Module Interface Specification for ANN (Artificial Neural Network)

Tanya Djavehrpour

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1 Revision History

Date	Version	Notes
March 19	1.0	Initial Draft

2 Symbols, Abbreviations and Acronyms

See SRS Documentation Djavaherpour (2024b) at HERE.

Contents

1	Rev	vision 1	History						
2	Symbols, Abbreviations and Acronyms								
3	Introduction								
4	Notation								
5	Mo	dule D	Decomposition						
6	MIS	S of A	NN Control Module						
	6.1	Modu	le						
	6.2	Uses							
	6.3	Synta	x						
		6.3.1	Exported Constants						
		6.3.2	Exported Access Programs						
	6.4	Seman	ntics						
		6.4.1	State Variables						
		6.4.2	Environment Variables						
		6.4.3	Assumptions						
		6.4.4	Access Routine Semantics						
		6.4.5	Local Functions						
7	MIS	S of Sa	aved ANN Model Module						
	7.1	Modu	le						
	7.2	Uses							
	7.3	Synta	x						
		7.3.1	Exported Constants						
		7.3.2	Exported Access Programs						
	7.4	Seman	$\operatorname{ntics}^{\circ}$						
		7.4.1	State Variables						
		7.4.2	Environment Variables						
		7.4.3	Assumptions						
		7.4.4	Access Routine Semantics						
		7.4.5	Local Functions						
3	MIS	S of O	utput Module						
	8.1	Modu	de						
	8.2								
	8.3		X						
		8.3.1	Exported Constants						
		8.3.2	Exported Access Programs						

	8.4	Seman	ntics					 							8
		8.4.1	State Variables					 							8
		8.4.2	Environment Variables					 							8
		8.4.3	Assumptions					 							8
		8.4.4	Access Routine Semantics					 							8
		8.4.5	Local Functions												9
9	MIS	of In	put Classifier Module												10
	9.1		le					 							10
	9.2														10
	9.3	Syntax	x					 							10
		9.3.1	Exported Constants												10
		9.3.2	Exported Access Programs												10
	9.4	Seman	•												10
		9.4.1	State Variables												10
		9.4.2	Environment Variables					 							10
		9.4.3	Assumptions												10
		9.4.4	Access Routine Semantics												11
10		-	put Image Module												12
			le												12
															12
	10.3		X												12
			Exported Constants												12
			Exported Access Programs												12
	10.4		ntics												12
			State Variables												12
			Environment Variables												12
			Assumptions												12
			Access Routine Semantics												13
		10.4.5	Local Functions			•		 	•	•	 ٠	•	•	•	13
11	MIS	of Tr	aining Model Module												14
			le					 							14
	11.2	Uses						 							14
			x												14
			Exported Constants												14
			Exported Access Programs												14
	11.4		ntics												14
			State Variables												14
			Environment Variables												14
			Assumptions												14
			Access Routine Semantics												14

		11.4.5	Local Functions	15
12	Inpı	ıt Prep	paring and Preprocessing Module	16
	12.1	Module	e	16
	12.2	Uses .		16
	12.3	Syntax		16
		12.3.1	Exported Constants	16
		12.3.2	Exported Access Programs	16
	12.4	Semant	tics	16
		12.4.1	State Variables	16
		12.4.2	Environment Variables	16
		12.4.3	Assumptions	16
		12.4.4	Access Routine Semantics	16
		12.4.5	Local Functions	17
	_	_		
13		_	aring and Preprocessing Module	18
			e	18
				18
	13.3			18
			Exported Constants	18
			Exported Access Programs	18
	13.4		tics	18
			State Variables	18
			Environment Variables	18
			Assumptions	19
			Access Routine Semantics	19
		13.4.5	Local Functions	19
11	МТ	of Tra	aining and Testing Module	21
14			6	21
	14.0		Exported Constants	21
			Exported Access Programs	21
	1/1/		tics	21
	14.4		State Variables	21
			Environment Variables	$\frac{21}{21}$
			Assumptions	$\frac{21}{22}$
			Access Routine Semantics	22
			Local Functions	$\frac{22}{23}$

3 Introduction

The following document details the Module Interface Specifications for ANN (Artificial Neural Network). This document specifies how every module is interfacing with every other parts.

Complementary documents include the System Requirement Specifications (SRS) Djavaherpour (2024b) and Module Guide (MG) Djavaherpour (2024a). The full documentation and implementation can be found at Github repository for ANN.

