

Summary of tracking hands-on exercise

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Setup

- exercise built on existing jupyter notebooks
- launched from caltech cluster, no user account creation needed
- authentication via github
- several notebooks with different approaches to same problem

Feedback on structure

- group had wide range of experience with relevant tools (jupyter, python, keras, tracking, etc)
- some CMS DAS-style asynchronous “pre-exercises” might be useful to add
- insight into NN structures came through conversations with leaders
- more time would be needed to do something substantive beyond running the existing notebooks

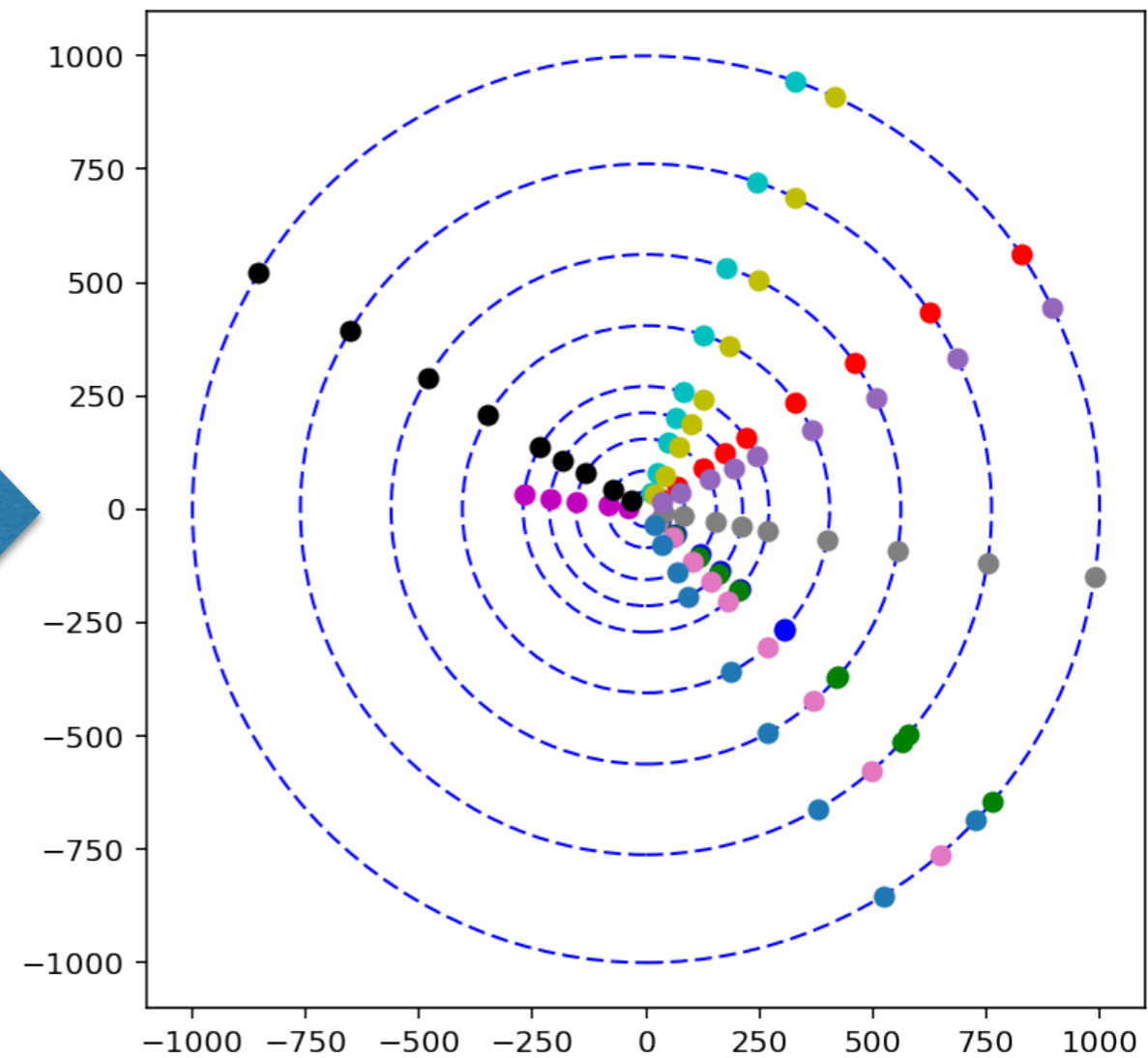
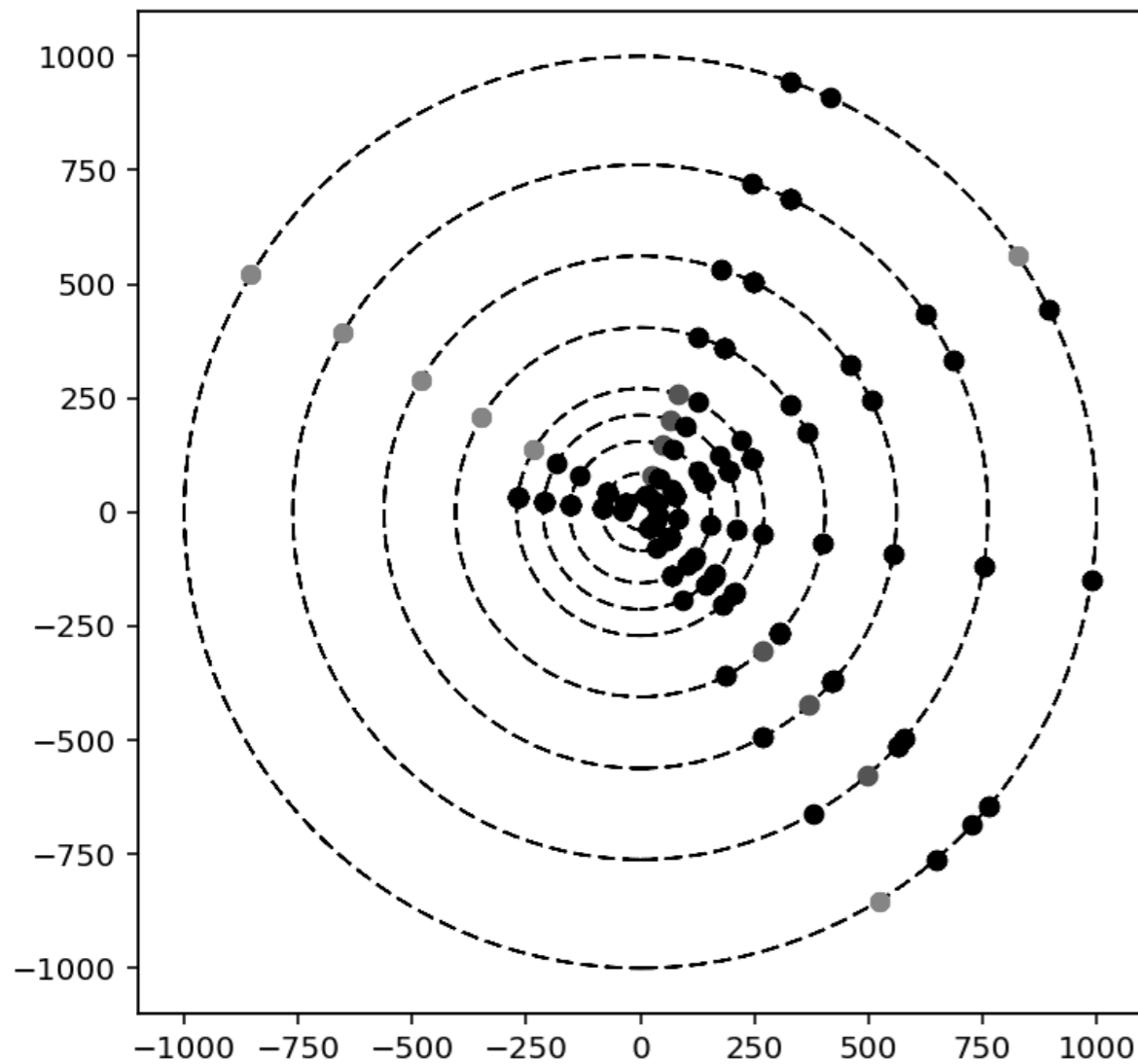
lots of useful
conversations

