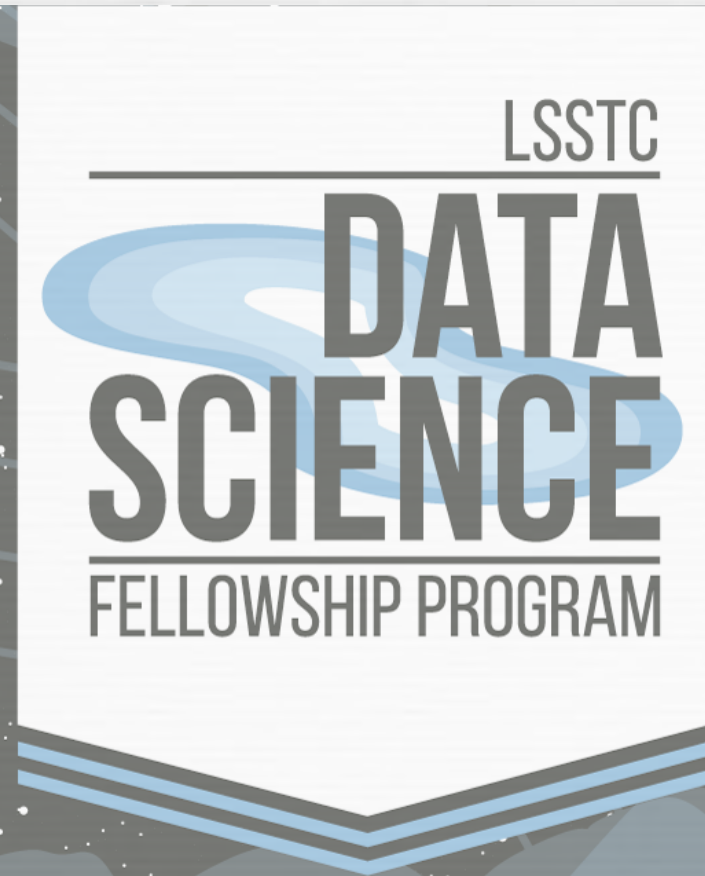


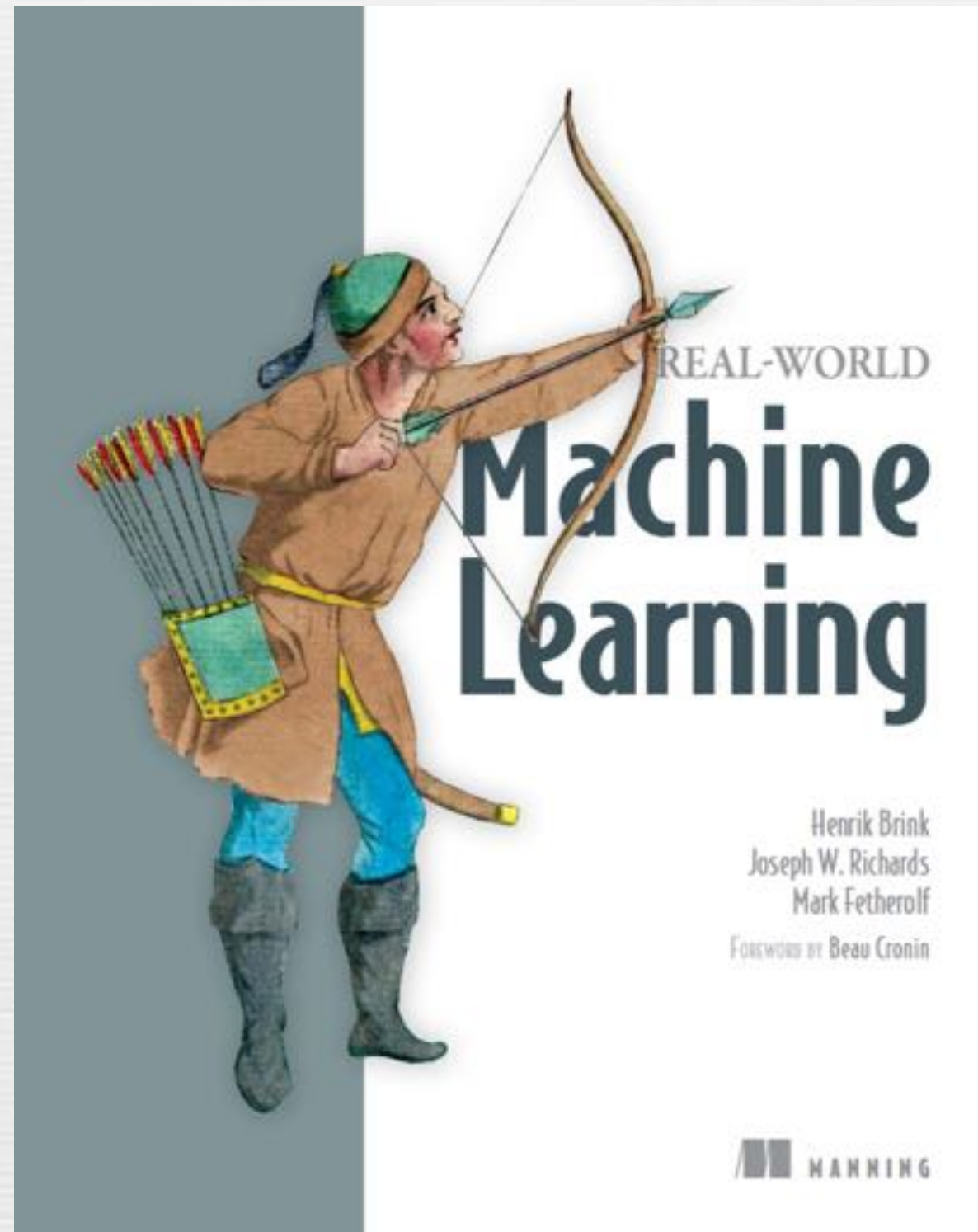
# Developing the Machine Learning Workflow