4 Notation

The structure of the MIS for modules comes from Hoffman and Strooper (1995), with the addition that template modules have been adapted from Ghezzi et al. (2003). The mathematical notation comes from Chapter 3 of Hoffman and Strooper (1995). For instance, the symbol := is used for a multiple assignment statement and conditional rules follow the form $(c_1 \Rightarrow r_1|c_2 \Rightarrow r_2|...|c_n \Rightarrow r_n)$.

The following table summarizes the primitive data types used by ANN.

Data Type	Notation	Description
1D array	\mathbf{A}_i	A linear sequence of elements
2D matrix	\mathbf{M}_{ij}	A collection of elements arranged in rows and columns
3D matrix	\mathbf{M}_{ijk}	A structure composed of elements arranged in a grid with three dimensions
boolean	bool	True or False
dictionary	dict	A dictionary to store data as keys and val-
		ues
string	str	A sequence of characters
character	char	a single symbol or digit
integer	\mathbb{Z}	a number without a fractional component
		in $(-\infty, \infty)$
positive Integer	\mathbf{Z}_{+}	a number without a fractional component
		in $(0, \infty)$
natural number	\mathbb{N}	a number without a fractional component
		in $[1, \infty)$
real	\mathbb{R}	any number in $(-\infty, \infty)$

The specification of ANN uses some derived data types: sequences, strings, and tuples. Sequences are lists filled with elements of the same data type. Strings are sequences of

characters. Tuples contain a list of values, potentially of different types. In addition, ANN uses functions, which are defined by the data types of their inputs and outputs. Local functions are described by giving their type signature followed by their specification.

5 Module Decomposition

The following table is taken directly from the Module Guide Djavaherpour (2024a) document for this project.

Level 1	Level 2
Hardware-Hiding	
Behaviour-Hiding	ANN Control Module Saved ANN Model Module Output Module Input Classifier Module Input Image Module Training Model Module
Software Decision	Input Preparing and Preprocessing Module Data Preparing and Preprocessing Module Training and Testing Module

Table 1: Module Hierarchy

6 MIS of ANN Control Module

6.1 Module

main

6.2 Uses

- Hardware-Hiding Module
- Saved ANN Model Module (7)
- Output Module (8)

6.3 Syntax

6.3.1 Exported Constants

None.

6.3.2 Exported Access Programs

Name	In	Out	Exceptions
main	-	-	=

6.4 Semantics

6.4.1 State Variables

None.

6.4.2 Environment Variables

None.

6.4.3 Assumptions

- The ANN Control Module assumes that the Hardware-Hiding Module, Saved ANN Model Module, and Output Module are implemented according to their specifications. However, it does include error handling to manage unexpected behaviors or failures in these modules.
- The system environment (operating system, hardware) is assumed to be stable. Also, essential libraries and dependencies are presumed to be correctly installed and configured.

6.4.4 Access Routine Semantics

main():

• transition: Initializes the program.

Note: As the ANN Control Module mainly serves as a coordinator between different modules without maintaining its own state or producing output, its primary function is to ensure the correct sequence of operations and interactions between these modules. It relies on the robustness of the called modules' error handling.

6.4.5 Local Functions

7 MIS of Saved ANN Model Module

7.1 Module

model

7.2 Uses

- Hardware-Hiding Module
- Training and Testing Module (14)

7.3 Syntax

7.3.1 Exported Constants

None.

7.3.2 Exported Access Programs

Name	In	Out	Exceptions
save_model	-	bool	PermissionError
${\tt load_trained_classifier}$	\mathbf{A}_i, str	\mathbf{Z}_{+}	-

7.4 Semantics

7.4.1 State Variables

None.

7.4.2 Environment Variables

None.

7.4.3 Assumptions

None.

7.4.4 Access Routine Semantics

save_model():

- transition: Writes the current state of the ANN model (weights and biases) to a .npy file as a *dict*.
- output: Returns True if model is saved successfully.

• exception: Raises PermissionError if the module lacks the necessary permissions to write to modelFile. It may also raise an IOError if there are issues with the file system, such as insufficient storage space.

load_trained_classifier(input_image, model_name):

- transition: Loads a trained classifier and uses it to predict a class for the input image.
- output: The predicted class as \mathbf{Z}_{+} , based on the vector of input image and name of the saved model.
- exception: None.

7.4.5 Local Functions

load_model(file_name):

- transition: Load model parameters from a specified code.npy file.
- output: The model parameters stored in the file as a *dict*.
- exception: Raises FileNotFoundError if modelFile does not exist or cannot be accessed. Additionally, an exception may be raised for data corruption or format mismatch, indicating issues with the integrity or compatibility of the stored model data.

