoded input values for validation.

## **Testcase- ScalarEncoder\_EncodeIntoArray\_RangeOfInputValues\_ReturnsCorrectArrays:**

This program is a unit test for the ScalarEncoder class EncodeIntoArray function. The test confirms that, for a range of input values between a specified minimum and maximum value, the method returns the right array.

The test constructs an instance of the encoder using the ScalarEncoder constructor with the following values: numBits = 101, MinValue = 0, MaxValue = 100, and period = MaxValue – MinValue. The test then runs the EncodeIntoArray function to obtain the anticipated and actual output arrays while iterating through a range of input values between MinValue and MaxVal ue in steps of 0.1.

The test verifies that each element in the expected array and the actual array are the same length. is equal to the corresponding element in the actual array. If the arrays are not equal, the test creates a PNG image of the expected and actual arrays side-by-side with the input value as the filename.

## **Testcase- TestGetBucketInfo:**

The code is a unit test in C# that tests the functionality of the GetBucketInfo method of the ScalarEncoder class. The ScalarEncoder is a class in the Hierarchical Temporal Memory (HTM) library that encodes scalar values into sparse distributed representations suitable for use as inputs to HTM algorithms.

The first test, TestGetBucketInfoNonPeriodic, tests the method with a non-periodic encoder. The encoder is initialized with the following parameters:

W: The number of bits in the output representation of each input value. In this case, it is set to 21.

N: The number of bits in the output representation of the entire encoder. In this case, it is set to 100.

Radius: The radius of the input space. A value of -1.0 means that the radius is automatically determined based on the range of the input space.

MinVal: The minimum value of the input space. In this case, it is set to 0.0.

MaxVal: The maximum value of the input space. In this case, it is set to 100.0.

Periodic: Whether the input space is periodic or not. In this case, it is set to false, indicating that the input space is not periodic.

Name: The name of the encoder. In this case, it is set to "scalar nonperiodic".

ClipInput: Whether to clip input values to the range of the input space or not. In this case, it is set to false, indicating that input values outside the range of the input space will be encoded.

NumBuckets: The number of buckets used to encode the input space. In this case, it is set to 100.

The test cases involve passing different input values to the GetBucketInfo method and verifying that the returned bucket information is correct. For example, the first test case passes the value 49.0 to the method and expects the bucket information to be [49, 50, 49, 50]. This means that the input value is in the bucket with index 49, its center is at bucket index 50, its start is at bucket index 49, and its end is at bucket index 50.

The second test, TestGetBucketInfoPeriodic, tests the method with a periodic encoder. The encoder is initialized with similar parameters as before, except that the Periodic parameter is set to true, indicating that the input space is periodic.

The test cases are similar to the non-periodic case, except that the periodicity of the input space affects the bucket information. For example, the first test case passes the value 49.0 to the method and expects the bucket information to be [49, 49, 49, 50]. This means that the input value is in the bucket with index 49, its center is at bucket index 49, its start is at bucket index 49, and its end is at bucket index 50.

# **Conclusion**

In summary, the Scalar Encoder with Buckets implementation offers a practical method for encoding continuous data into a sparse distributed representation appropriate for usage in HTM (Hierarchical Temporal Memory) systems. The Scalar Encoder with Buckets can recognize intricate patterns in the input data and maintain spatial and temporal linkages by bucketing the range of input values into a predetermined number of buckets and encoding the values into binary arrays using overlapping Gaussian curves. The encoder's parameters, such as the number of buckets and the width of the Gaussian curves, can be changed to customize the encoding process for various applications. The Scalar Encoder with Buckets' effective implementation indicates its potential for application in a its potential for application in variety of domains, including anomaly detection, prediction, and classification. The

The test cases performed effectively during the execution of the scalar encoder with buckets. The test cases covered a range of scenarios including encoding and decoding values within the min/max range, decoding values outside of the min/max range, encoding and decoding values with periodicity, and encoding and decoding values with different bit widths. The tests were designed to ensure that the scalar encoder with buckets was functioning as expected and was able to accurately encode and decode scalar values.

ScalarEncoder class passed all of the test cases, indicating that it is operating appropriately because it delivered the desired outcomes for each input. Scalar values could be encoded into arrays by the encoder and then decoded to their original values by the decoder. It is possible that the ScalarEncoder is dependable for encoding and decoding scalar values because the test cases included a variety of inputs, including edge cases.

The implementation of the scalar encoder with buckets was successful, as it provided a flexible and efficient method for encoding scalar values into sparse distributed representations. The use of buckets allowed for a more fine-grained control over the encoding process, which allowed for better encoding of values with high variance. Additionally, the use of periodicity allowed for the encoder to more effectively encode values that had a cyclic nature.

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