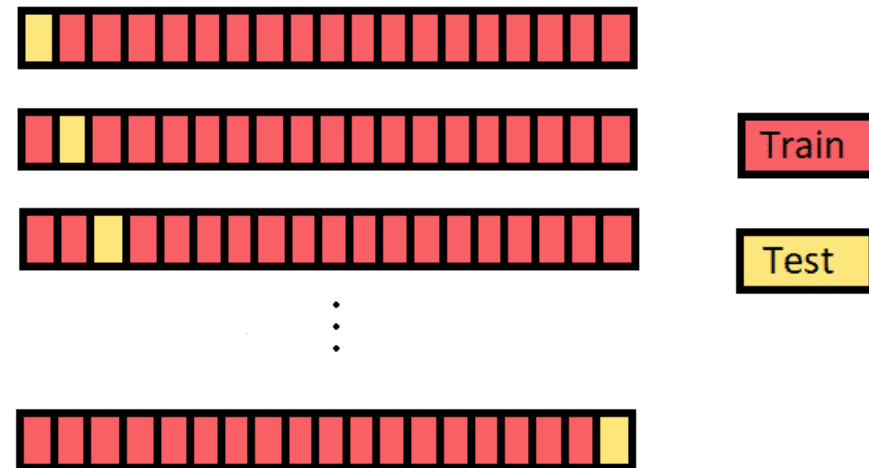
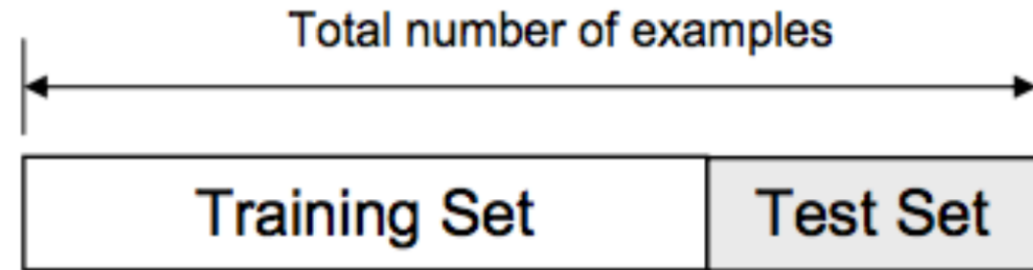


