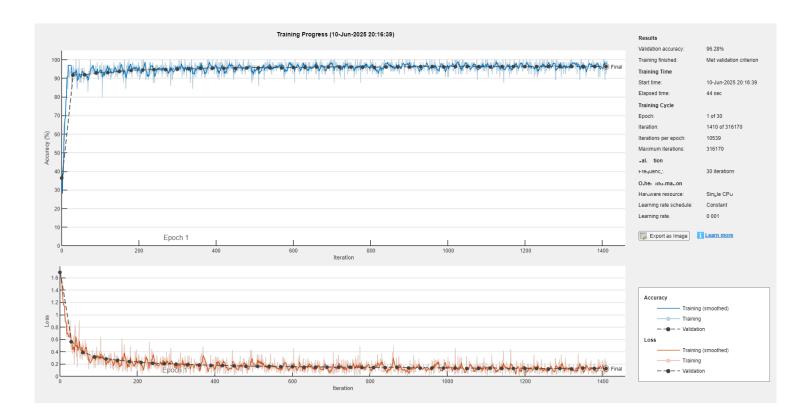
```
% Features: n × m → m × n for deep learning input
XTrain = X_train_raw';
XVal = X_val';
XTest = X_test';

% Labels: Convert to categorical
YTrain = categorical(Y_train_raw);
YVal = categorical(Y_val);
YTest = categorical(Y_test);
```

```
options = trainingOptions('adam', ...
    'MaxEpochs', 30, ...
    'MiniBatchSize', 64, ...
    'Shuffle', 'every-epoch', ...
    'ValidationData', {XVal', YVal}, ...
    'ValidationFrequency', 30, ...
    'ValidationPatience',5,...
    'Verbose', false, ...
    'Plots', 'training-progress');
```

```
net = trainNetwork(XTrain', YTrain, layers, options);
```



```
% Predict labels
YPred = classify(net, XTest');

% Confusion matrix and accuracy
confusionchart(YTest, YPred);
```

F		125			63
N		131316	25	145	1146
True Class		329	760		107
SVEB		1578		562	641
VEB		1119	8	18	6606
	F N Q SVEB VEB Predicted Class				

```
accuracy = sum(YPred == YTest) / numel(YTest);
fprintf('Test Accuracy: %.2f%\n', accuracy * 100);
```

Test Accuracy: 96.33%

```
% Assume YTest and YPred are categorical arrays
classes = categories(YTest);
numClasses = numel(classes);
% Get confusion matrix
confMat = confusionmat(YTest, YPred);
% Preallocate
precision = zeros(numClasses, 1);
recall = zeros(numClasses, 1);
f1_score = zeros(numClasses, 1);
support = sum(confMat, 2); % True instances per class
% Compute per-class metrics
for i = 1:numClasses
    TP = confMat(i, i); % True Positives
    FP = sum(confMat(:, i)) - TP; % False Positives
    FN = sum(confMat(i, :)) - TP; % False Negatives
    precision(i) = TP / (TP + FP + eps);
```

```
recall(i) = TP / (TP + FN + eps);
f1_score(i) = 2 * (precision(i) * recall(i)) / (precision(i) + recall(i) + eps);
end

% Display as table
metrics_table = table(classes, precision, recall, f1_score, support, ...
    'VariableNames', {'Class', 'Precision', 'Recall', 'F1_Score', 'Support'});
disp(metrics_table)
```

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