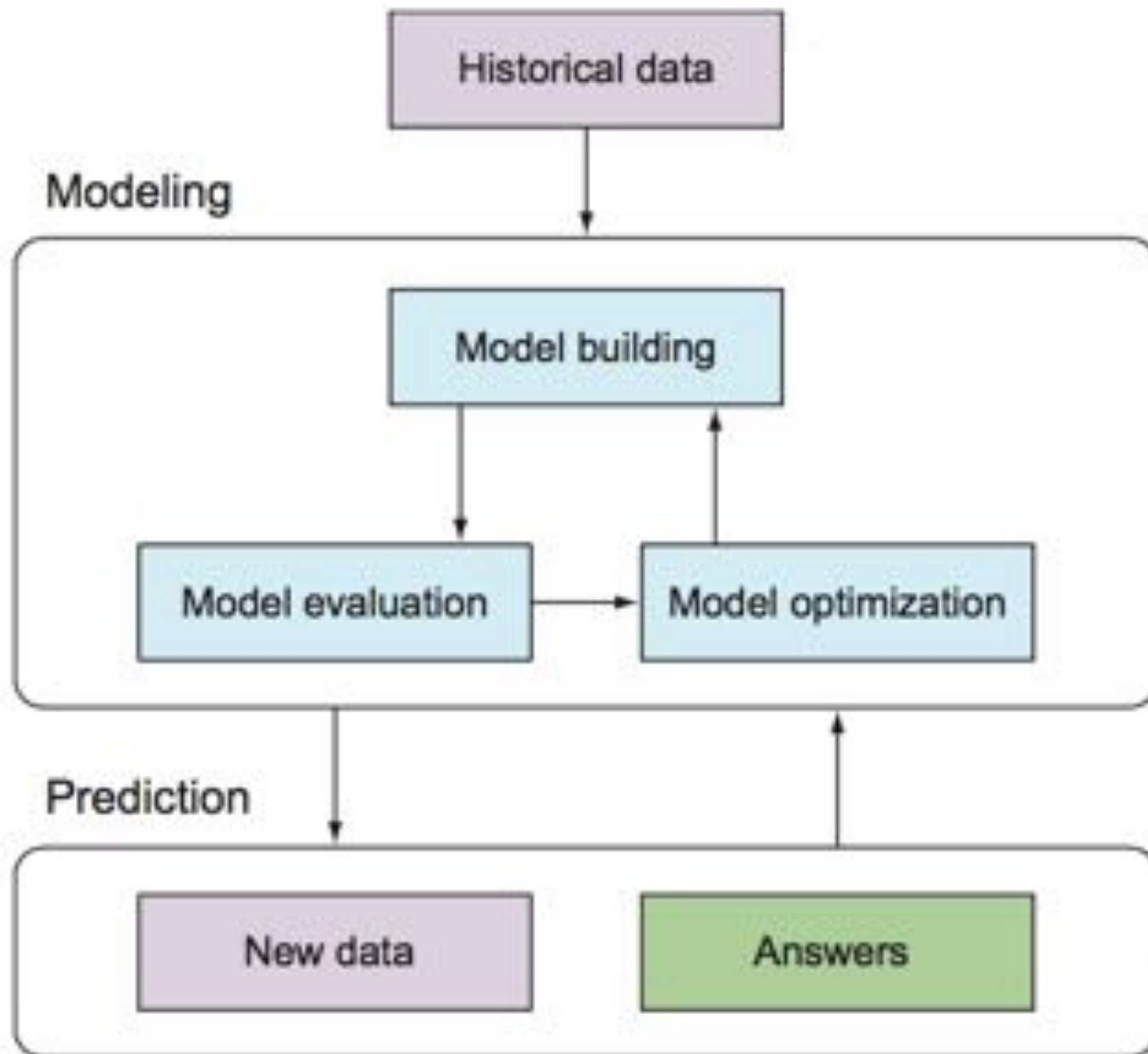
**Adam A Miller**

Northwestern/Adler Planetarium

# Developing the Machine Learning Workflow



# The Machine Learning Workflow



# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled



**Worry  
About  
The Data**

# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled

Select features to use in the model

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Select features to use in the model