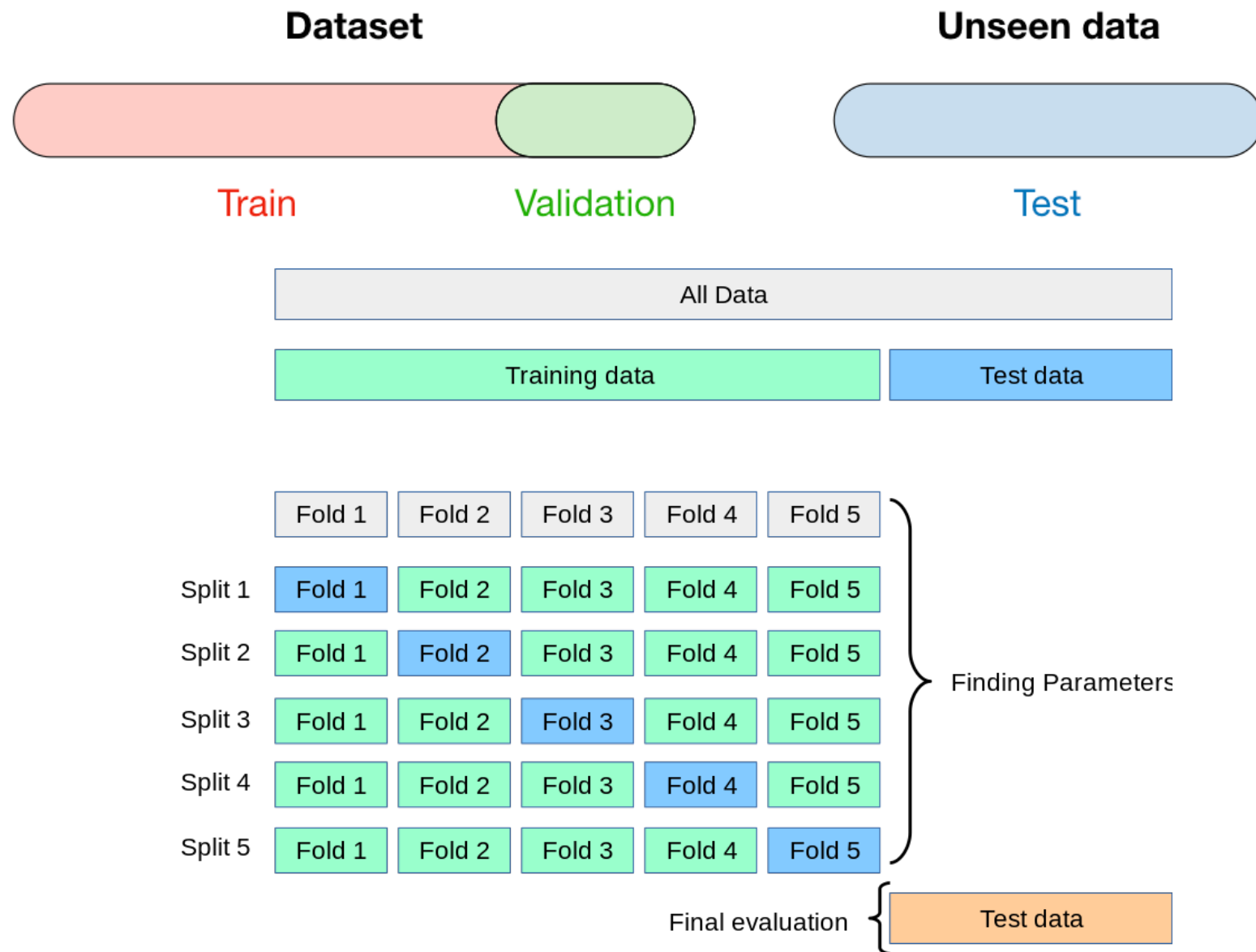
Метрики

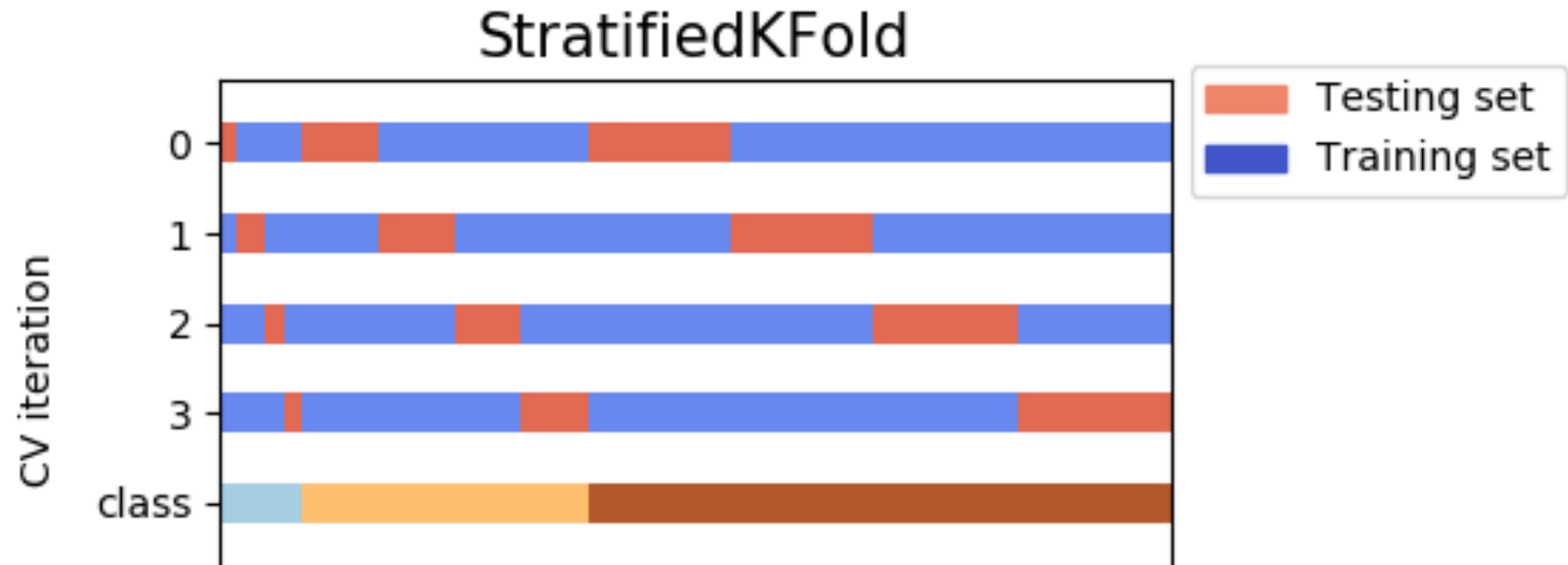
Сбертех, МФТИ

- Train-test split
- Cross-validation
- Leave-one-out

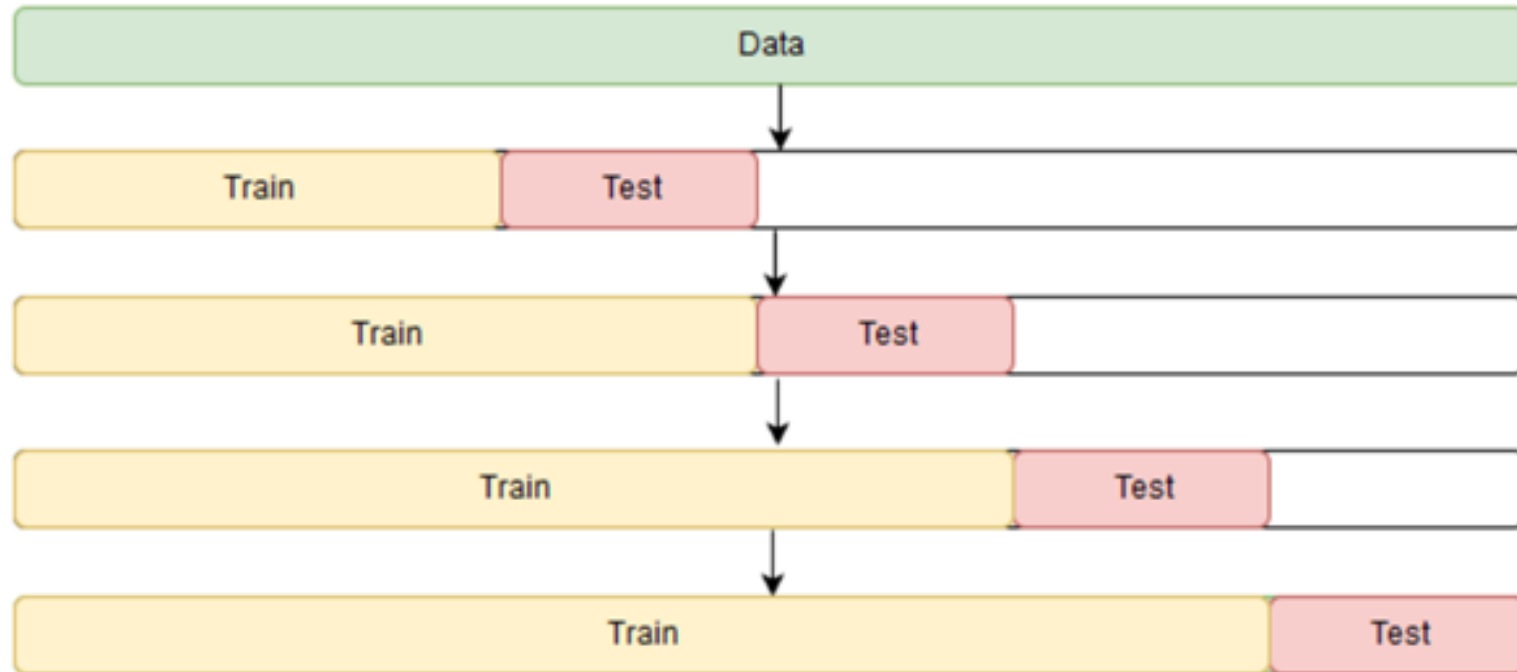


Стратегия валидации





Time series