8 MIS of Output Module

8.1 Module

output

8.2 Uses

- $\bullet\,$ Hardware-Hiding Module
- Input Classifier Module (9)

8.3 Syntax

8.3.1 Exported Constants

None.

8.3.2 Exported Access Programs

Name	In	Out	Exceptions
set_class_name	-	str	-
${\tt save_feedback}$	-	-	-

8.4 Semantics

8.4.1 State Variables

class_name: The str as the name of the class to which the image has been classified.

8.4.2 Environment Variables

None.

8.4.3 Assumptions

None.

8.4.4 Access Routine Semantics

set_class_name():

- transition: Determines the class name of an image using the Input Classifier Module (9).
- output: he class name determined by the Input Classifier Module (9).

• exception: None.

save_feedback():

• transition: ollects and saves user feedback on the classification result to a text file.

• output: None.

• exception: None.

8.4.5 Local Functions

append_to_file(file_path, sentence):

- transition: Appends a given feedback to a file and prints a success message or error message depending on the outcome of the file operation.
- output: None.
- exception: Raises FileNotFoundError if the destination .txt file is not found. IOError will be raised, if there is a problem with writing to the file.

9 MIS of Input Classifier Module

9.1 Module

classifier

9.2 Uses

- Hardware-Hiding Module
- Saved ANN Module (7)
- Input Preparing and Preprocessing Module (12)

9.3 Syntax

9.3.1 Exported Constants

None.

9.3.2 Exported Access Programs

Name	In	Out	Exceptions
set_image_pixe	el -	-	-
${ t get_class}$	-	str	-

9.4 Semantics

9.4.1 State Variables

- input_image: An array of input image's pixels after preprocessing.
- class_name: The str as the name of the class to which the image has been classified.

9.4.2 Environment Variables

None.

9.4.3 Assumptions

9.4.4 Access Routine Semantics

set_image_pixels():

- transition: Receives the array of input image from Input Preparing and Preprocessing Module (12) and saves in input_image.
- output: None.
- exception: None.

get_class():

- transition: Classifies the preprocessed image using a pretrained model from (7) and maps the output to a class name.
- output: The class of the input image.
- exception: None.

10 MIS of Input Image Module

10.1 Module

input

10.2 Uses

• Hardware-Hiding Module

10.3 Syntax

10.3.1 Exported Constants

- HEIGHT: A value (\mathbf{Z}_{+}) describing acceptable height of input image (currently 32).
- WIDTH: A value (\mathbf{Z}_{+}) describing acceptable width of input image (currently 32).
- IMAGE_FORMAT: A list of strings (str) of acceptable types of input image (currently PNG and JPEG).

10.3.2 Exported Access Programs

Name	In	Out	Exceptions
set_image	-	\mathbf{M}_{ijk}	FileNotFoundError,
			InvalidSize,
			${\tt InvalidFormat}$

10.4 Semantics

10.4.1 State Variables

None.

10.4.2 Environment Variables

None.

10.4.3 Assumptions

10.4.4 Access Routine Semantics

set_image():

- transition: Receives the input image from end user and returns its matrix as inputImage.
- output: None.
- exception: Raises FileNotFoundError if inputImagePath does not exist or cannot be accessed. Also, InvalidSize is raised when the size of input image in not compatible with HEIGHT or WIDTH. Additionally, InvalidFormat is thrown if the input image's format is not compatible with IMAGE_FORMAT.

10.4.5 Local Functions

11 MIS of Training Model Module

11.1 Module

training_model

11.2 Uses

• Data Preparing and Preprocessing Module (13)

11.3 Syntax

11.3.1 Exported Constants

- LAYERS_NUMBER: A value (\mathbf{Z}_+) describing the number of neural network's layers.
- LAYERS_NEURONS: An array including each layer's number of neurons.

11.3.2 Exported Access Programs

Name	In	Out	Exceptions
create_gradients	-	\mathbf{A}_i	-

11.4 Semantics

11.4.1 State Variables

None.

11.4.2 Environment Variables

None.

11.4.3 Assumptions

None.

11.4.4 Access Routine Semantics

create_gradients():

- transition: Creates zero arrays for all needed gradients based on the LAYERS_NUMBER and LAYERS_NEURONS.
- output: All gradients' zero vector (\mathbf{A}_i)
- exception: None.

