

---

# **panna Documentation**

*Release prerelease*

**pannadevs**

**Nov 15, 2018**



## CONTENTS:

<b>1</b>	<b>Simulation - the data</b>	<b>3</b>
<b>2</b>	<b>Input of the network</b>	<b>5</b>
<b>3</b>	<b>Checkpoint: saving and restart routine</b>	<b>7</b>
<b>4</b>	<b>Input parser</b>	<b>9</b>
<b>5</b>	<b>Network architecture</b>	<b>11</b>
<b>6</b>	<b>Regularizations</b>	<b>13</b>
<b>7</b>	<b>Layers of the network</b>	<b>15</b>
<b>8</b>	<b>Real Layers of the network</b>	<b>17</b>
<b>9</b>	<b>TF Variable helpers</b>	<b>19</b>
<b>10</b>	<b>Release notes &amp; To-do</b>	<b>21</b>
	<b>Python Module Index</b>	<b>23</b>
	<b>Index</b>	<b>25</b>



Automatically created by Sphinx using the documentation in the source code



## SIMULATION - THE DATA

```
class simulation.Example(g_vectors,      species_vector,      true_energy,      zeros=None,  
                        atomic_species=None, name=None)
```

Example class

```
simulation.iterator_over_tfddata(g_size, *args, **kwargs)
```

TFdata unpacker

### Parameters

- **g\_size** – size of the G's
- **args** – all the tfdata files that one wants to parse

**kwargs:** zeros: list of zeros, one per species

**Retrun:** iterator over the record in the files





## INPUT OF THE NETWORK

Utilities to handling the input system

`inputs.parse_fn_v1` (*example*, *g\_size*, *zeros*, *n\_species*)  
Parse TFExample records and perform simple data augmentation.

**Parameters**

- **example** – a batch of example obj
- **g\_size** – size of the *g\_vector*
- **zeros** – array of zero's one value per specie.
- **n\_species** – number of species

**Returns** Sparse Tensor, (n\_atoms) value in range(n\_species) *g\_vectors\_tensor*: Sparse Tensor, (n\_atoms, *g\_size*) *energy*: true energy value corrected with the zeros

**Return type** *species\_tensor*

`inputs.input_iterator` (*data\_dir*, *batch\_size*, *parse\_fn*, *name*, *shuffle\_buffer\_size\_multiplier*=10,  
*prefetch\_buffer\_size\_multiplier*=20, *num\_parallel\_readers*=8,  
*num\_parallel\_calls*=8, *\*args*, *oneshot*=None)

Construct input iterator.

**Parameters**

- **data\_dir** – directory for data, must contain a “train\_tf subfolder”
- **batch\_size** – batch size
- **parse\_fn** – function to parse the data from tfrecord file
- **name** – name scope
- **\*\_buffer\_size\_multiplier** – batchsize times this number
- **num\_parallel\_readers** – process that are doing Input from drive
- **num\_parallel\_calls** – call of the parse function
- **oneshot** – experimental, do not set
- **TODO** – construct a double system to handle in\_place evaluation of accuracy

**Returns** *initializable\_iterator*, recover input data to feed the model

---

**Note:**

- shuffling batch and buffer size multiplier default are randomly chosen by me

- initializable iterator can be changed to one shot iterator in future version to better comply with documentation
  - a maximum number of epoch should also be added to this routine.
-

## CHECKPOINT: SAVING AND RESTART ROUTINE

**class** checkpoint.**Checkpoint** (*filename, atoms\_list=None*)  
class to handle a Checkpoint

**class** checkpoint.**Parameters** (*file\_name, atoms\_list=None*)  
A class that load parameters form files generated by and is compatible with the Network object



## INPUT PARSER

`parser_callable.get_network_architecture (value)`  
parse the architecture format

**Parameters**

- **value** – string like
- **layer\_size** – layer2\_size...

**Returns** list of size per layer

`parser_callable.get_network_trainable (value)`  
parse trainable list

**Parameters**

- **value** – string like
- **1 – 0:1**

**Returns** list of trainable flag per layer



## NETWORK ARCHITECTURE

```
networks.network_A2A(batch_of_species, batch_of_gvects, layer_size, trainability, gvect_size,
                      batch_size, Nspecies, atomic_label, import_layer=None, reuse=None)
```

**New network with variable architecture annd** species resolved weights.

### Parameters

- **batch\_of\_species** – [batch size x max number of atoms per molecule] species vector for each element of the batch
- **batch\_of\_gvects** – batch of gvectors
- **layer\_size** – a list of lists with the size of each hidden layer, for each species
- **trainability** – a list of lists with the boolean flag of the trainable status of each hidden layer, for each species
- **gvect\_size** – gvector size
- **batch\_size** – number of calculations in a batch
- **Nspecies** – number of species
- **import\_layer** – a list of lists of tuple, each tuple has 2 elements, weights and biases that can be either a tensor with correct shape or None
- **reuse** – whether to reuse variables with the same name