Метрики бинарной классификации

$$Accuracy(\hat{y}, y) = \frac{1}{\ell} \sum_{i=1}^{\ell} [\hat{y}_i = y_i]$$

True Class

Positive

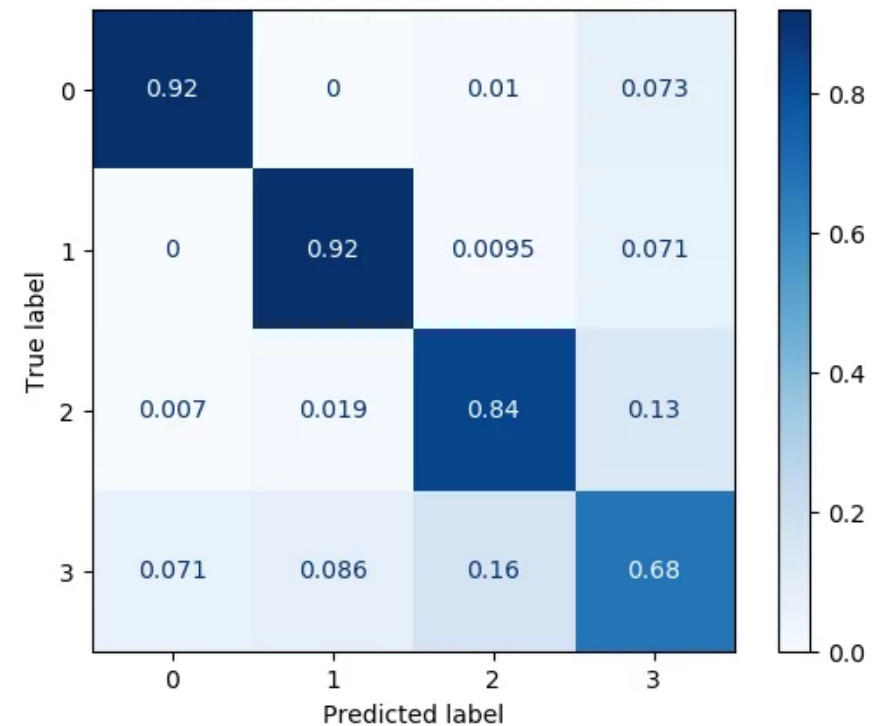
Negative

Predicted Class

Positive
Negative

Positive	TP	FP
Negative	FN	TN