Determine “ground truth” or labels for the training set

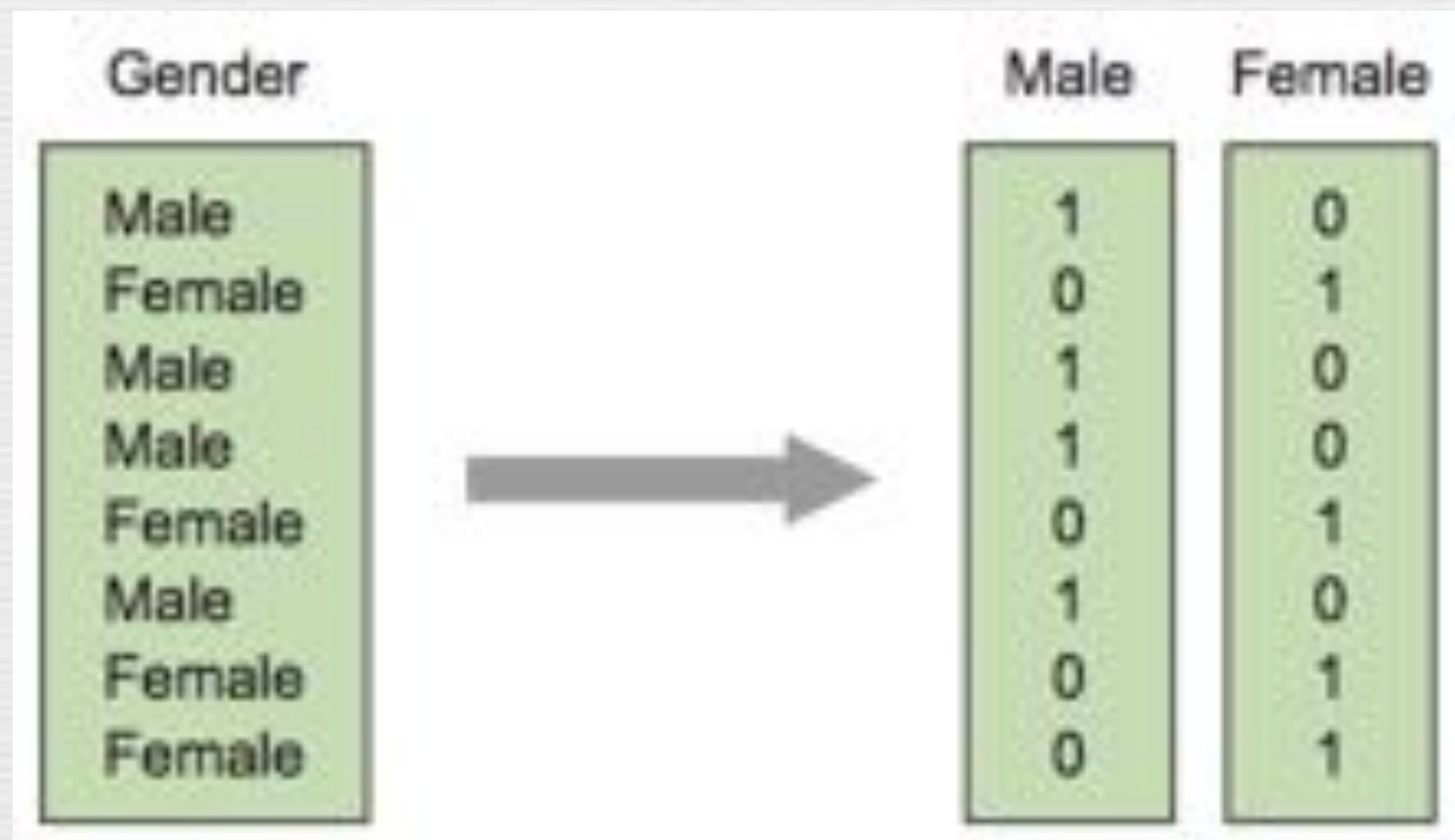
# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled

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Convert categorical features





# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled

Select features to use in the model

Determine “ground truth” or labels for the training set

Convert categorical features

Impute (or throw out?) missing data

# Data Preparation

## Strategies for missing data

Does the missing data have meaning?