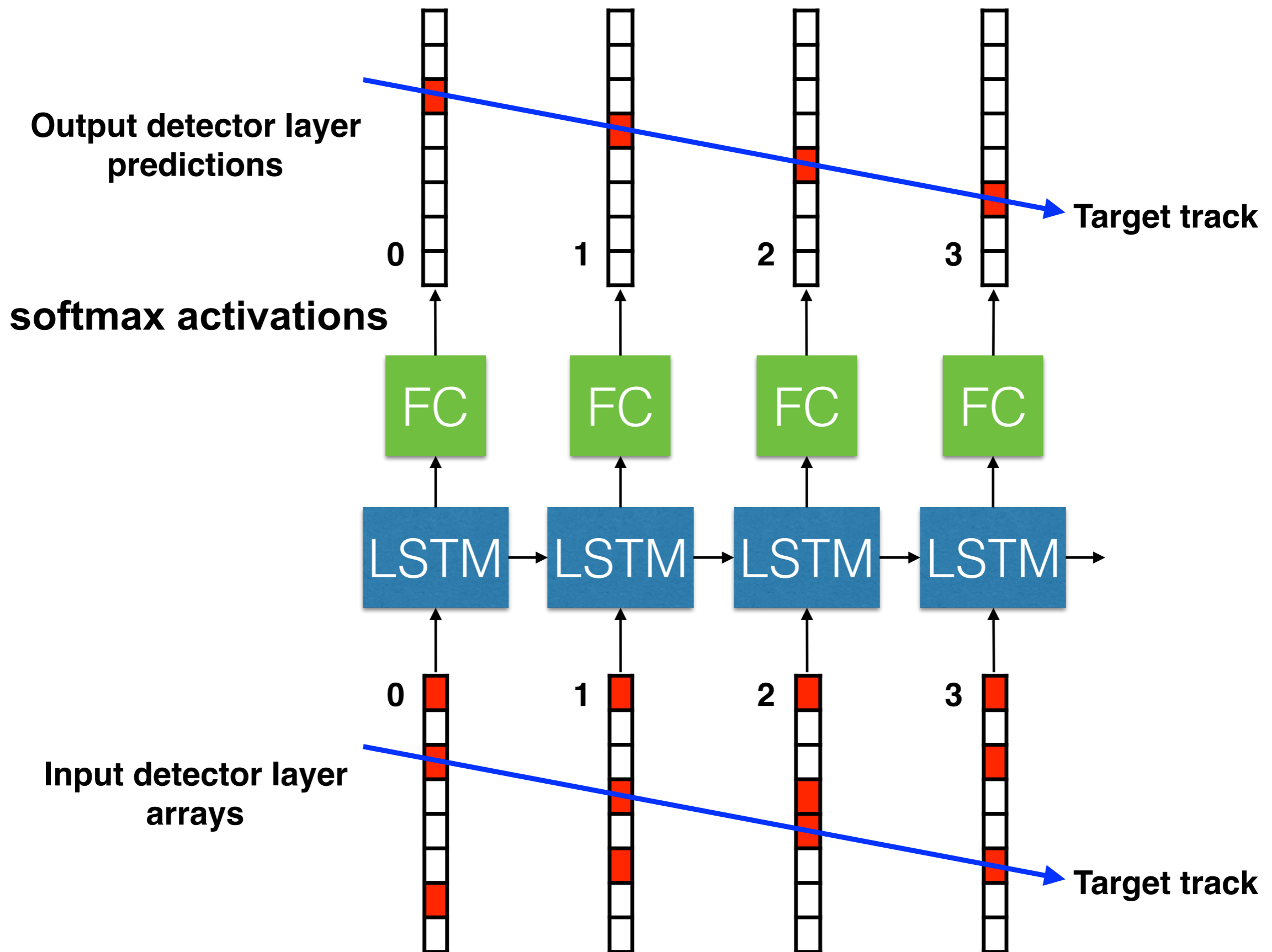


a boardful of useful conversations



problem: assign hits to tracks

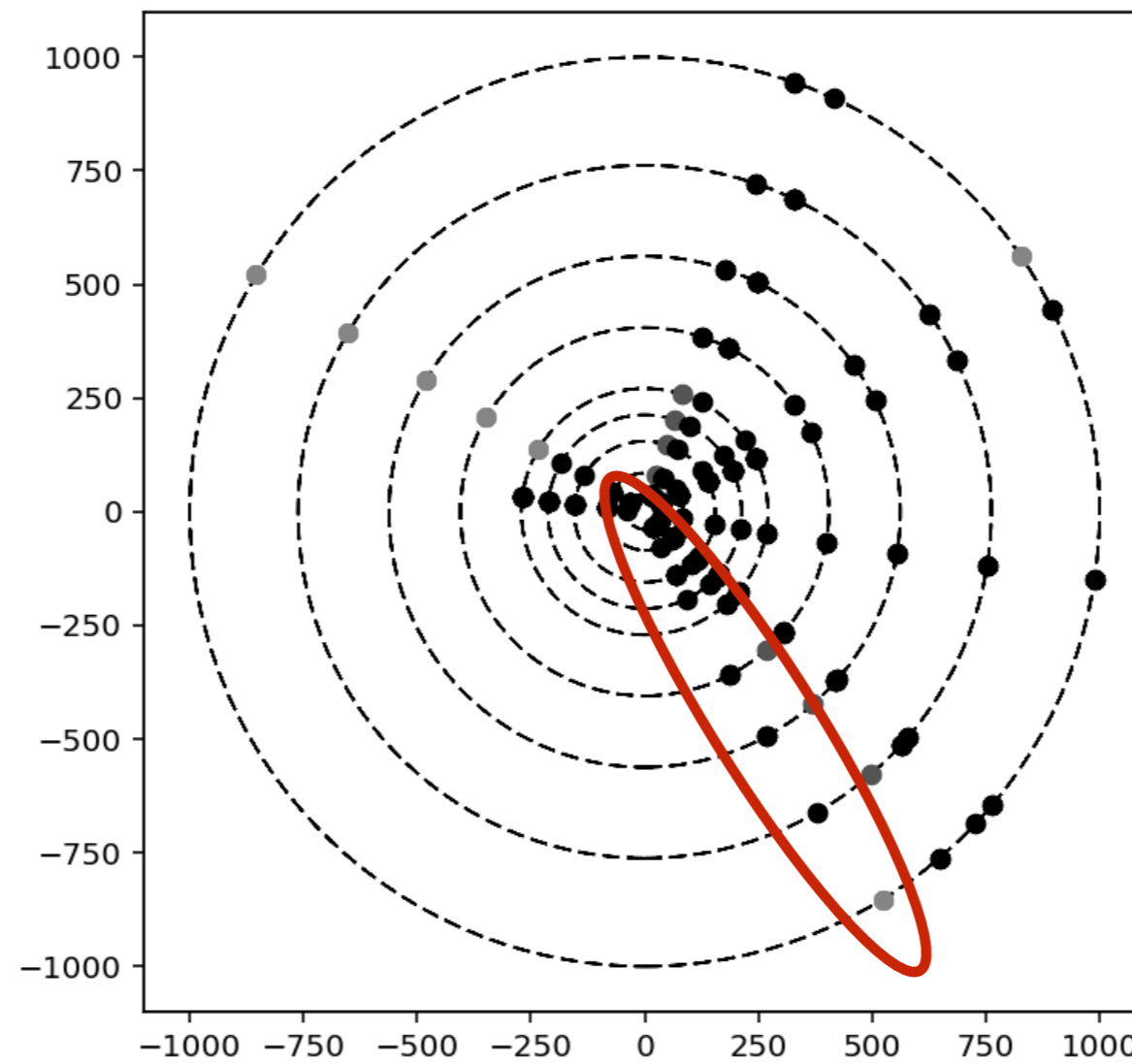
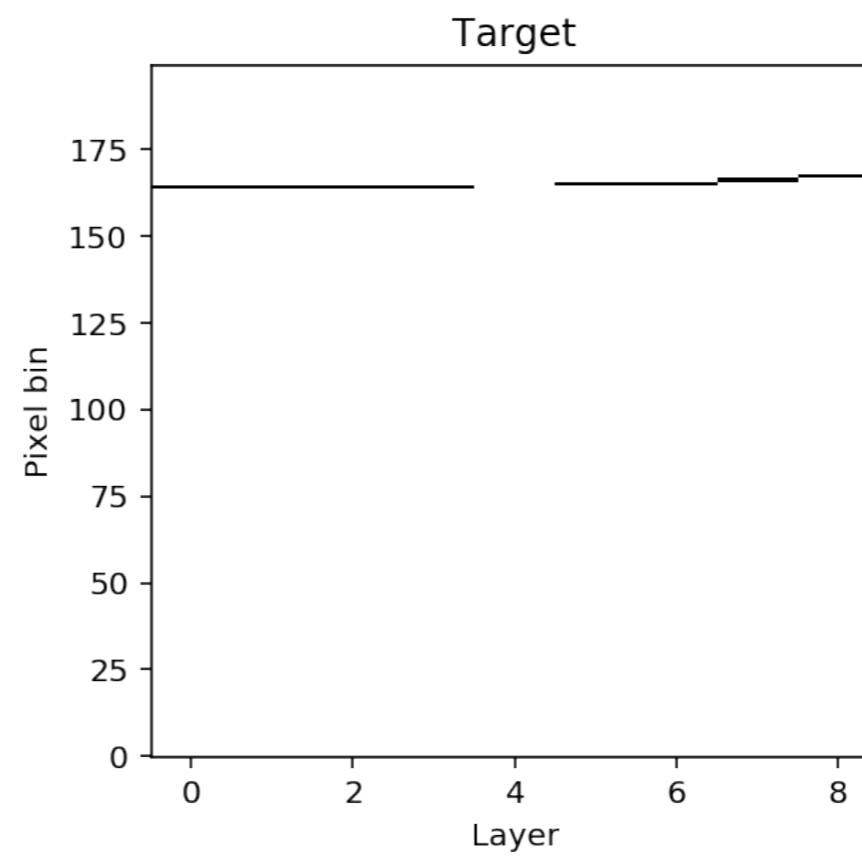
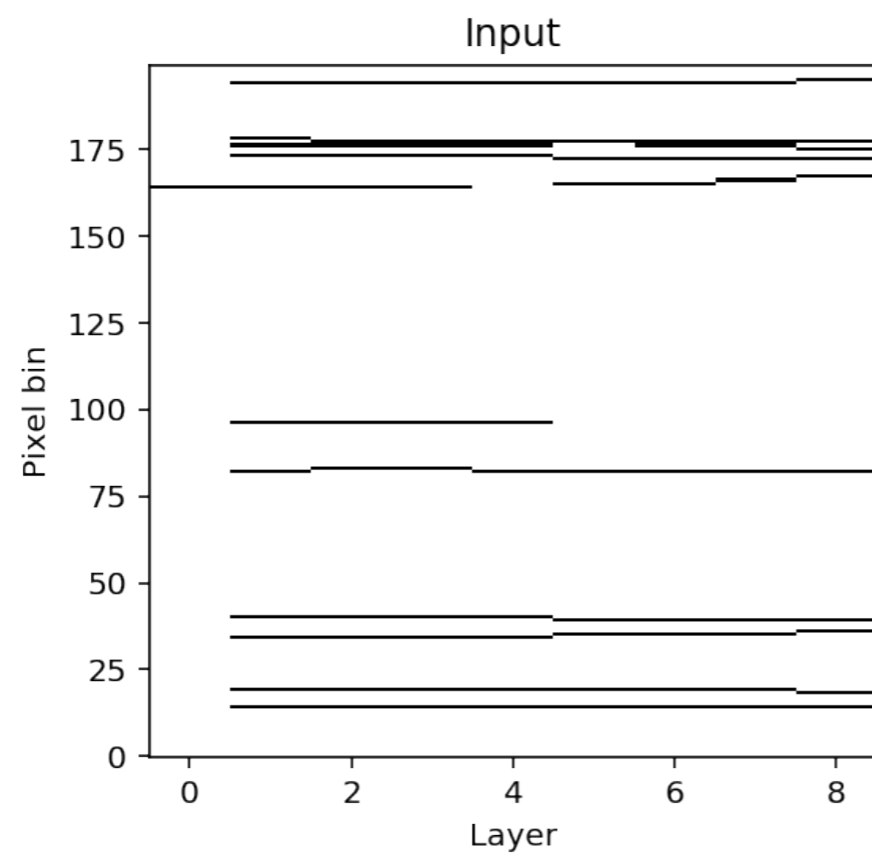


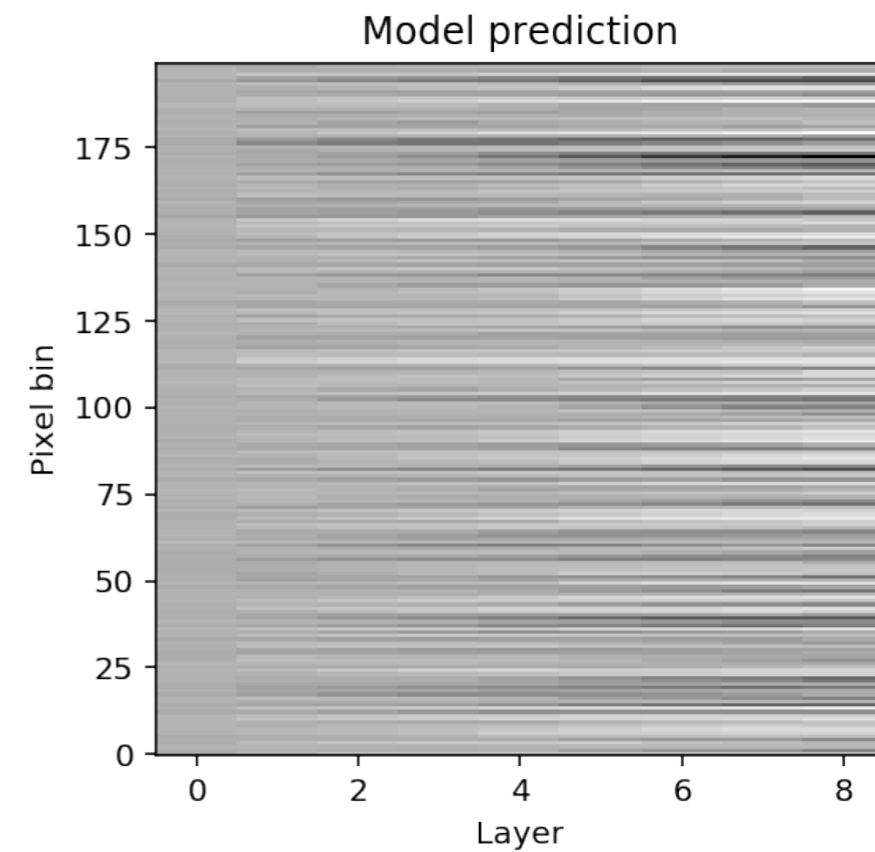
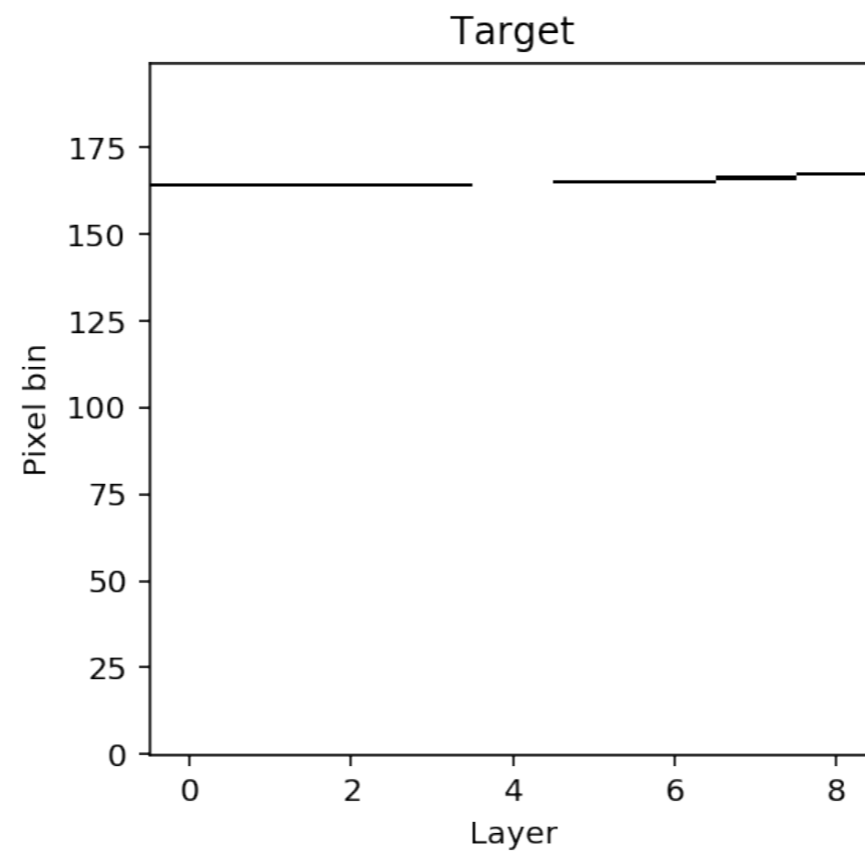
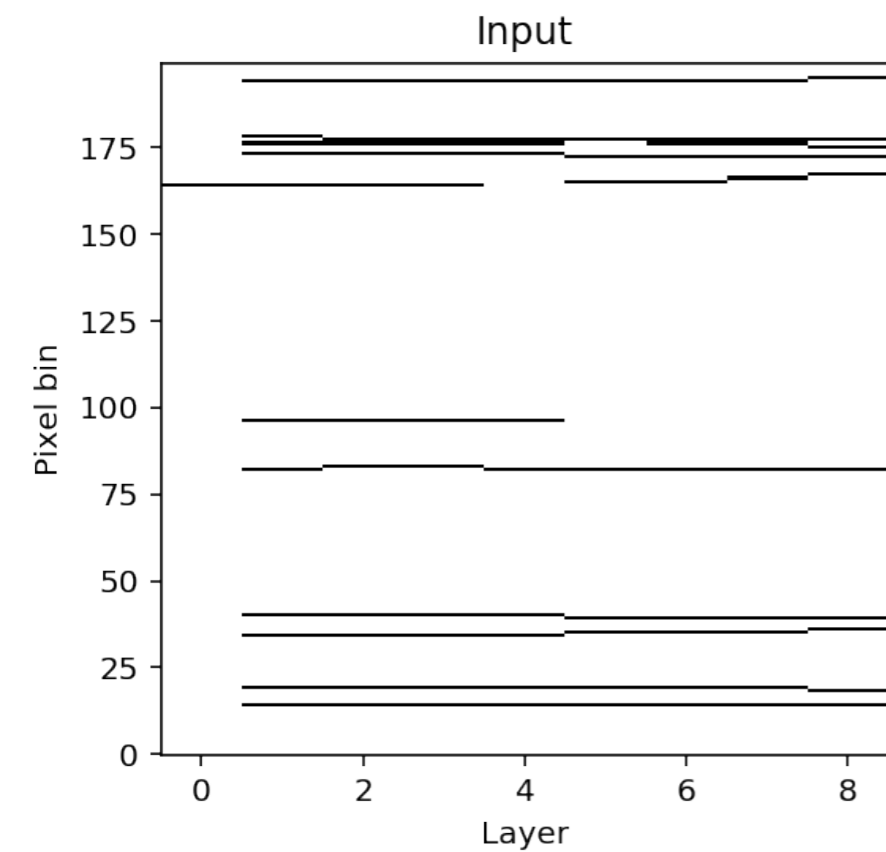