Confusion matrix for our classifier



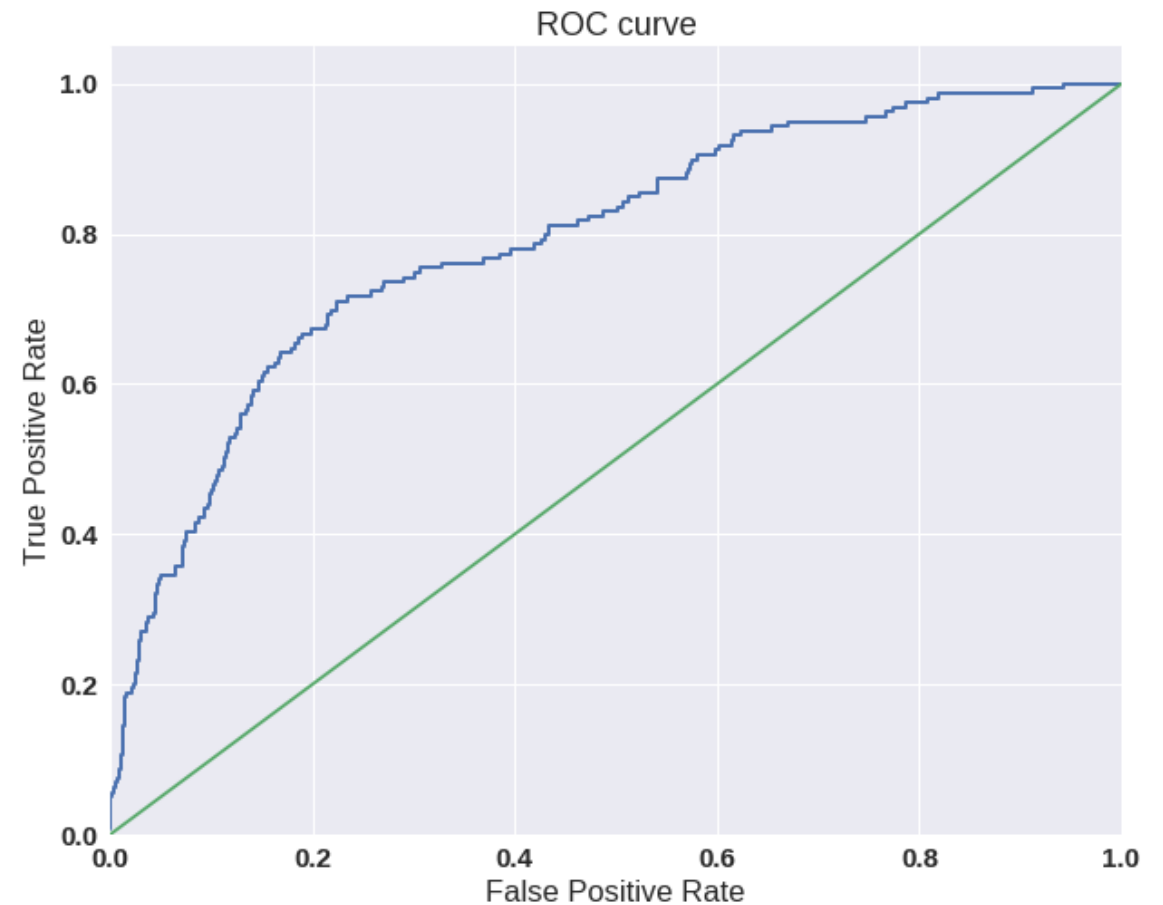
$$precision = \frac{TP}{TP + FP} \quad F_{\beta} = (1 + \beta^2) \frac{precision \cdot recall}{\beta^2 \cdot precision + recall}$$

$$recall = \frac{TP}{TP + FN} \quad F_1 = \frac{2 \cdot precision \cdot recall}{precision + recall}$$

Метрики бинарной классификации

$$TPR = \frac{TP}{TP + FN}$$

$$FPR = \frac{FP}{FP + TN}$$



<https://alexanderdyakonov.wordpress.com/2017/07/28/auc-roc-площадь-под-кривой-ошибок/>