Yes - replace with numerical value (-999) or new categorical variable

No - if data set is large with few missing values:

- remove objects with missing data

- else if dataset is large and temporal:

  - replace missing values with preceding value or interpolate

- else if dataset has simple distribution:

  - replace missing values with mean or median

- else:

  - build separate ML model to impute (predict) missing values

# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled

Select features to use in the model

Determine “ground truth” or labels for the training set

Convert categorical features

Impute (or throw out?) missing data

Normalize the features

# Data Preparation

Query, observe, simulate, etc. - collect data that needs to be modeled

Select features to use in the model

Determine “ground truth” or labels for the training set

Convert categorical features

Impute (or throw out?) missing data

Normalize the features

Visualize the data



**Worry  
About  
The Data**

# Feature Engineering

Add new features - if necessary

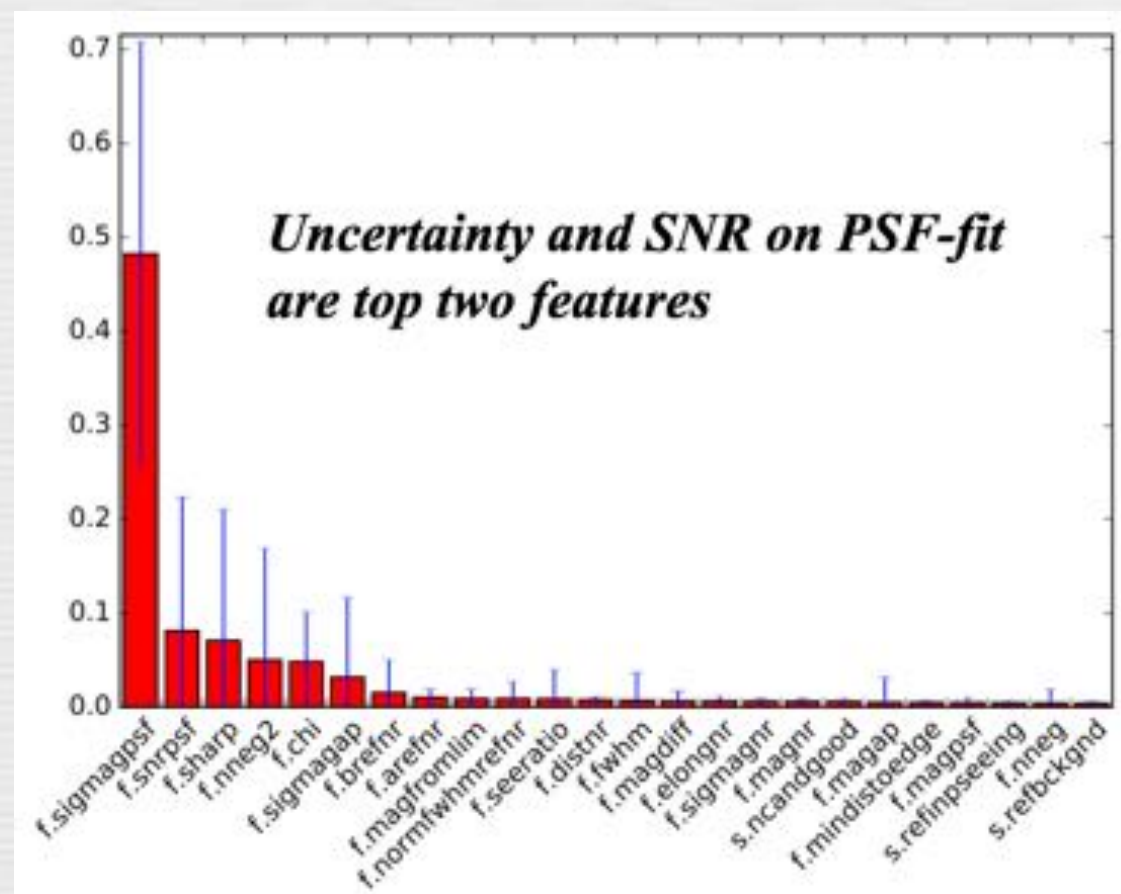
Utilize domain knowledge to create/compute new features