Where the magic happens

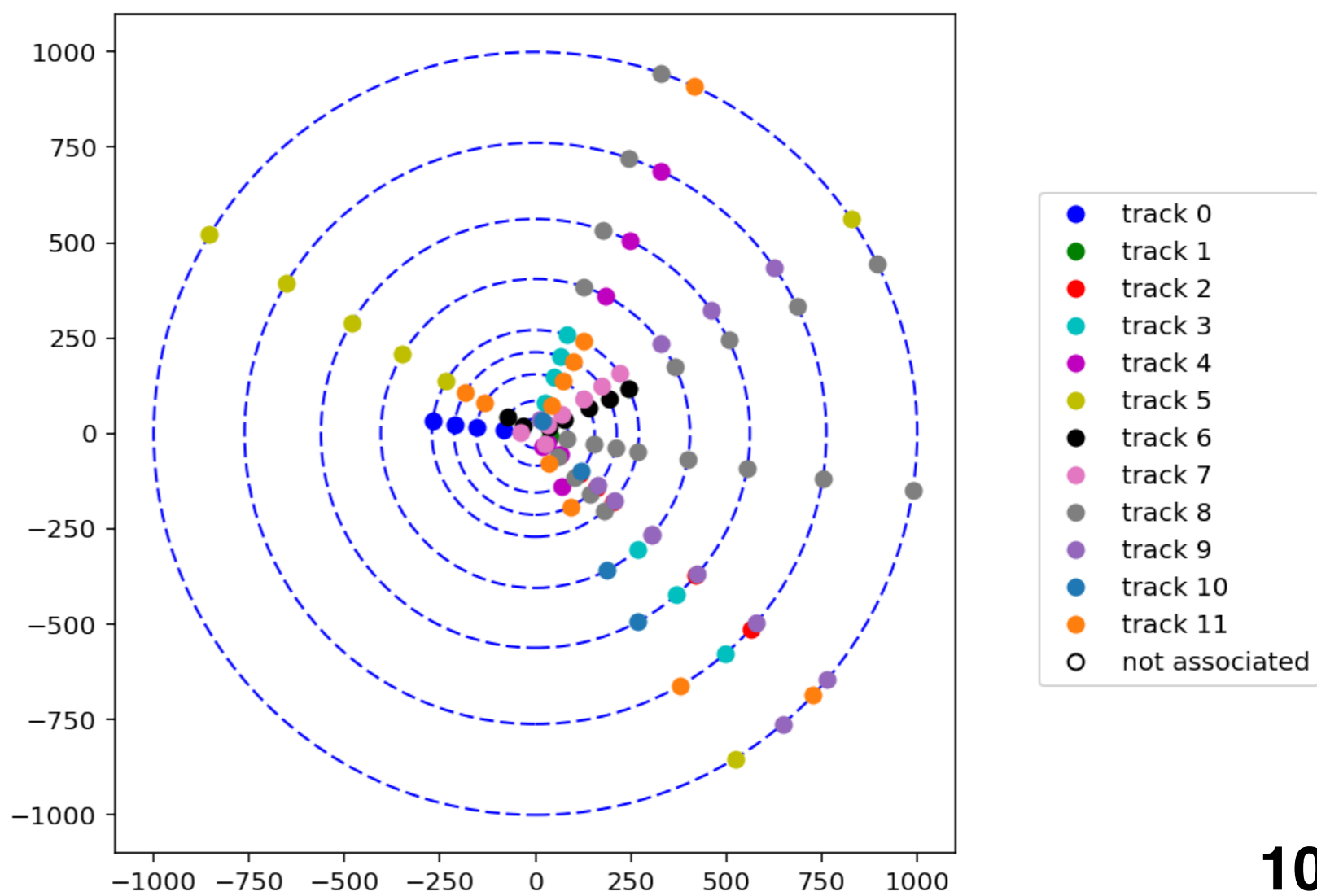
```
def build_model(num_hidden, length, dim,
                loss='categorical_crossentropy',
                optimizer='Nadam', metrics=['accuracy']):
    inputs = layers.Input(shape=(length, dim))
    hidden = layers.LSTM(output_dim=num_hidden, return_sequences=True)(inputs)
    outputs = layers.TimeDistributed(layers.Dense(dim, activation='softmax'))(hidden)
    model = models.Model(input=inputs, output=outputs)
    model.compile(loss=loss, optimizer=optimizer, metrics=metrics)
    return model
```

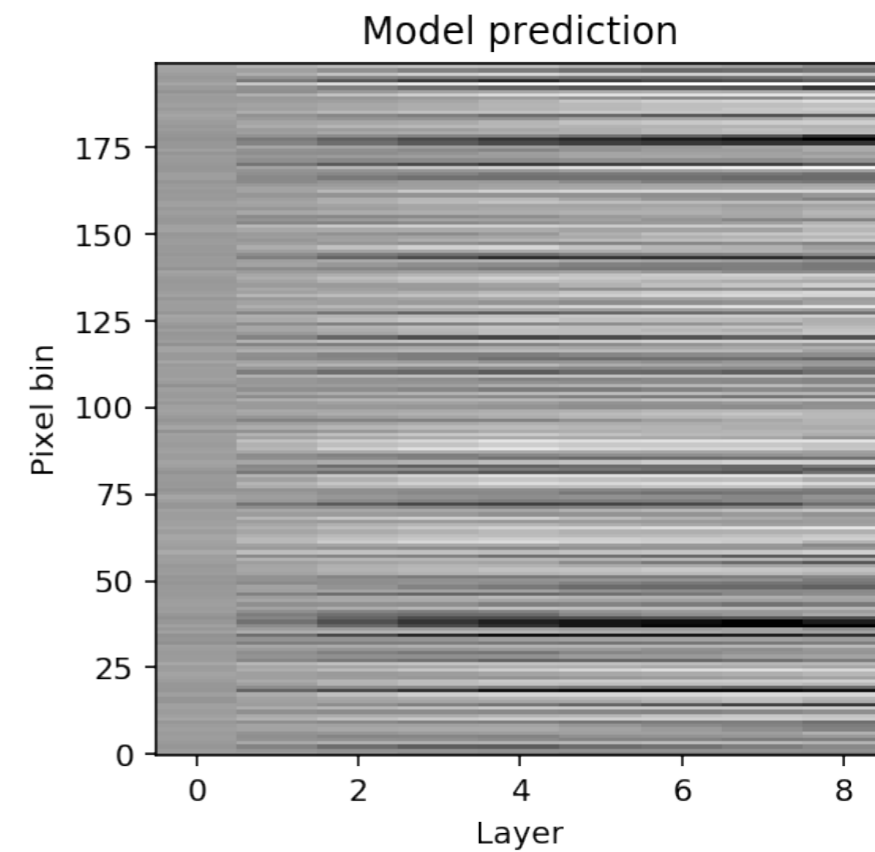
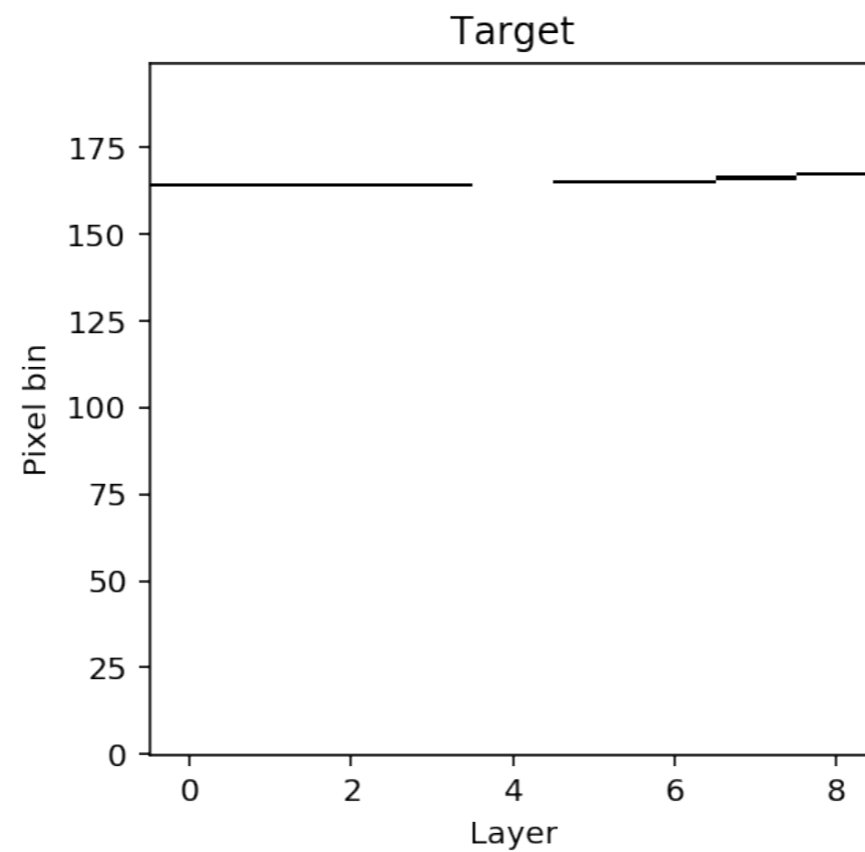
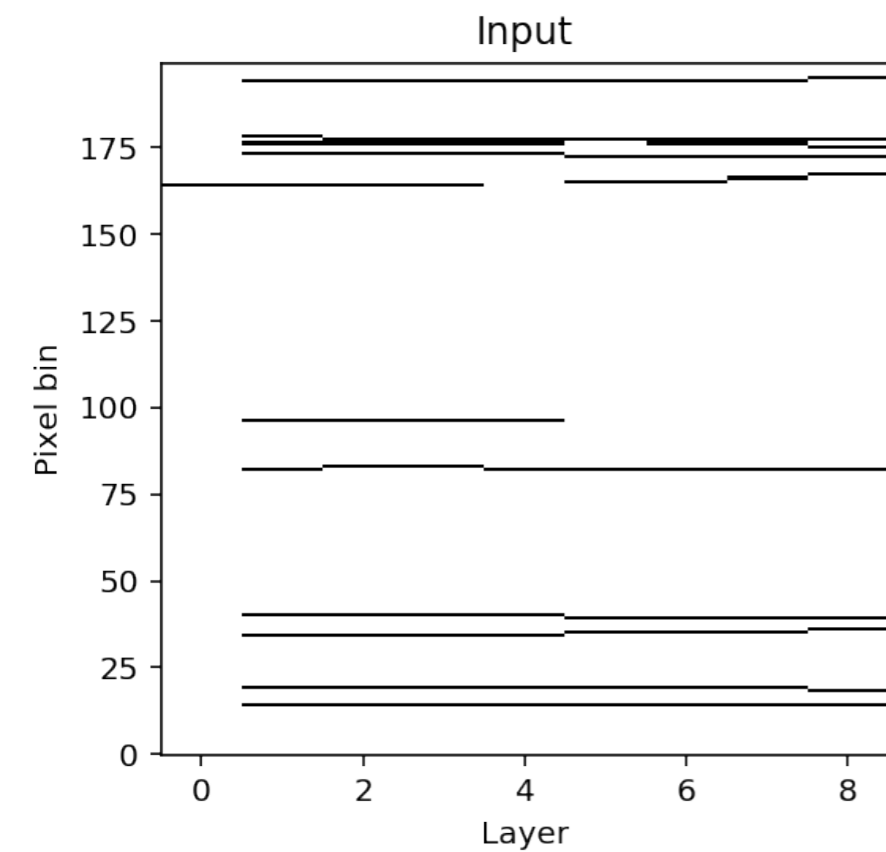
```
num_phi_bin=200
hidden_dim=50
batch_size=128
num_epoch=5
tracker = Clusterer(num_phi_bin=num_phi_bin, hidden_dim=hidden_dim,
                    batch_size=batch_size, num_epoch=num_epoch)
```