11.4.5 Local Functions

12 Input Preparing and Preprocessing Module

12.1 Module

 ${\tt input_prep}$

12.2 Uses

Data Preparing and Preprocessing Module (13) Input Image Module (10)

12.3 Syntax

12.3.1 Exported Constants

None.

12.3.2 Exported Access Programs

Name	In	Out	Exceptions
get_input	-	\mathbf{A}_i	-

12.4 Semantics

12.4.1 State Variables

• input_image: Incling input image pixels.

12.4.2 Environment Variables

None.

12.4.3 Assumptions

None.

12.4.4 Access Routine Semantics

input_prep():

- transition: Prepares and preprocess the input image to change it the way model can use it to predicts the class.
- output: an A_i including prepared and preprocessed input image.
- exception: None.

12.4.5 Local Functions

- set_image_pixels():
 - transition: Loads an image using the Input Image Module (10). and store it in input_image.
 - output: None.
 - exception: None.
- rgb2gray():
 - transition: Uses Data Preparing and Preprocessing Module (13) to convert RGB data (input_image) into grayscale to reduce the complexity.
 - output: None.
 - exception: None.
- prep_pixels():
 - transition: Normalizes grayscaled (input_image) to change the range of data between 0 and 1, using Data Preparing and Preprocessing Module (13).
 - output: None.
 - exception: None.
- flat_data():
 - transition: Data is flatten since (input_image) should be vectorized with the size of 1024. After grayscaling input image is a \mathbf{M}_{ij} , this should be an \mathbf{A}_i to be used by implemented model.
 - output: None.
 - exception: None.

13 Data Preparing and Preprocessing Module

13.1 Module

data

13.2 Uses

None.

13.3 Syntax

13.3.1 Exported Constants

None.

13.3.2 Exported Access Programs

Name	In	Out	Exceptions
get_dataset	-	$[\mathbf{M}_{ij},\mathbf{A}_i],[\mathbf{M}_{ij},\mathbf{A}_i]$	-
rgb2gray	$[\mathbf{M}_{ij}],\ bool$	-	-
prep_pixels	$[\mathbf{M}_{ij}]$	-	-

13.4 Semantics

13.4.1 State Variables

- train_data: Data structure holding train data images and their labels. Since train images after processing are vectors (\mathbf{A}_i) , a list of these images is a matix (\mathbf{M}_{ij}) . Alos, labels are saving in a vector (\mathbf{A}_i) . Consequently, this data structure is a list in $[\mathbf{M}_{ij}, \mathbf{A}_i]$ format.
- test_data: Data structure holding test data images and their labels. Since test images after processing are vectors (\mathbf{A}_i) , a list of these images is a matix (\mathbf{M}_{ij}) . Alos, labels are saving in a vector (\mathbf{A}_i) . Consequently, this data structure is a list in $[\mathbf{M}_{ij}, \mathbf{A}_i]$ format.
- train_images: A list incling traing data images.
- test_images: Alist incling test data images.

13.4.2 Environment Variables

13.4.3 Assumptions

None.

13.4.4 Access Routine Semantics

get_dataset():

- transition: Processes and retrieve the training and testing datasets.
- output: train_data, test_data
- exception: None.

rgb2gray(images, input_image):

- transition: Converts RGB data into grayscale in order to reduce complexity. First input is a numpy array representing one or more images in RGB format. If a single image, it should be a \mathbf{M}_{ijk} array (height, width, color_channels). If multiple images, it should be a 4D array (number_images, height, width, color_channels). Second input as a bool. It is a flag to indicate if the provided 'images' parameter is a single image (True) or a batch of images (False). Default is False.
- output: The transformed grayscale images. If 'images' was a single image, the return is a \mathbf{M}_{ij} (height, width). If 'images' was a batch of images, the return is a \mathbf{M}_{ijk} array (number_images, height, width).
- exception: None.

prep_pixels(images):

- transition: Normalizes grayscaled images to change the range of data between 0 and 1.
- output: Normalized images.
- exception: None.

13.4.5 Local Functions

- save_data():
 - transition: Downloads the file from Google Drive and saves it locally.
 - output: None.
 - exception: This function raises UnableToDownload, when there is a problem with downloading or extracting data.
- load_data():

- transition: Loads train_data and test_data, ecnods the labels. This function updates train_images and test_images as well.
- output: None.
- exception: This function raises UnableToDownload, when there is a problem with downloading or extracting data.