Метрики бинарной классификации

$$TPR = \frac{TP}{TP + FN} = \frac{90}{90 + 10} = 0.9$$

$$precision = \frac{TP}{TP + FP} = 90 / (90 + 10) = 0.9$$

$$FPR = \frac{FP}{FP + TN} = \frac{10}{10 + 999890} = 0.00001$$

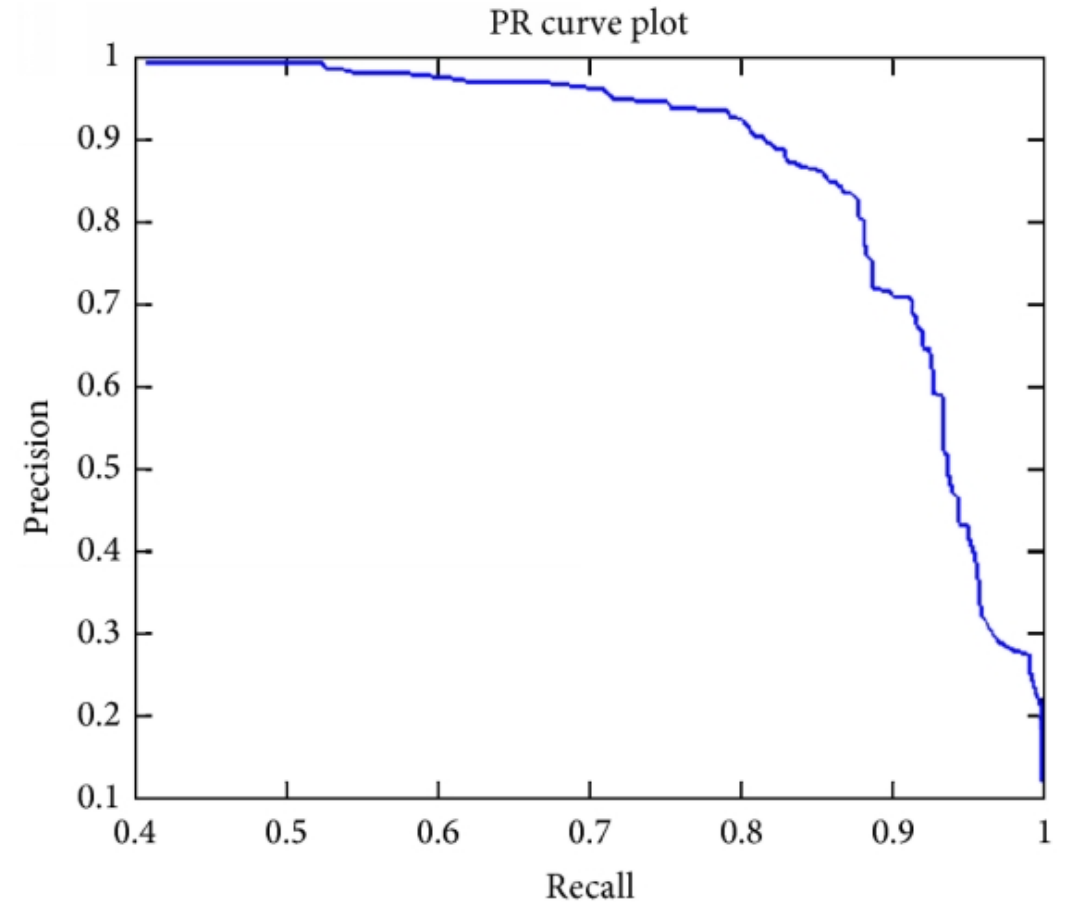
$$recall = \frac{TP}{TP + FN} = 90 / (90 + 10) = 0.9$$