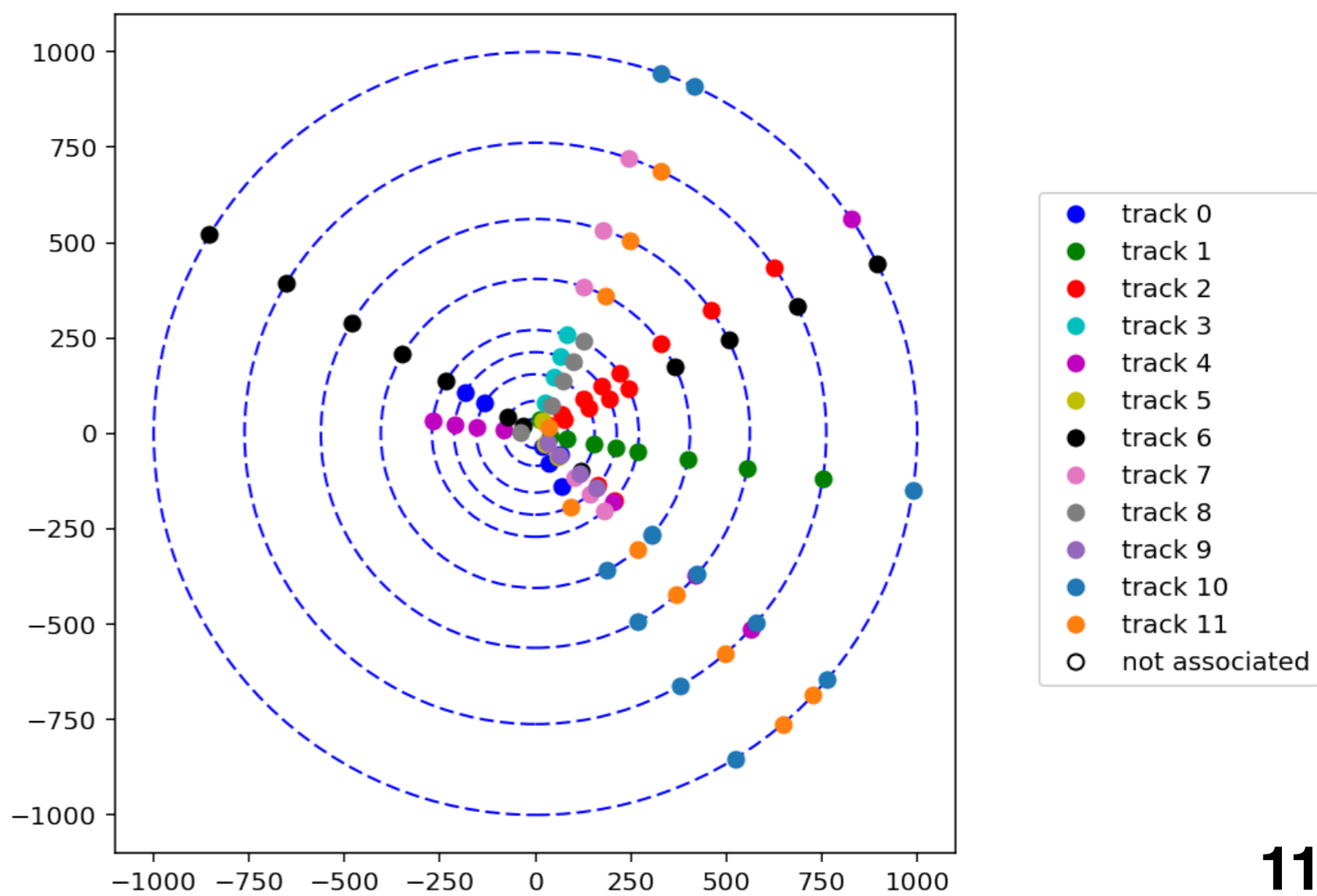


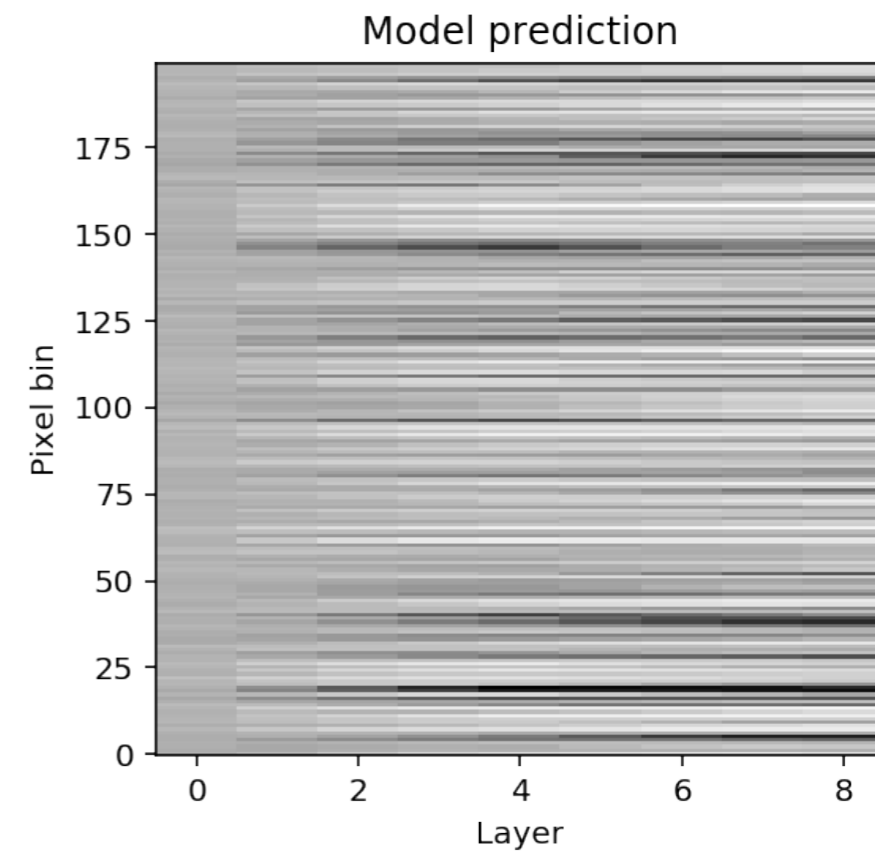
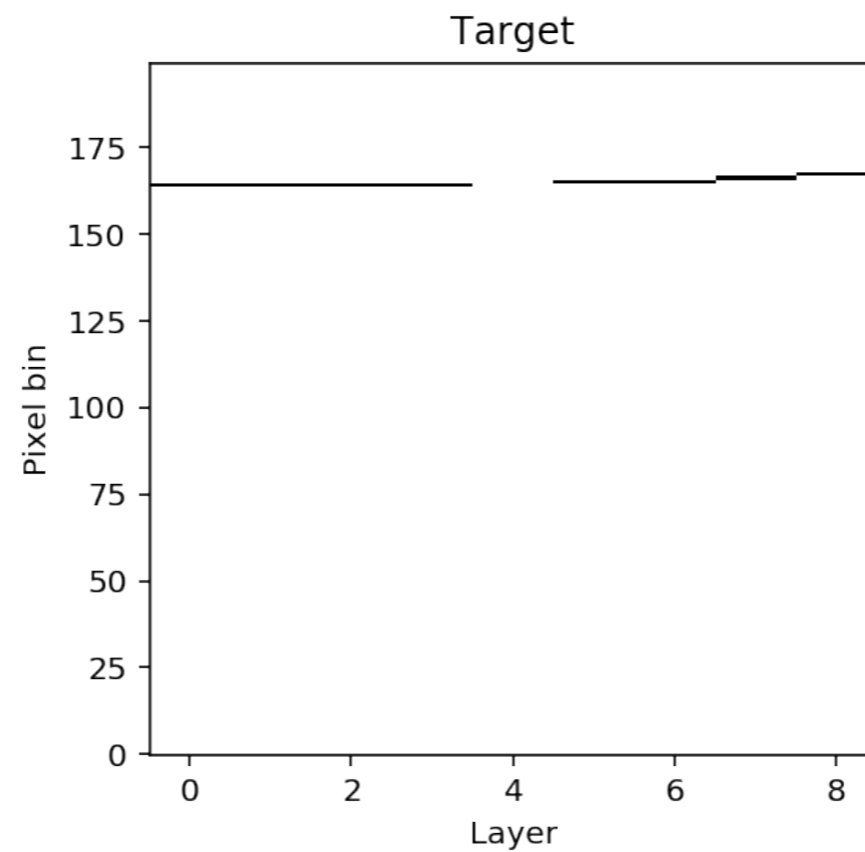
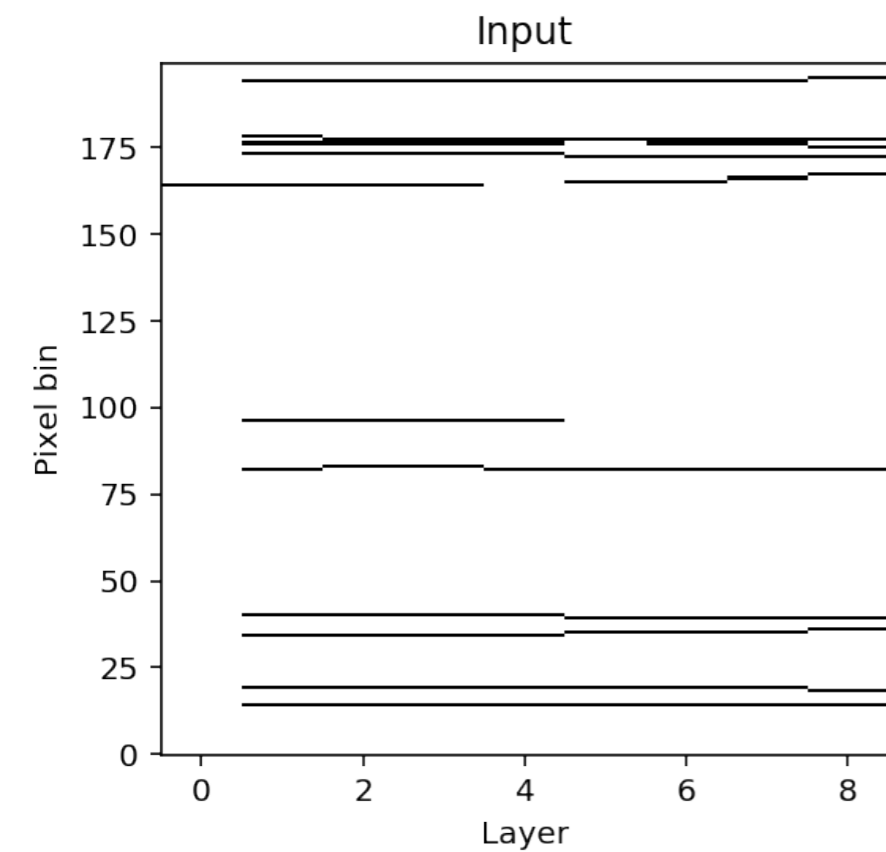
training set:
10 events