Combine features or represent in an alternative fashion

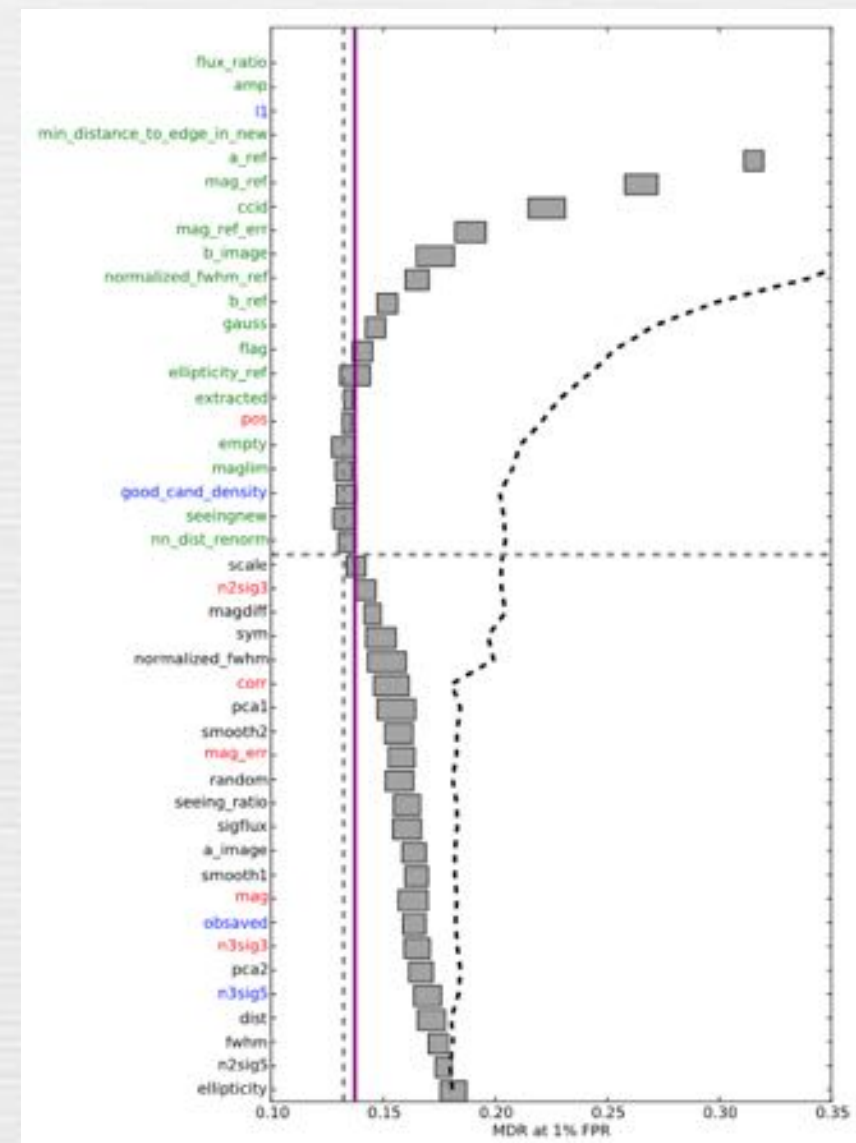
Remove noisy/uninformative features - if necessary

Determine feature importance (RF)