• flat_data(images):

- transition: Data is flatten since images should be vectorized with the size of 1024. After grayscaling each image is a \mathbf{M}_{ij} . These images should be an \mathbf{A}_i to be used in training and testing process by implemented model.
- output: Flattened images.
- exception: None.

• shuffle_data(data, images):

- transition: Combines images with their corresponding labels, one-hot encodes the labels, and then shuffles the combined data for randomness.
- output: train_data and test_data.
- exception: A tuple as $[\mathbf{M}_{ij}, \mathbf{A}_i]$, where each tuple contains a shuffled image and its label.

14 MIS of Training and Testing Module

The details of functions using here are described in SRS document Djavaherpour (2024b).

14.1 Module

train_and_test

14.2 Uses

- Training Model Module (11)
- Data Preparing and Preprocessing Module (13)

14.3 Syntax

14.3.1 Exported Constants

- BATCH_SIZE: The partition size of the dataset for each step of learning, typically a power of two.
- LEARNING_RATE: Speed at which the model learns; controls adjustments to the weights.
- EPOCHS: Total number of training cycles through the entire dataset.

14.3.2 Exported Access Programs

Name	In	Out	Exceptions
train	-	$dict, \mathbf{A}_i, \mathbf{Z}_+, \mathbf{Z}_+$	-
result	$dict, \mathbf{A}_i$	-	-
calculate_percentage_of_accuracy	$\mathbf{M}_{ij},dict,bool$	-	-
	or $[\mathbf{M}_{ijk}, \mathbf{M}_i],$	-	-
	dict,bool		

14.4 Semantics

14.4.1 State Variables

• layers: Layers dimensions are defined based on the model architecture and are saved in an A_i named layers.

14.4.2 Environment Variables

14.4.3 Assumptions

None.

14.4.4 Access Routine Semantics

train():

- transition: Trains the model based on constant variables defined in 11. Gradient arrays and train data are needed for training. The training process involves initializing parameters, performing forward and backward passes for each batch, and updating the parameters using the computed gradients. The process is repeated for a specified number of epochs.
- output: The final trained parameters, cost for each epoch, start time, and end time of training.
- exception: None.

result(epochs_costs, trained_params):

- transition: Displays the accuracy percentages for training and testing datasets and plots the training cost over epochs. This method calculates and prints the accuracy on both the training and testing datasets using the trained model parameters. It also generates a plot of the training costs over epochs to visually assess the model's learning progress.
- output: This method prints the accuracy percentages for the training and test datasets to the console and displays a line plot of the training costs over epochs, showing changes in cost with each epoch.
- exception: None.

calculate_percentage_of_accuracy(data, parameters, input_image = False):

- transition: Calculates the accuracy of the neural network model on a given dataset or finds the predicted class of an input image. Accuracy is determined by comparing the predicted labels against the actual labels and calculating the percentage of correct predictions.
- output: Accuracy percentage if input_image is False, or class index if True.
- exception: None.

14.4.5 Local Functions

set_layers():

- transition: layers is set based on Train Model Module (11) and the gradients.
- output: None.
- exception: None.
- sigmoid(x):
 - transition: Claculates sigmoid function for x, as the activation function.
 - output: Sigmoid value of x.
 - exception: None.
- initialize_parameters(layers):
 - transition: Allocates random normal weights (\mathbf{M}_{ij}) and zero biases (\mathbf{A}_i) for each layer.
 - output: Returns a dictionary (named parameters) that the keys define weights or biases, and the values are allocated random numbers to each of them.
 - exception: None.
- compute_cost(predicted, actual):
 - transition: Claculates the sum of the squared errors based on the predicted and actual values.
 - output: Returns the sum of the squared errors
 - exception: None.
- feed_forward(predicted, parameters, layersNumb):
 - transition: Claculates feedforwarding process as described in SRS Djavaherpour (2024b). This is done by using the predicted value of previous step, parameters and the number of layers, parameters dictionary and the number of layers.
 - output: Returns the new predicted value and a cache including new and old parameters.
 - exception: None.
- extract_parameters(cache):
 - transition: Extracts parameters saved during forwardfeeding from the cache, based on layers, parameters dictionary and the number of layers.

- output: Returns extracted parameters.
- exception: None.
- backpropagation(cache, predicted, actual, layers):
 - transition: Claculates backpropagation process as described in SRS Djavaherpour (2024b) to calculate gradients of wights and biases.
 - output: A dictionary as gradients that keys are labels of gradients to define weights or biases, and values are gradients.
 - exception: None.

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