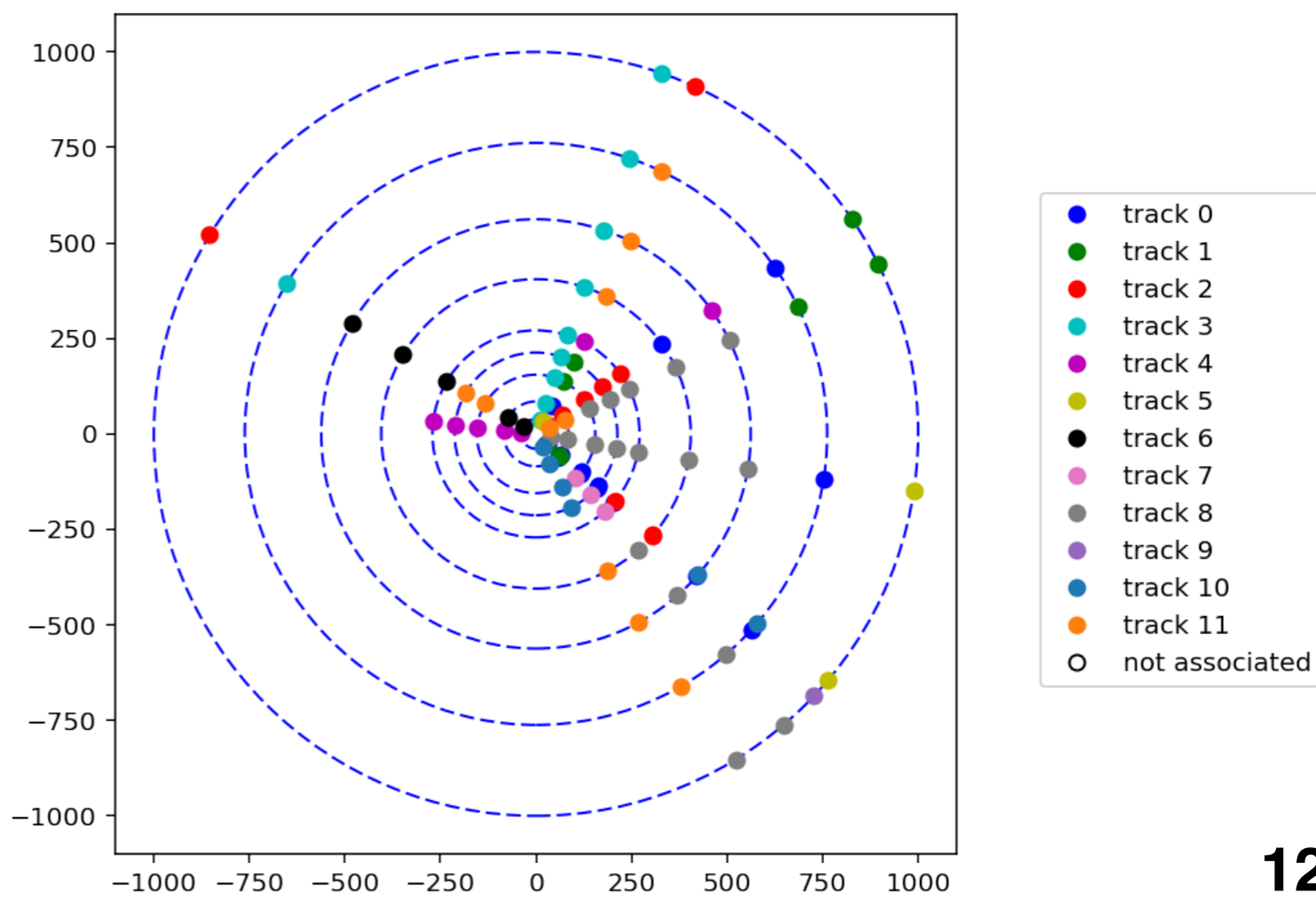


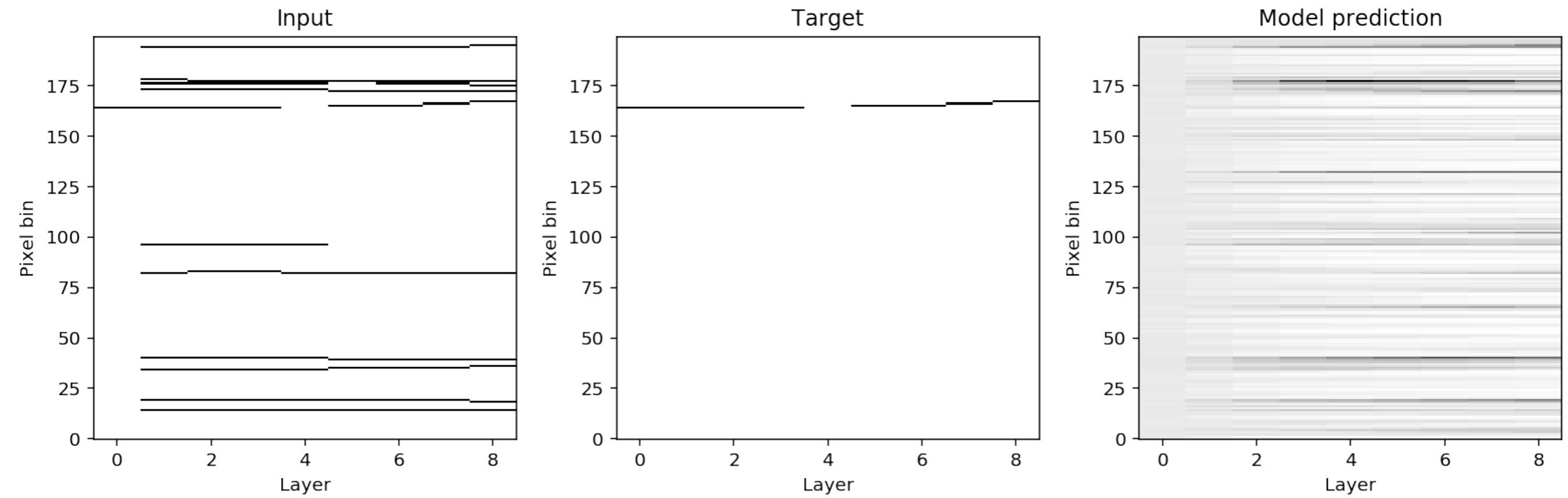
training set:
50 events



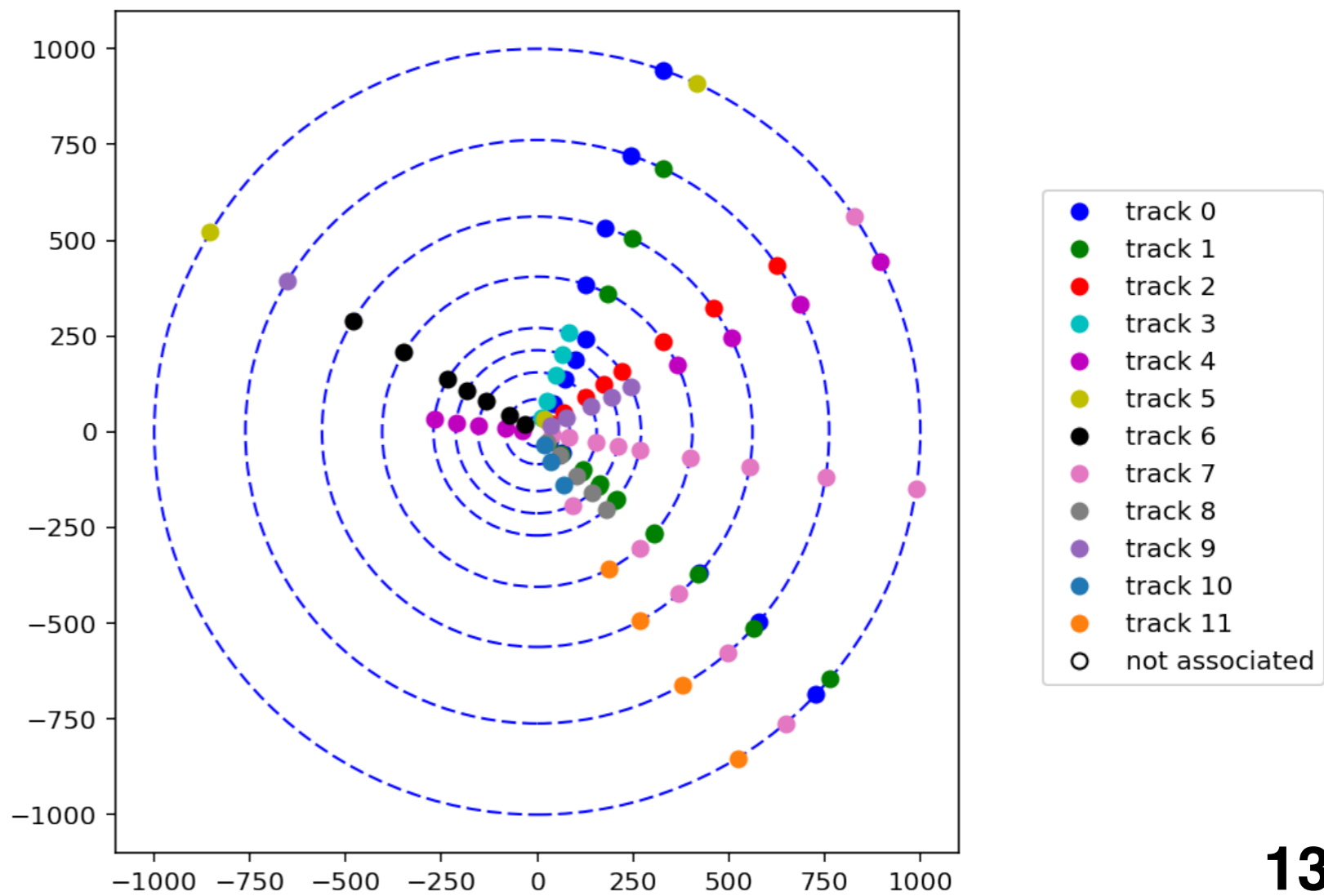


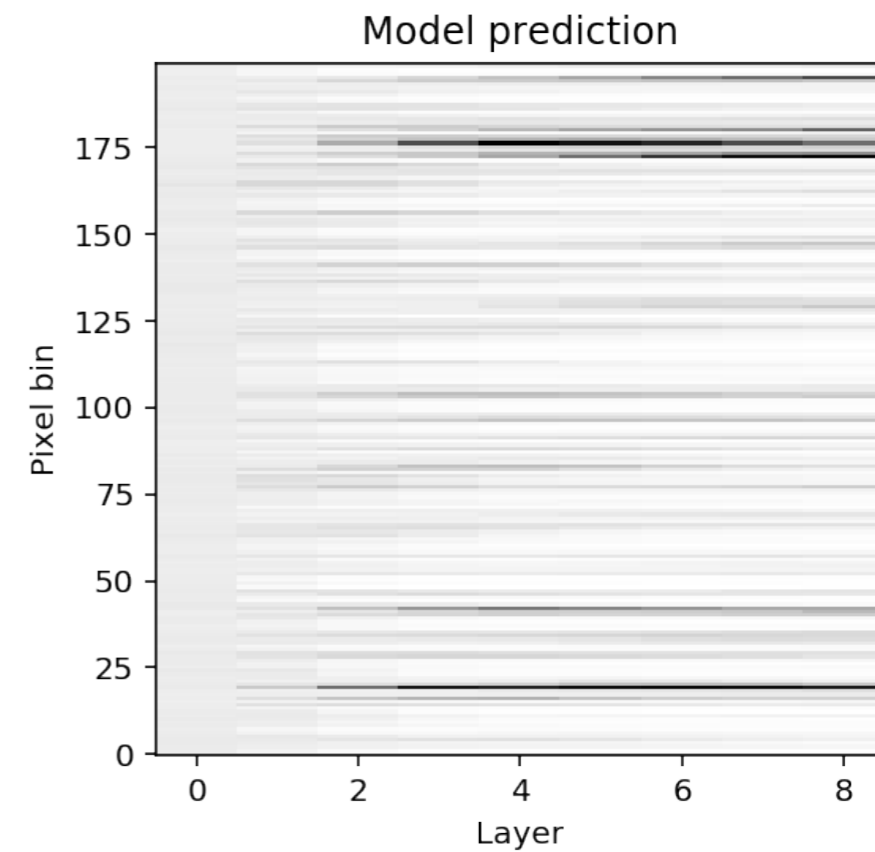
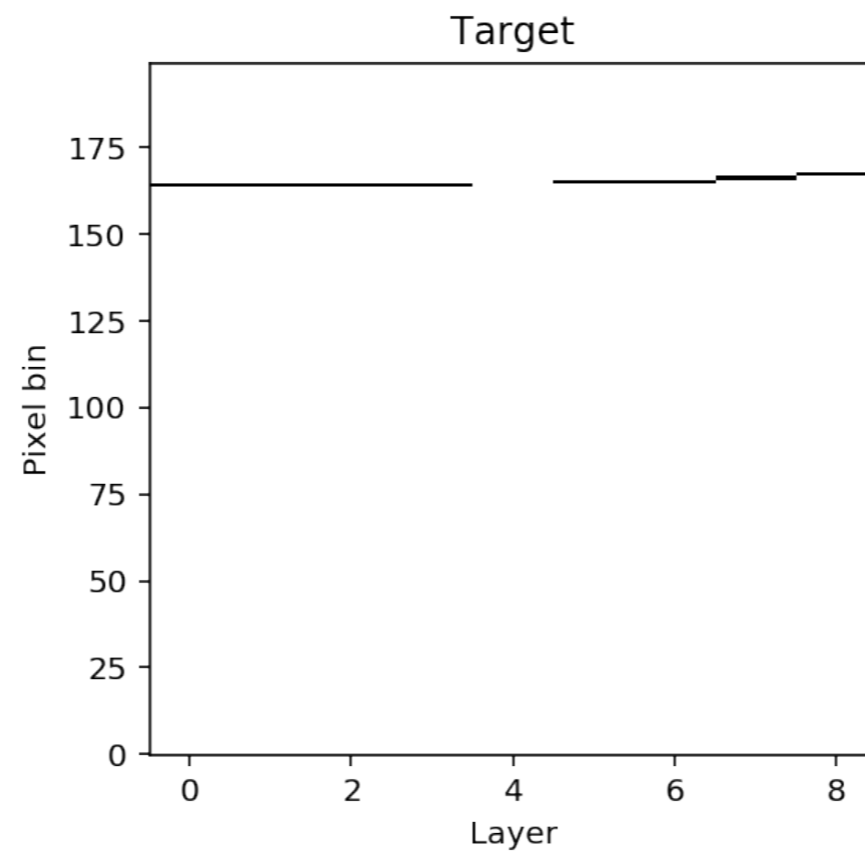
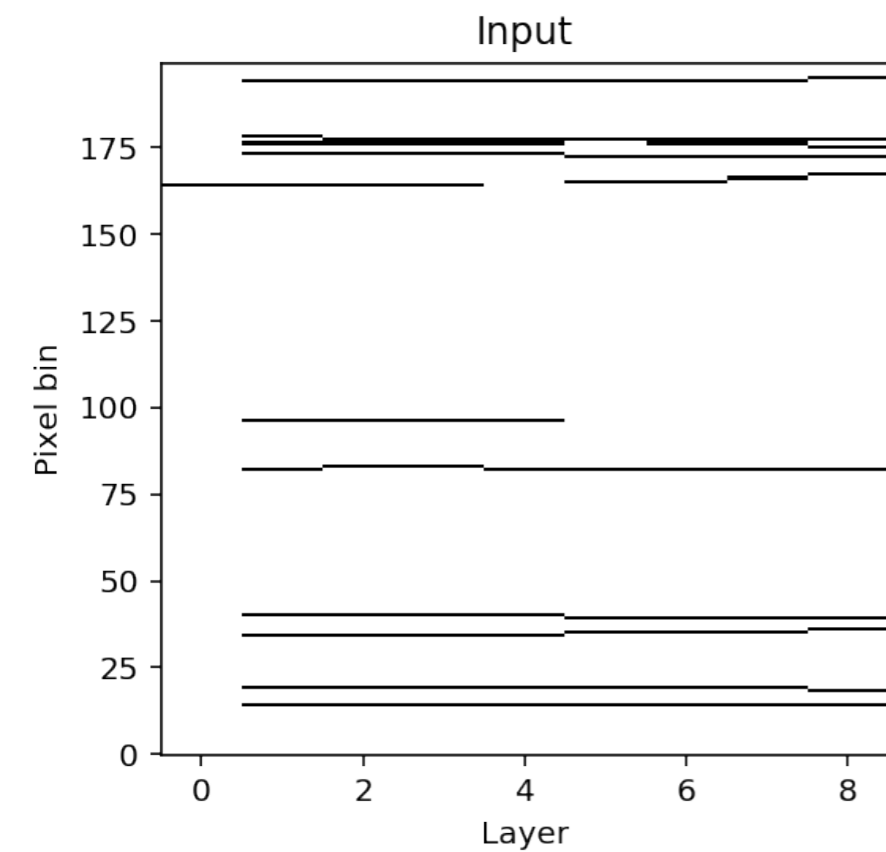
training set:
100 events