### Returns

tf.tensor of energies natoms\_batch: tf.tensor of number of atoms for each element of the batch

### Return type

Energy

```
networks.loss_NN(batch_energies, batch_energies_dft, batch_natoms)
```

this is simply our cost function

### Parameters

- = **prediction of the network**(*batch\_energies*) –
- = **energies labels**(*batch\_energies\_dft*) –
- = **number of atoms**(*batch\_natoms*) –

**Returns** the loss value tensor with delta\_e for each element of the batch

```
class networks.eval_network_A2A(checkpoint)
```

A2A network implementation





## REGULARIZATIONS

`regularizations.1112_regularizations` (*wscale\_l1*, *wscale\_l2*, *bscale\_l1*, *bscale\_l2*)

Apply required regularizator.

**Parameters**

- **w** – weights, **b** : biases
- **11** – norm one prefactor if zero nothing gets applied
- **12** – norm two prefactor if zero nothing gets applied

**Returns** A scalar representing the overall regularization penalty for W A scalar representing the overall regularization penalty for B



## LAYERS OF THE NETWORK

`train_ops.train_NN` (*loss, global\_step, lr, atomic\_sequence*)

Train NN model, optimization step.

Create an optimizer and apply to all trainable variables. Add moving average for all trainable variables.

### Parameters

- **loss** – quantity to minimize
- **global\_step** – Integer Variable counting the number of training steps processed.
- **lr** – learning rate
- **atomic\_sequence** – just for now here to simplify creation of histogram. . .

**Returns** op for training.

**Return type** train\_op

TODO: refactor this routine. . . . it is too big and does too many stuff



## REAL LAYERS OF THE NETWORK

`layers.hidden_layer_gauss` (*in\_tensor*, *in\_size*, *out\_size*, *trainable*, *init\_values*=(None, None))

**Define an all to all connected layer with species division and** Gaussian activation function.

### Parameters

- **in\_tensor** – input to be computed,
- **in\_size** – last dimension of the input,
- **out\_size** – last dimension of the output,
- **trainable** – whether we should train these weights
- **init\_values** – numpy arrays to initialize the tensors, weights and biases None = default initialization

**Returns** Output of the layer

weights variable will be named “weights” bias variable will be named “bias”

`layers.hidden_layer_linear` (*in\_tensor*, *in\_size*, *out\_size*, *trainable*, *init\_values*=(None, None))

**Define an all to all connected layer with species division and** linear activation function. TODO: Make a single layer that accepts activation function

### Parameters

- **in\_tensor** – input to be computed,
- **in\_size** – last dimension of the input,
- **out\_size** – last dimension of the output,
- **trainable** – whether we should train these weights
- **init\_values** – numpy arrays to initialize the tensors, weights and biases None = default initialization

**Returns** Output of the layer

weights variable will be named “weights” bias variable will be named “bias”



## **TF VARIABLE HELPERS**

These are specifically for TF variables .. automodule:: variable\_helpers .. autofunction:: \_variable\_on\_cpu .. autofunction:: \_variable\_random\_uniform





## RELEASE NOTES & TO-DO

Here can be release notes change log and to do

---

**Todo:** Update the below link when we add new guides on these.

---

- [genindex](#)
- [modindex](#)
- [search](#)



## PYTHON MODULE INDEX

### **c**

checkpoint, [7](#)

### **i**

inputs, [5](#)

### **l**

layers, [17](#)

### **n**

networks, [11](#)

### **p**

parser\_callable, [9](#)

### **r**

regularizations, [13](#)

### **s**

simulation, [3](#)

### **t**

train\_ops, [15](#)



## INDEX

### C

Checkpoint (*class in checkpoint*), 7  
checkpoint (*module*), 7

### E

eval\_network\_A2A (*class in networks*), 11  
Example (*class in simulation*), 3

### G

get\_network\_architecture() (in module  
parser\_callable), 9  
get\_network\_trainable() (in module  
parser\_callable), 9

### H

hidden\_layer\_gauss() (in module layers), 17  
hidden\_layer\_linear() (in module layers), 17

### I

input\_iterator() (in module inputs), 5  
inputs (*module*), 5  
iterator\_over\_tfdata() (in module simulation),  
3

### L

l1l2\_regularizations() (in module regulariza-  
tions), 13  
layers (*module*), 17  
loss\_NN() (in module networks), 11

### N

network\_A2A() (in module networks), 11  
networks (*module*), 11

### P

Parameters (*class in checkpoint*), 7  
parse\_fn\_v1() (in module inputs), 5  
parser\_callable (*module*), 9

### R

regularizations (*module*), 13

### S

simulation (*module*), 3

### T

train\_NN() (in module train\_ops), 15  
train\_ops (*module*), 15