

# Compiler to JVM

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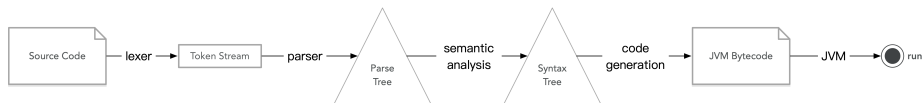
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# Language Specification

```
namespace Test {  
    struct MyStruct {  
        int a = f();  
    }  
  
    double d = 4;  
  
    def void main(String[] args) {  
        print("Hello world!\n");  
    }  
  
    def int f() {  
        return 3;  
    }  
}
```

# Compiler Architecture



# Third Party Library

haha

# Type I and Type II Errors

## False Negative Errors

A false negative is an error in which a test/study result is negative but in reality it is positive.

## True Positive Rate, TPR

$$TPR = \frac{TP}{TP + FN} \quad (1)$$

## False Positive Rate, FPR

$$FPR = \frac{FP}{FP + TN} \quad (2)$$

## True Positive Rate, TPR

$$TPR = \frac{TP}{TP + FN} \quad (3)$$

## False Positive Rate, FPR

$$FPR = \frac{FP}{FP + TN} \quad (4)$$

## ROC Space

An ROC space is defined by **FPR** and **TPR** as x and y axes, respectively, which depicts relative trade-offs between true positive (benefits) and false positive (costs)

- Each prediction result represents one point in the ROC space.



- What if a result point locates in the diagonal line?

## Accuracy

$$Accuracy = \frac{TP + TN}{TP + FP + FN + TN} \quad (5)$$

- Suppose the number of positive results equals to the number of negative results in reality.

$$TP + FN = FP + TN \quad (6)$$

- Then a result point located in the diagonal line means:

$$\begin{aligned} \frac{TP}{TP + FN} &= \frac{FP}{FP + TN} \\ \Rightarrow Accuracy &= \frac{TP + TN}{TP + FP + FN + TN} = 50\% \end{aligned} \quad (7)$$

- So the diagonal line are called line of no-discrimination (line of random guess)

- The result located in the diagonal line has 50% accuracy

- The closer a result is to the upper left corner, the better it predicts.
- The closer a result is to the lower right corner, the worse it predicts.

## ROC Curve

Fix binary classifier model and change the threshold, the result points will generate the ROC curve.

## Example

# AUC

## AUC

Area under the curve of ROC curve.

## Example

■  $AUC \leq 1$

## Example

- $AUC = 1$ , perfect classifier
- $0.5 < AUC < 1$ , better than random guess
- $AUC = 0.5$ , same as random guess
- $AUC < 0.5$ , worse than random guess



# Balance of FPR and FNR

- With the decrease of FPR, FNR increases.

## Example

- **Prediction of earthquakes:** we prefer to predict every earthquake successfully, which means that we can tolerate the increase of FPR compared to FNR.

- **Conviction:** By the presumption of innocence, we prefer to be cautious when involving criminal justice, so we have to tolerate larger FNR compared to FPR.