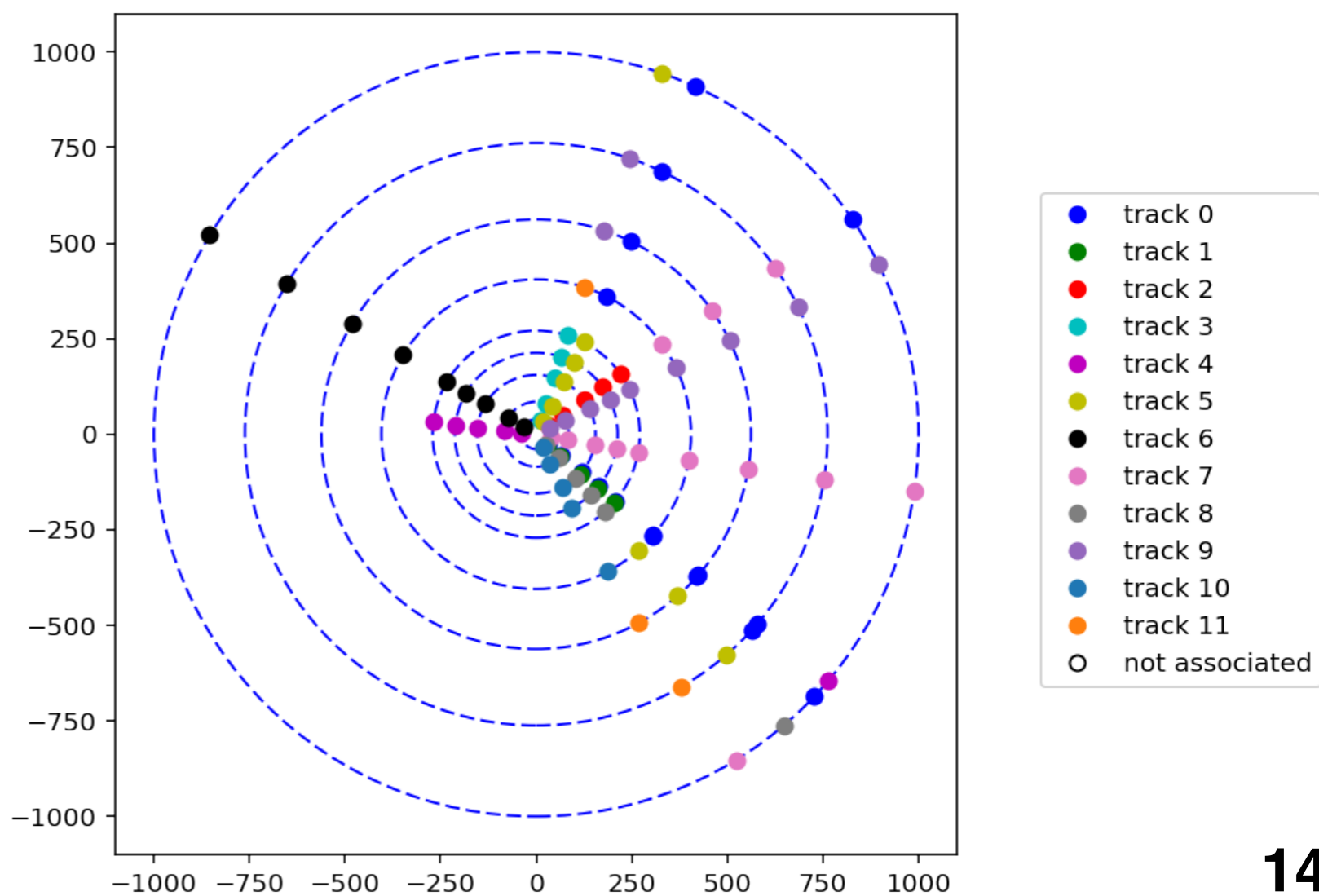


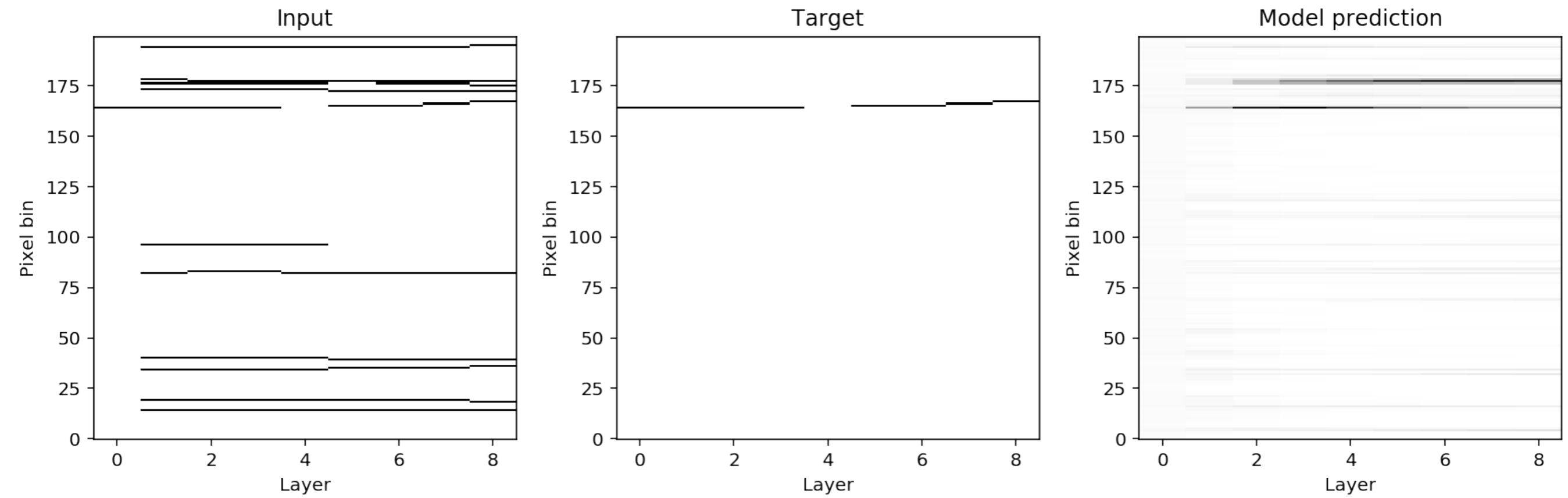
training set:
250 events



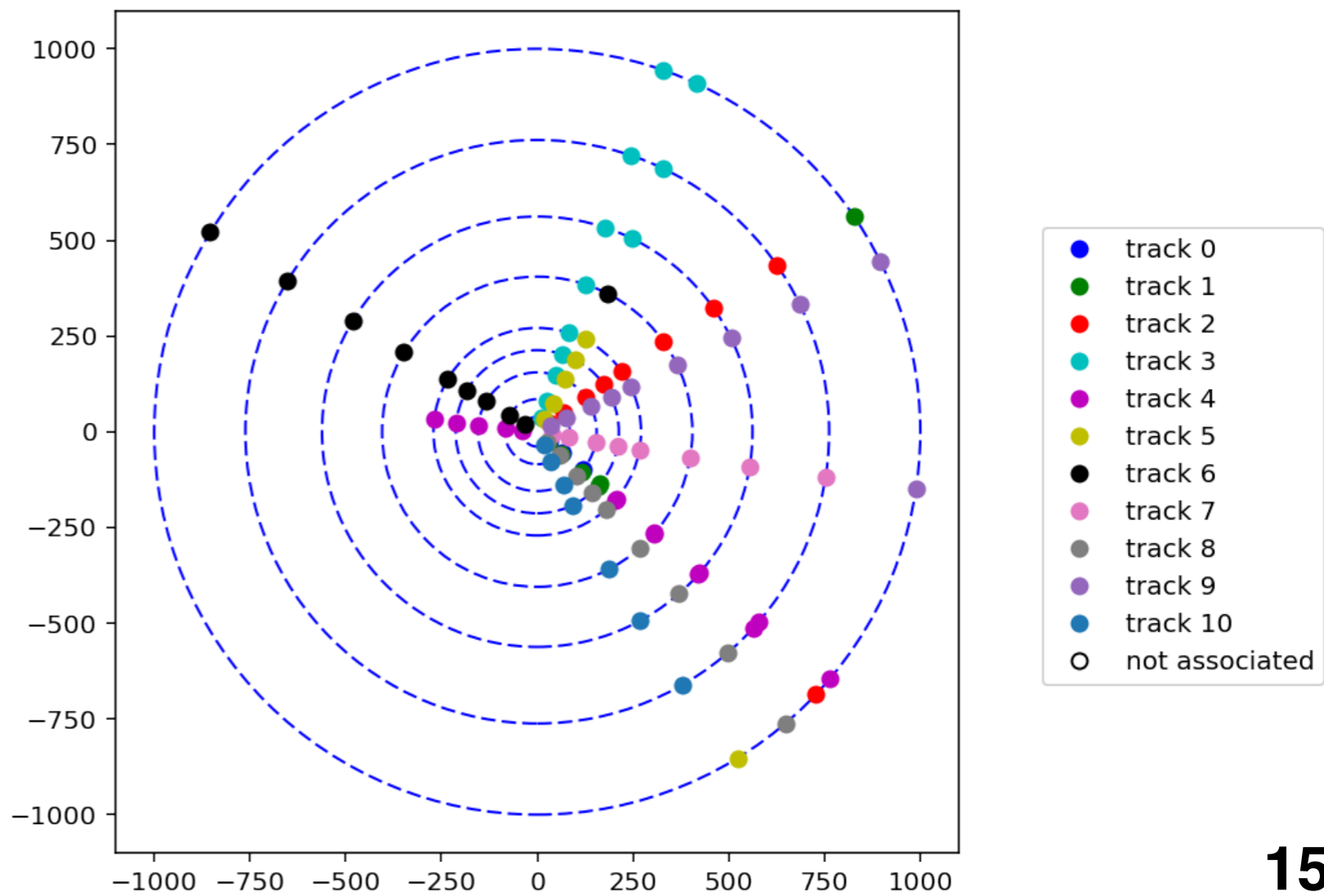


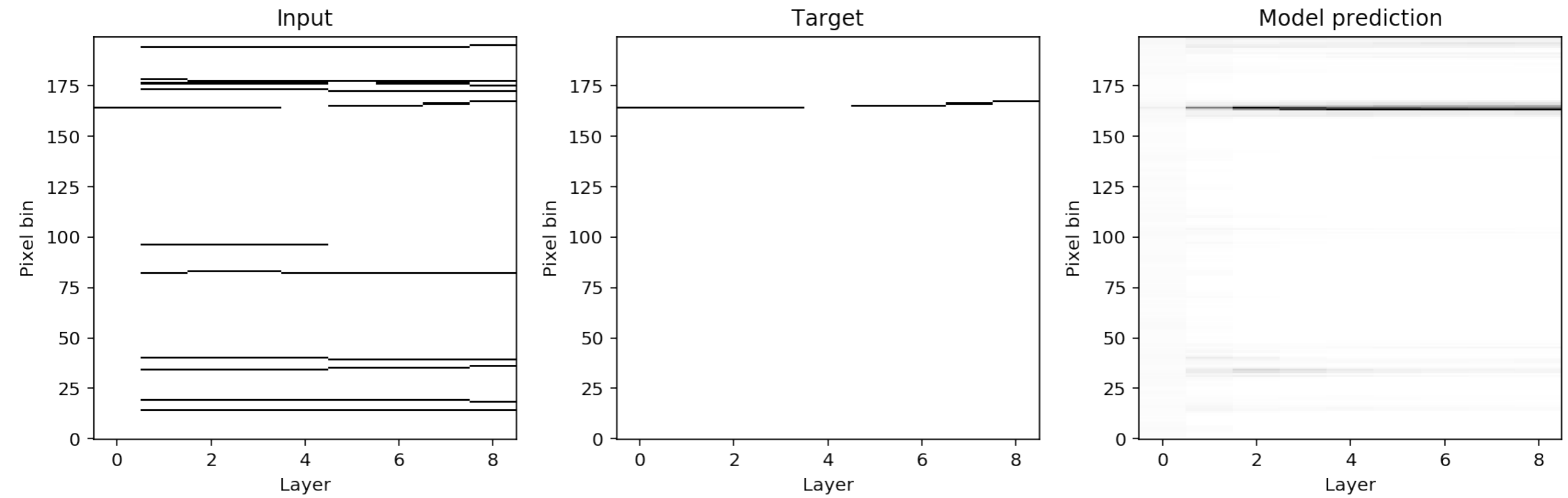
training set:
500 events



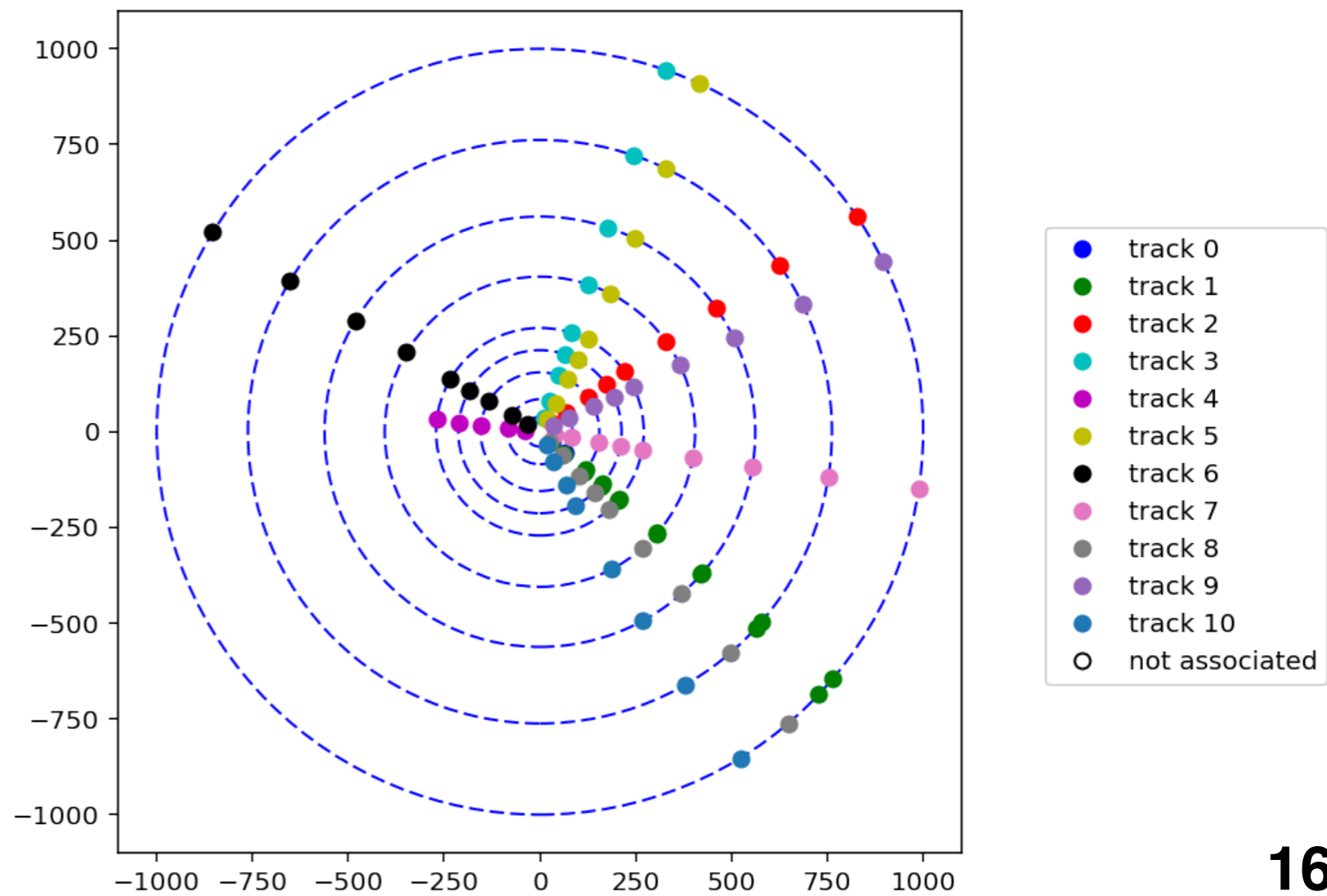


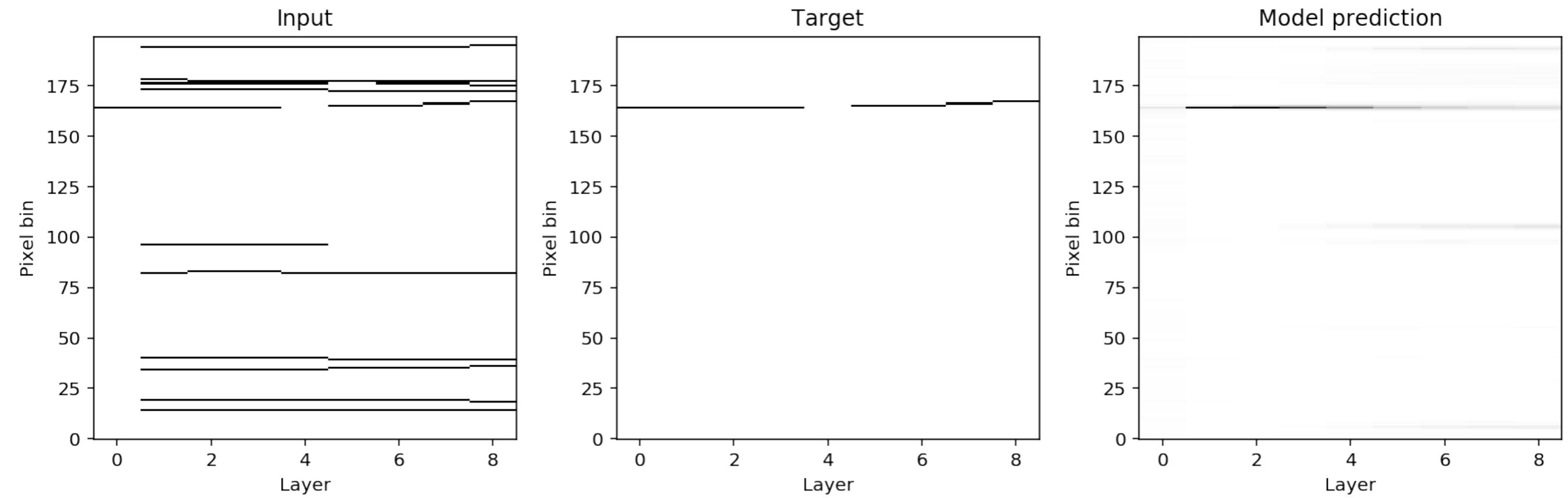
training set:
1000 events



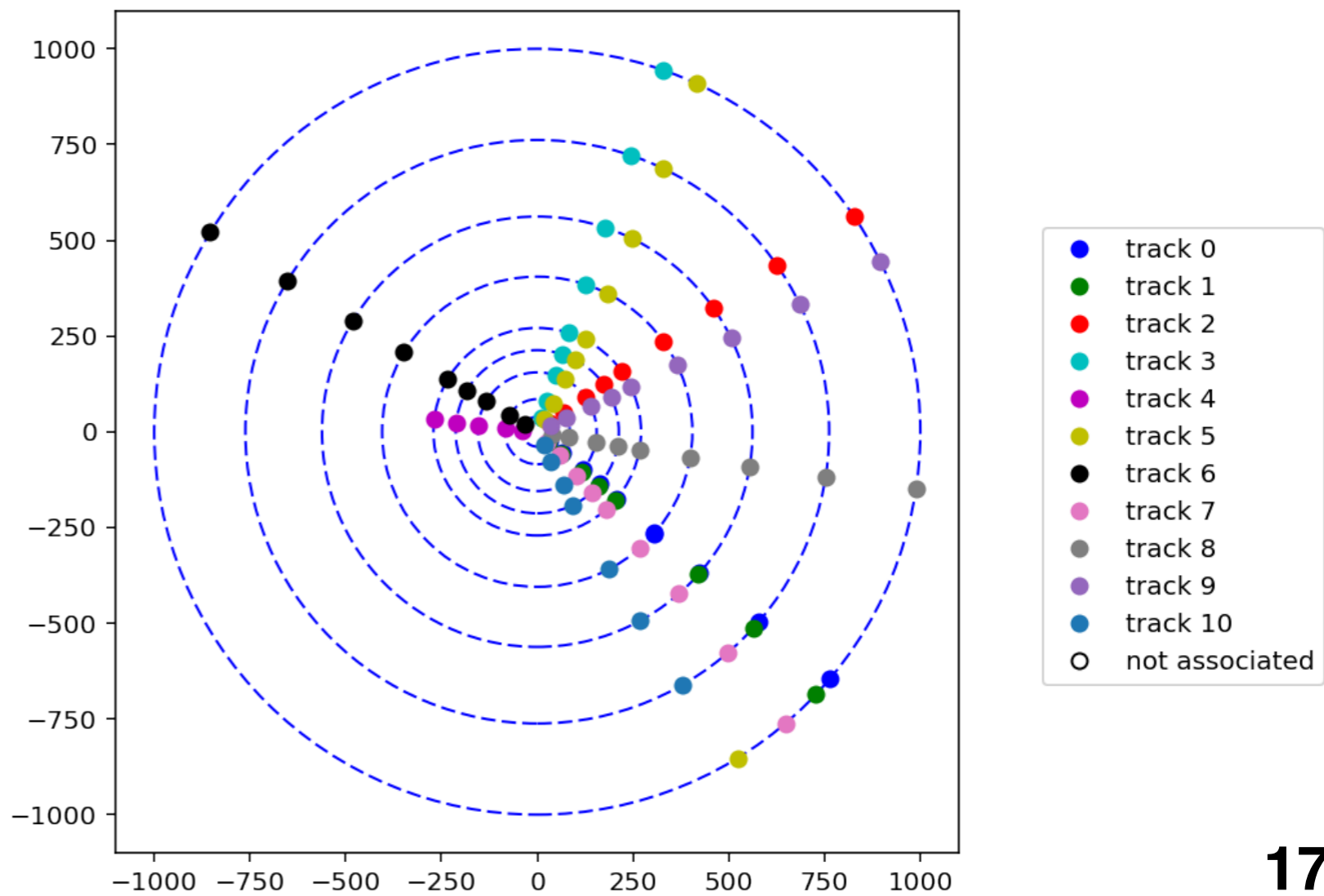


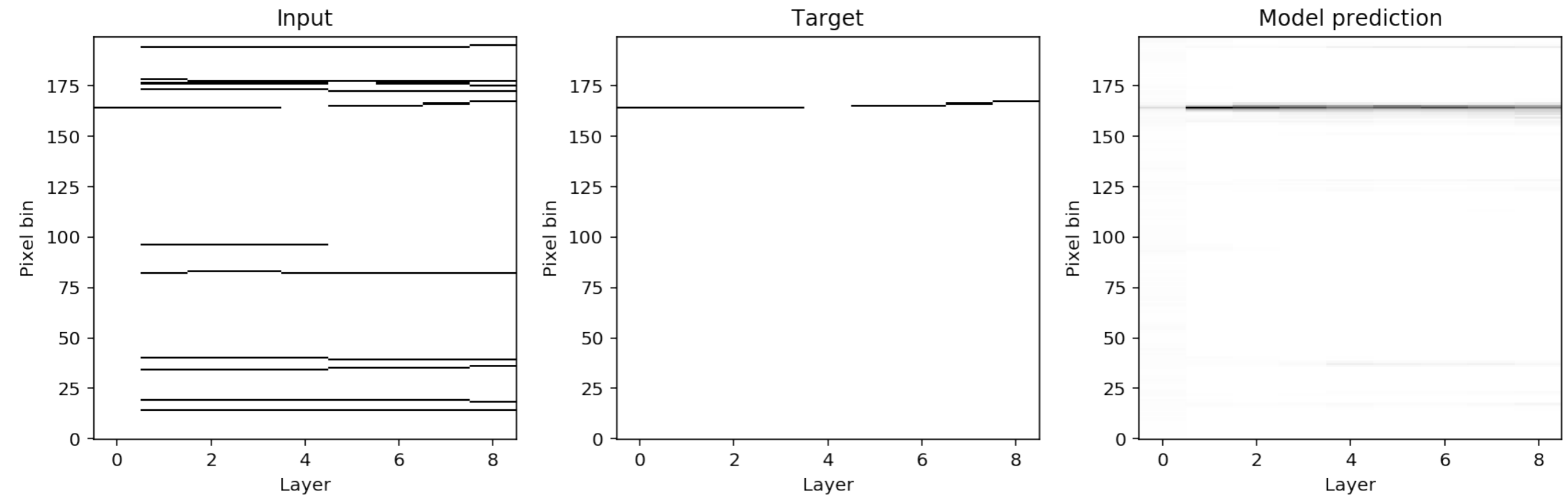
training set:
2000 events



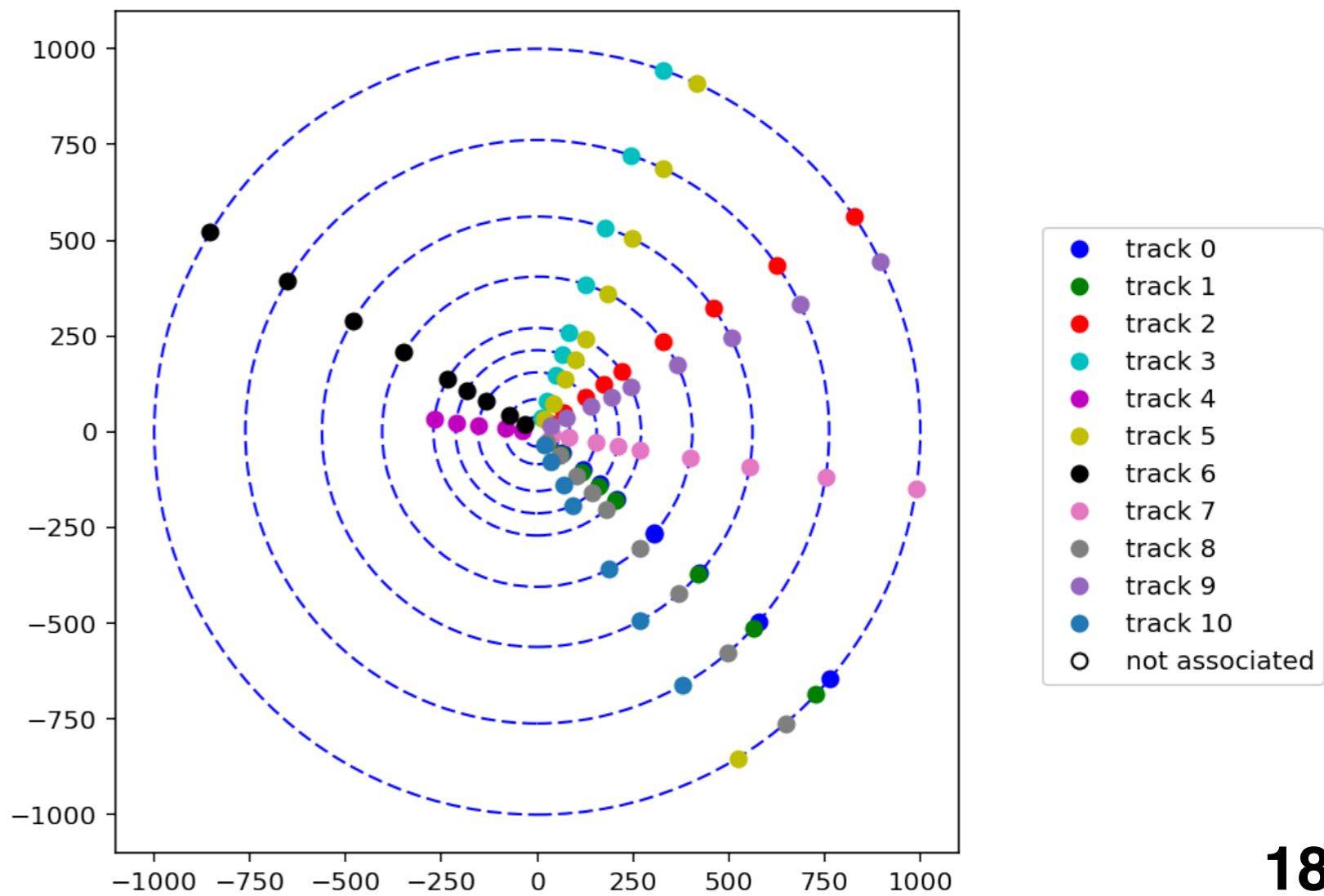