Forward/backward selection to iteratively remove features



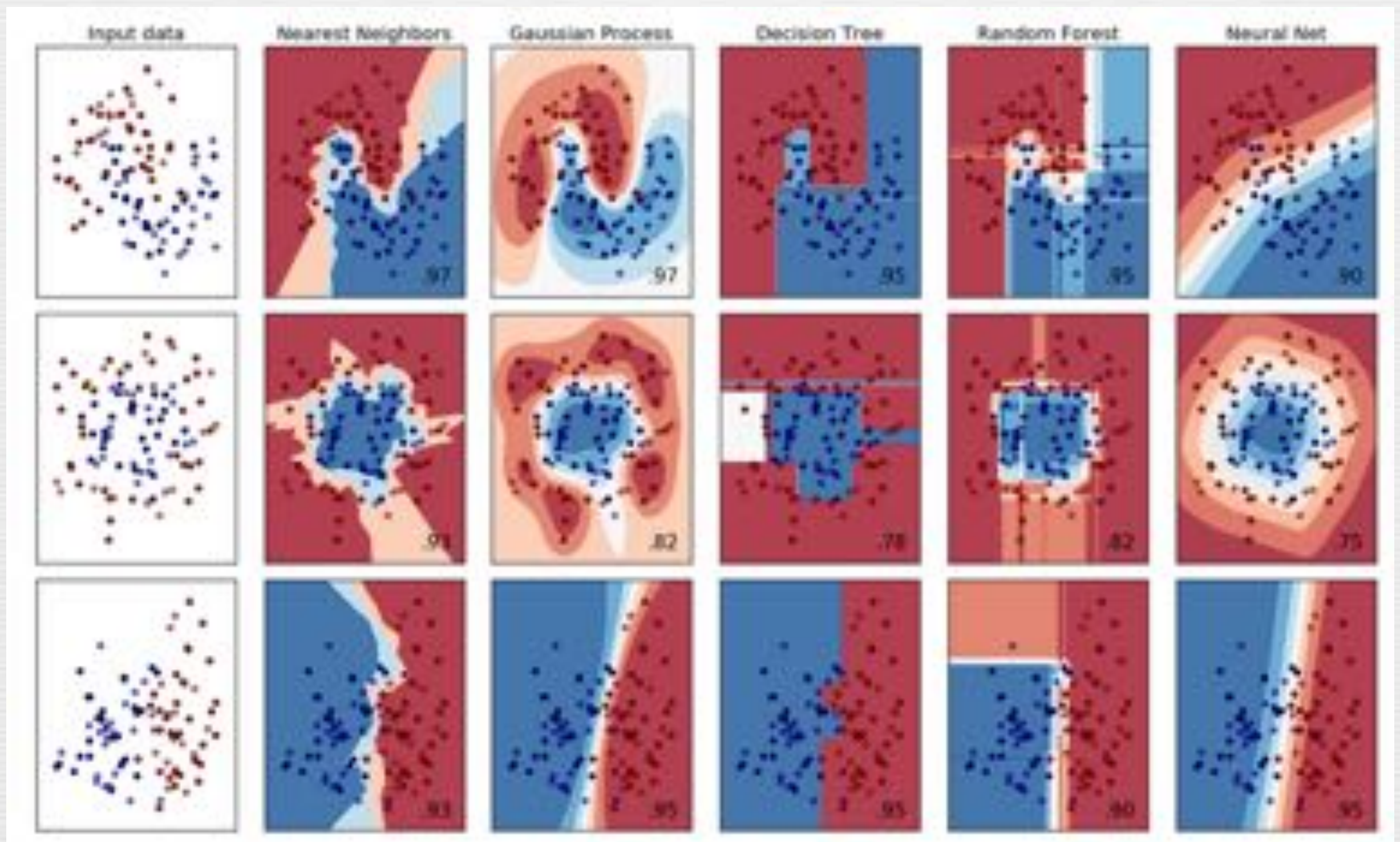
Rebbapragada+16



Brink+13



# Model Selection

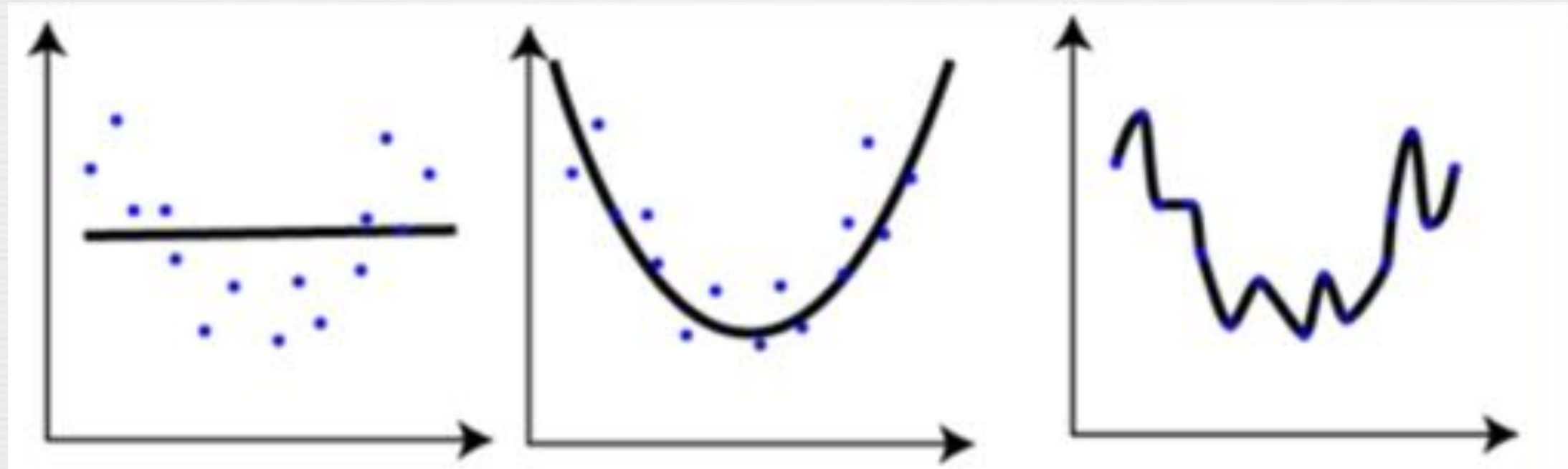


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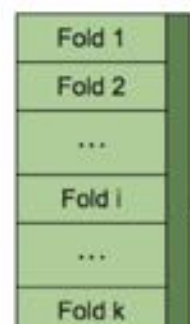
# Model Evaluation

Avoid under- and over-fitting

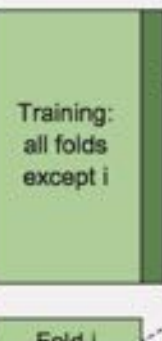


Daniela's lecture (Day 2)