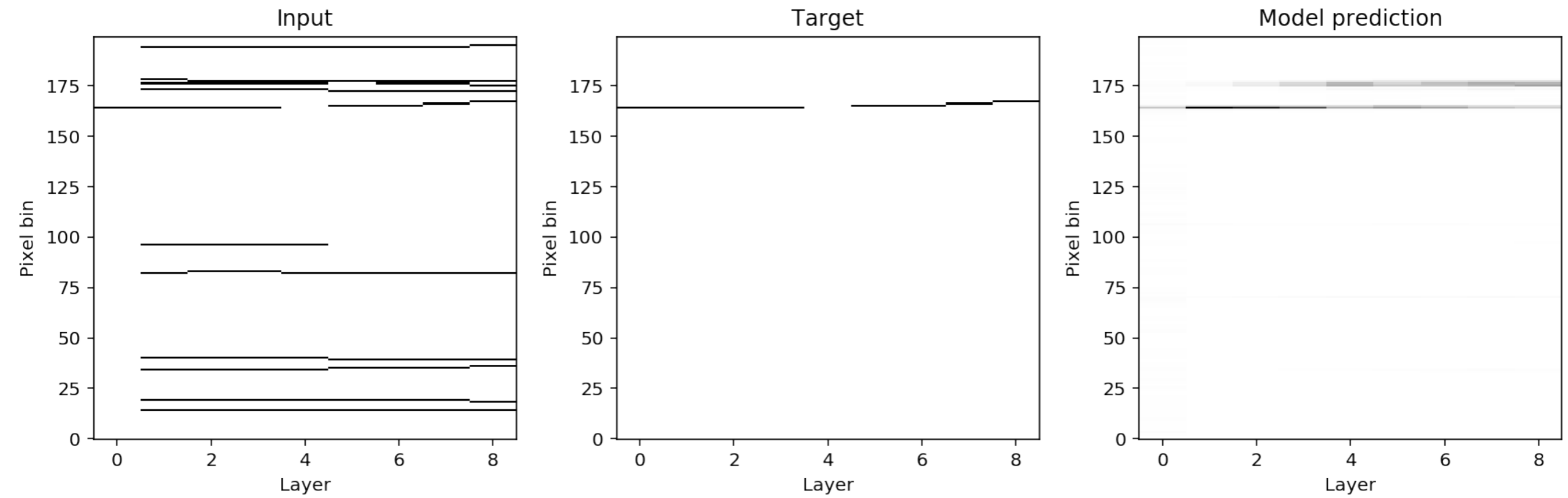
training set:
3000 events



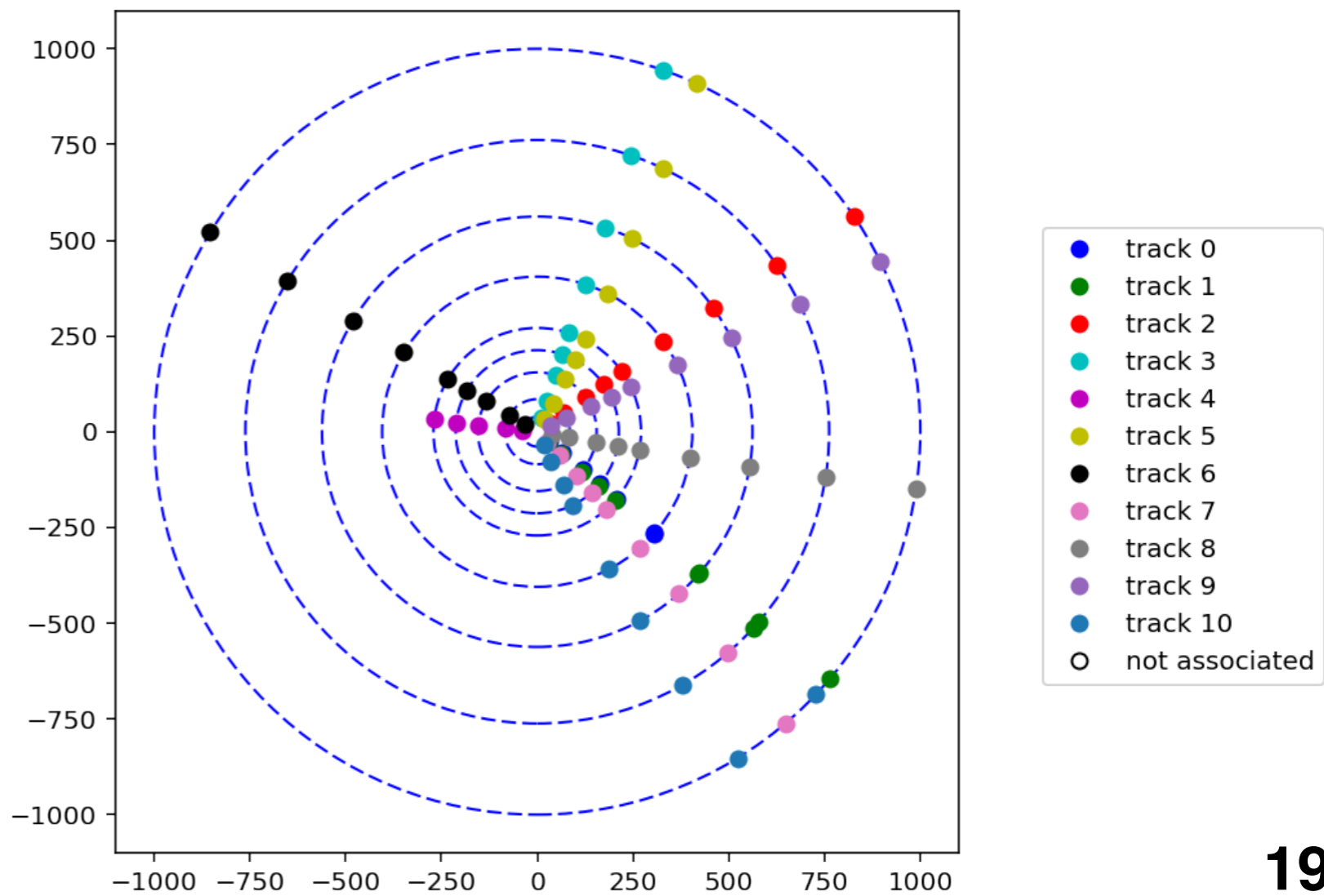


training set:
4000 events

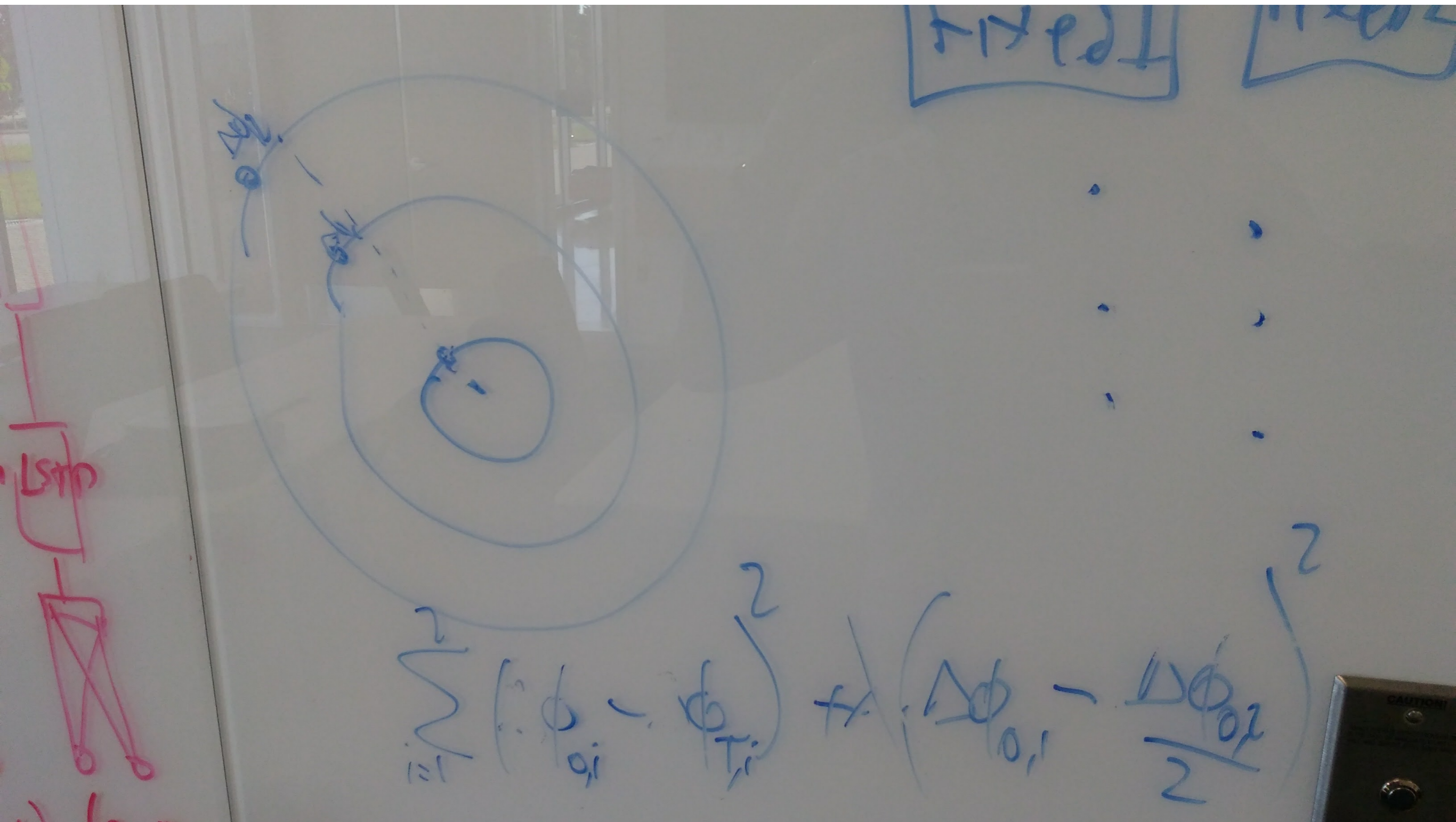




training set:
5000 events



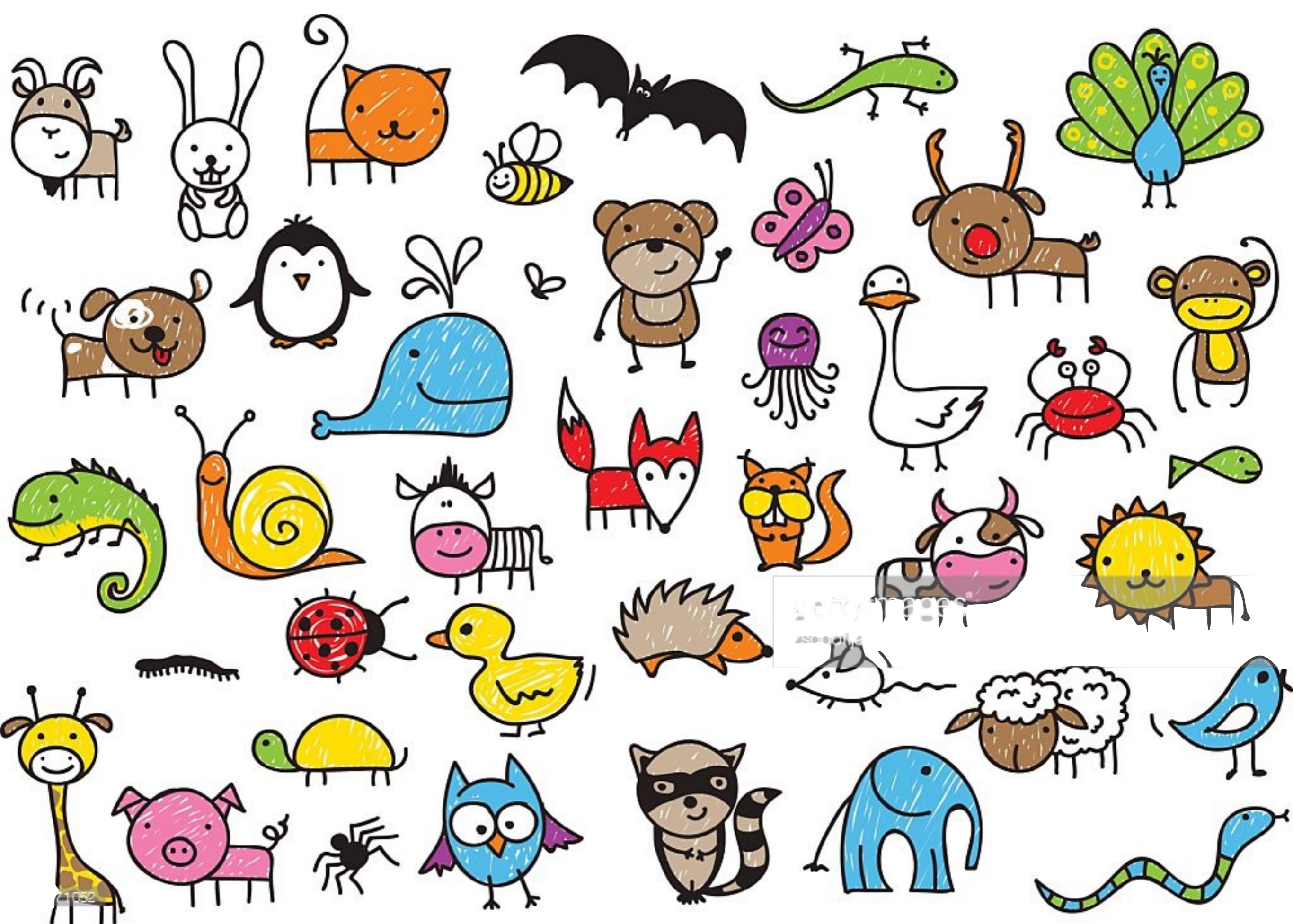
trying to add domain knowledge

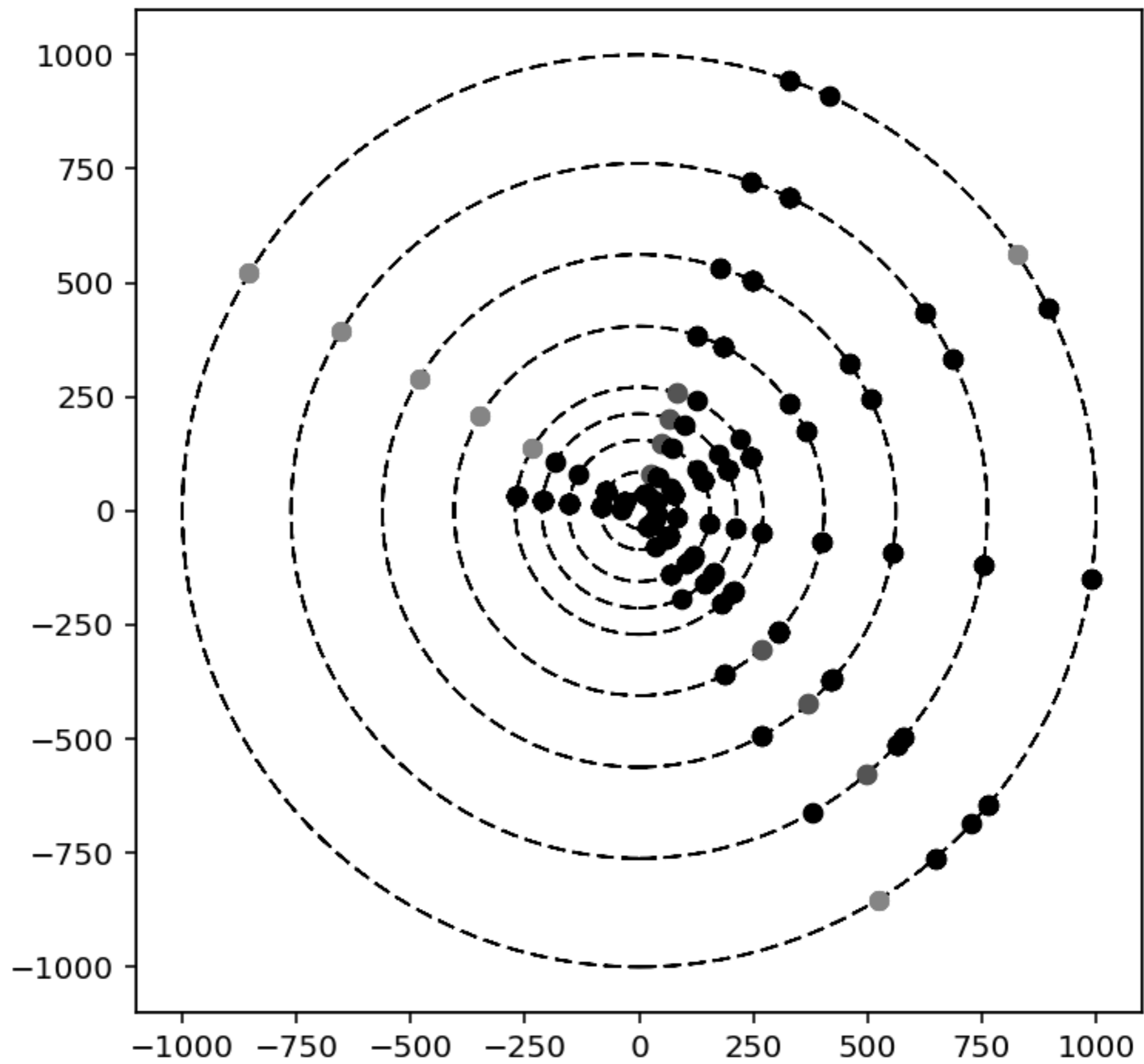


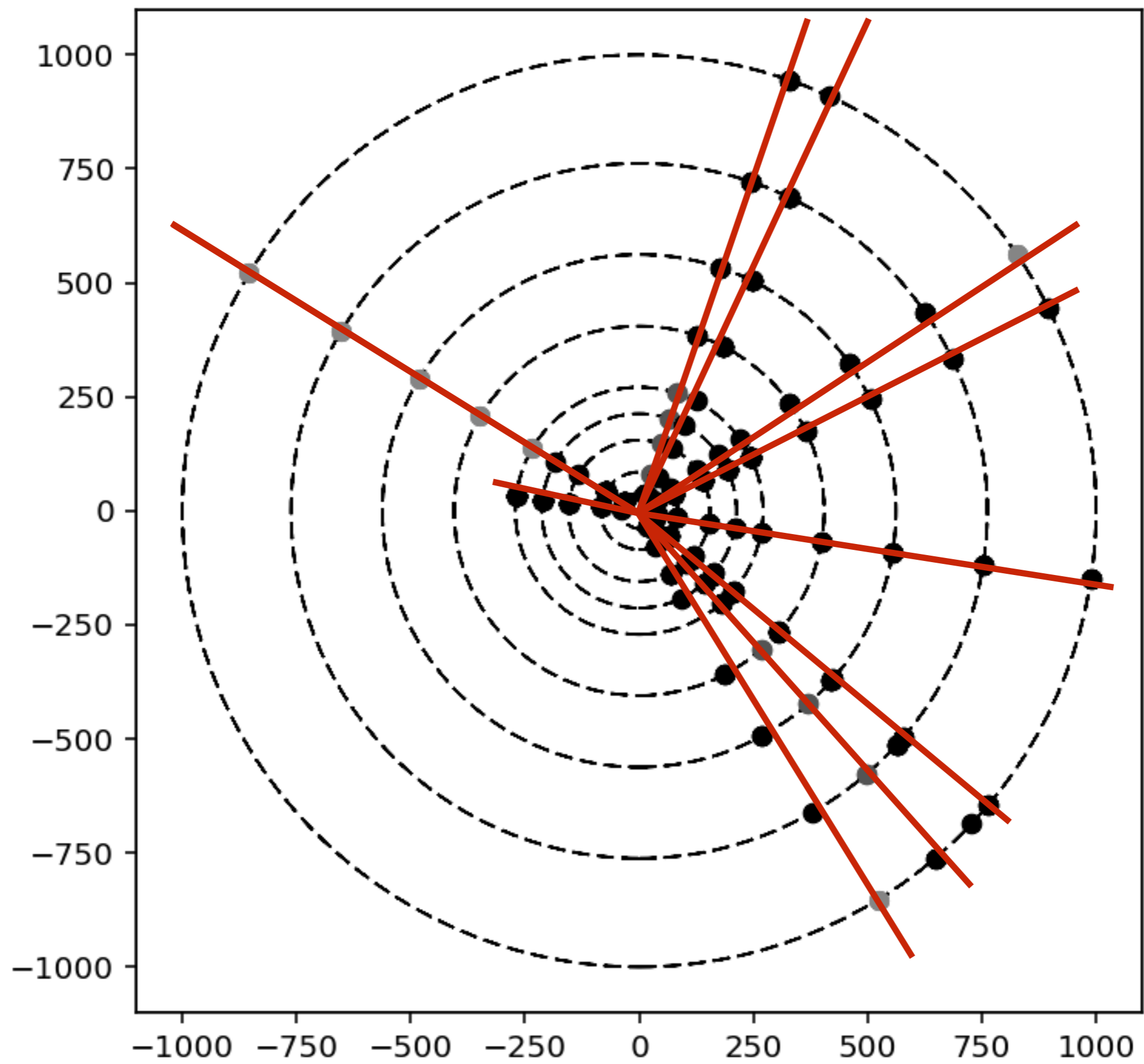
Op-ed: The ML black box

- I was surprised to realize that the algorithm wasn't told anything about tracking
- And how hard it was to ask it what it had learned in the training
- One can define a network in $O(10)$ lines of code
- Understanding and improving the network to address observed failure modes seems like the real challenge









Takeaway

- I came away excited to gain more familiarity with the tools
- And to understand what makes a problem a good candidate for ML-based approaches
- I'm also a bit leery of handing over the reigns to our future robot overlords