1. Randomly split training instances into  $k$  equal-sized subsets



For  $i$  in  $1:k$

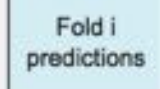


2. Train an ML model on the training subset



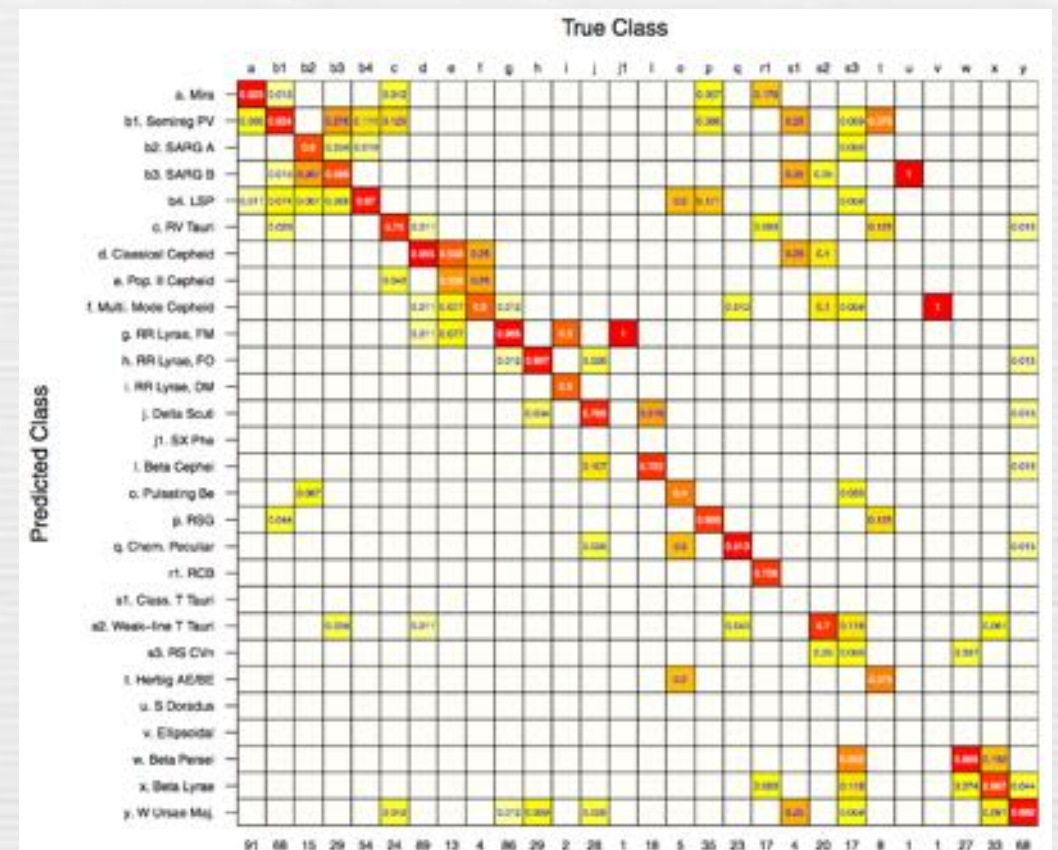
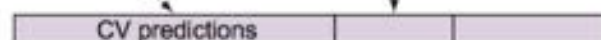
Ignore target when predicting

3. Make predictions on fold  $i$  subset



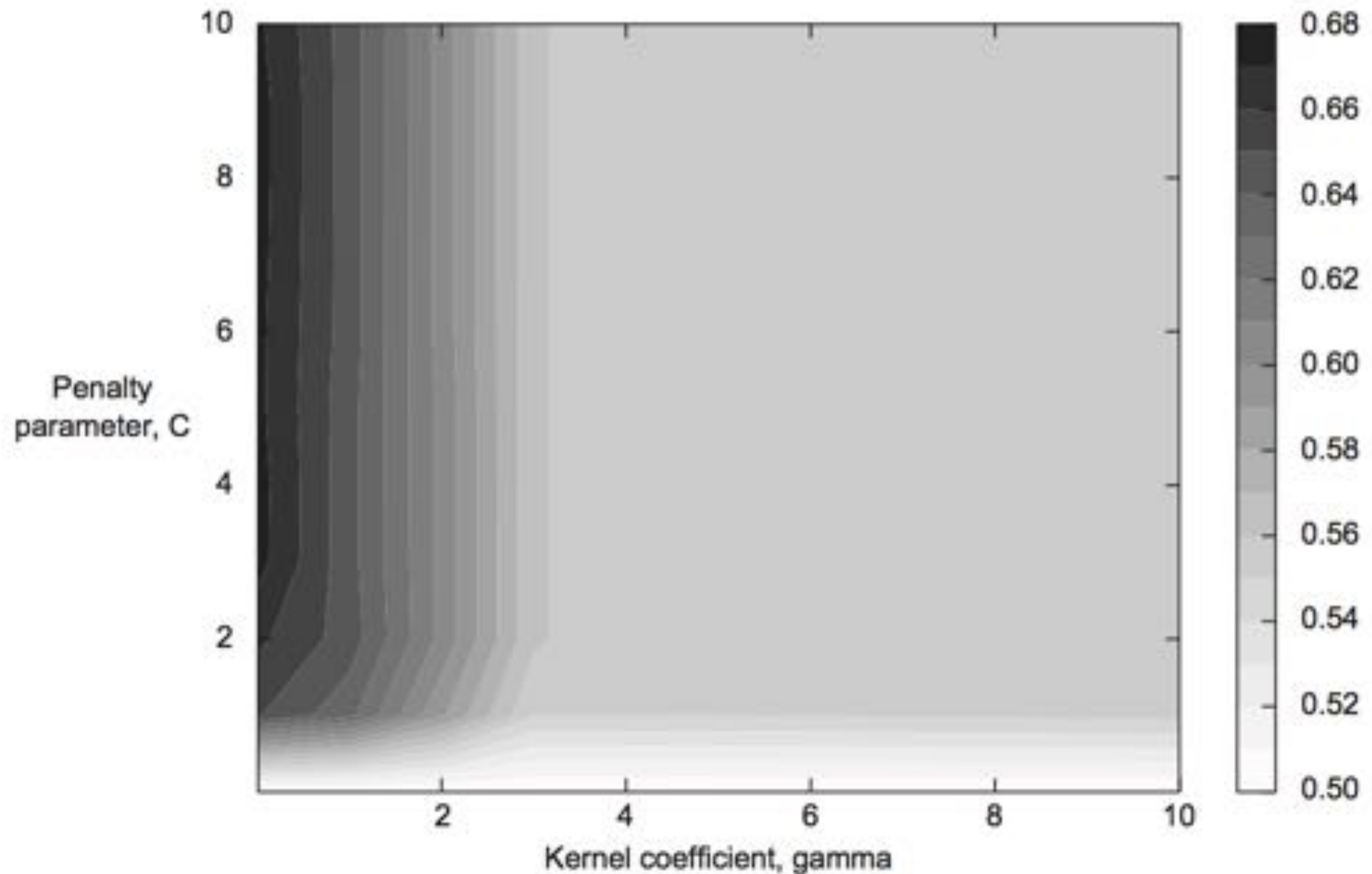
5. Compare CV predictions to target to assess accuracy

4. Store fold  $i$  predictions in the CV predictions array

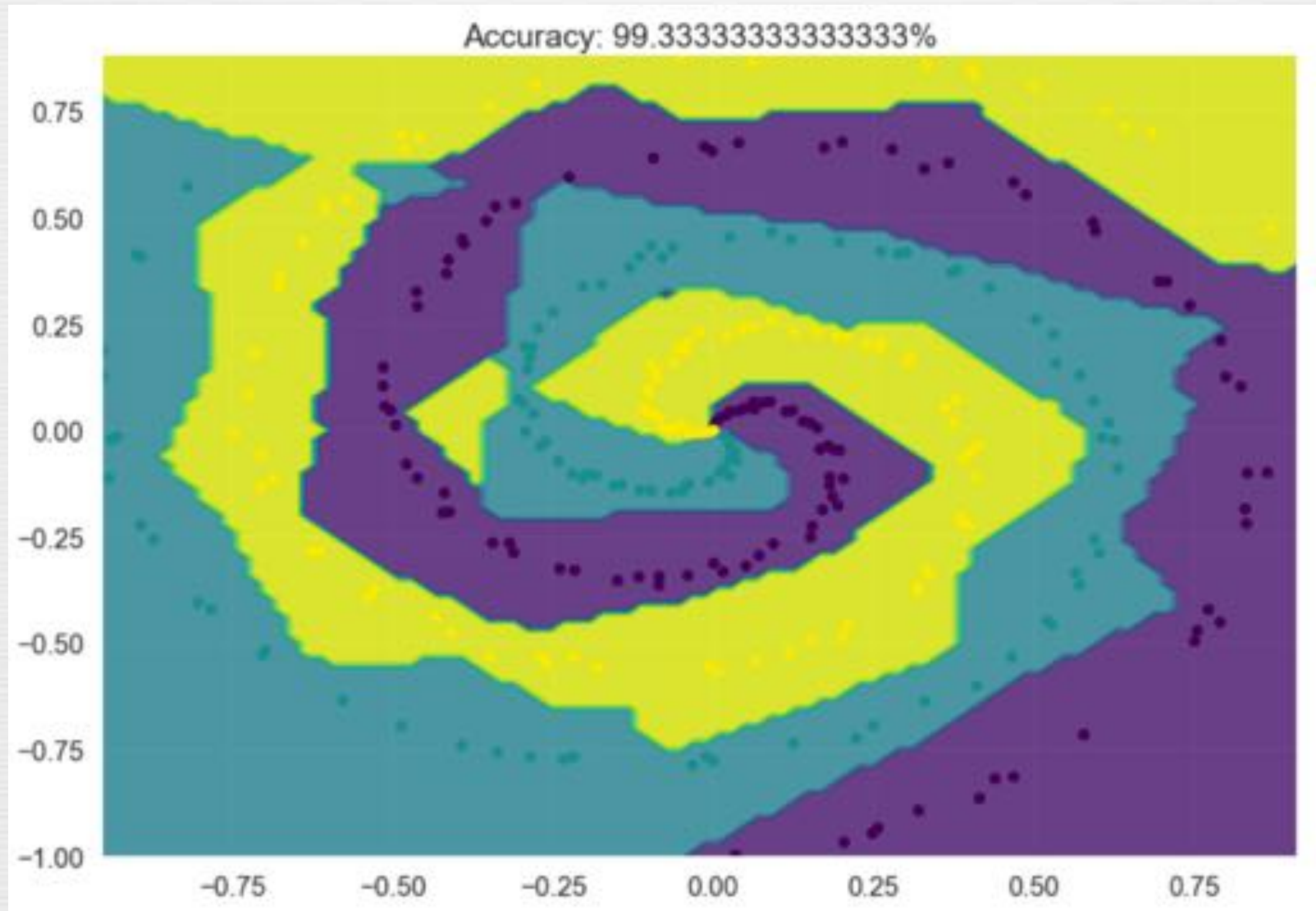


# Model Optimization

Identify optimal tuning parameters via grid search



# Model Prediction





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# The Machine Learning Workflow