$$TPR = \frac{TP}{TP + FN} = \frac{90}{90 + 10} = 0.9$$

$$precision = \frac{TP}{TP + FP} = \frac{90}{90 + 1910} = 0.045$$

$$FPR = \frac{FP}{FP + TN} = \frac{1910}{1910 + 997990} = 0.001$$

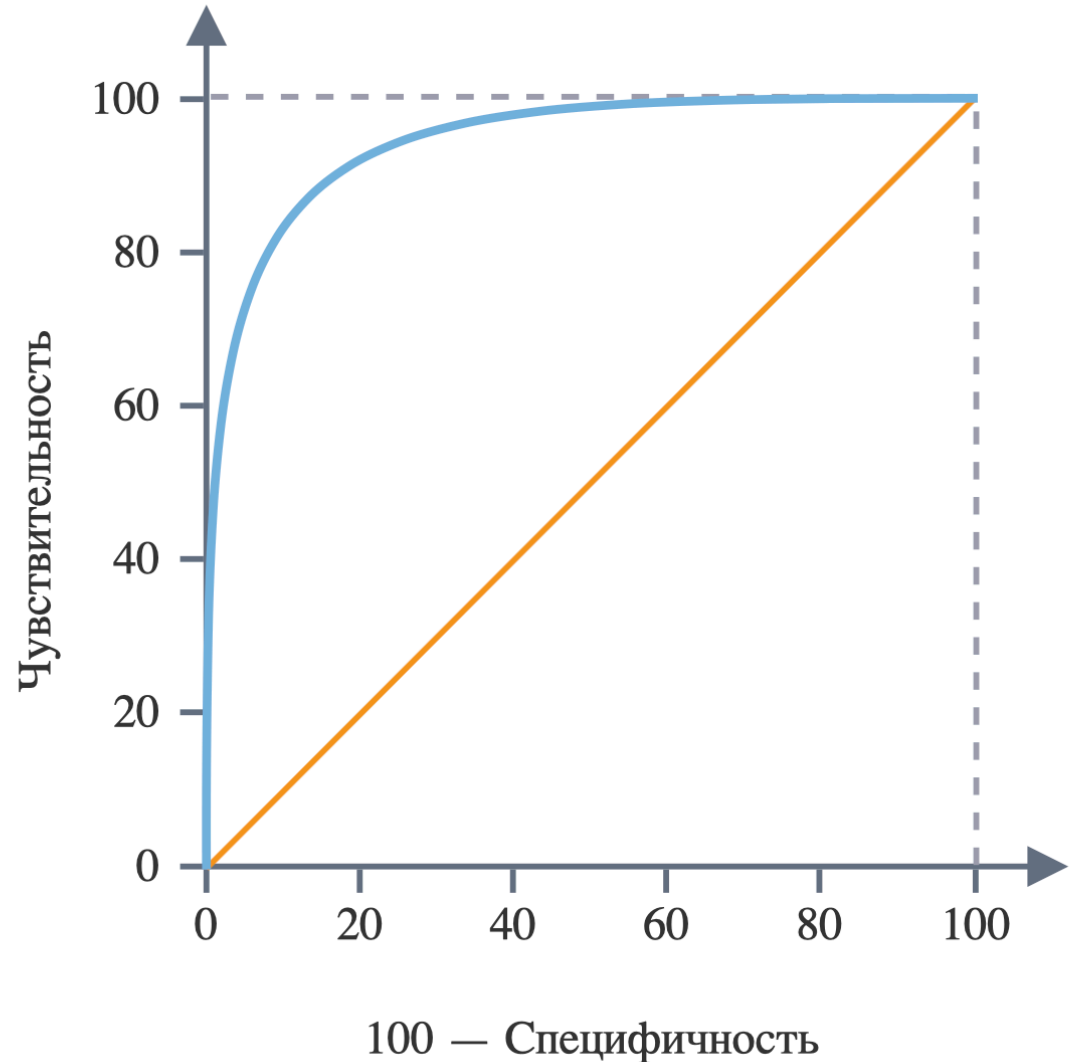
$$recall = \frac{TP}{TP + FN} = \frac{90}{90 + 10} = 0.9$$



Метрики бинарной классификации

$$Sensitivity = \frac{TP}{TP+FN}$$

$$Specificity = \frac{TN}{TN+FP}$$

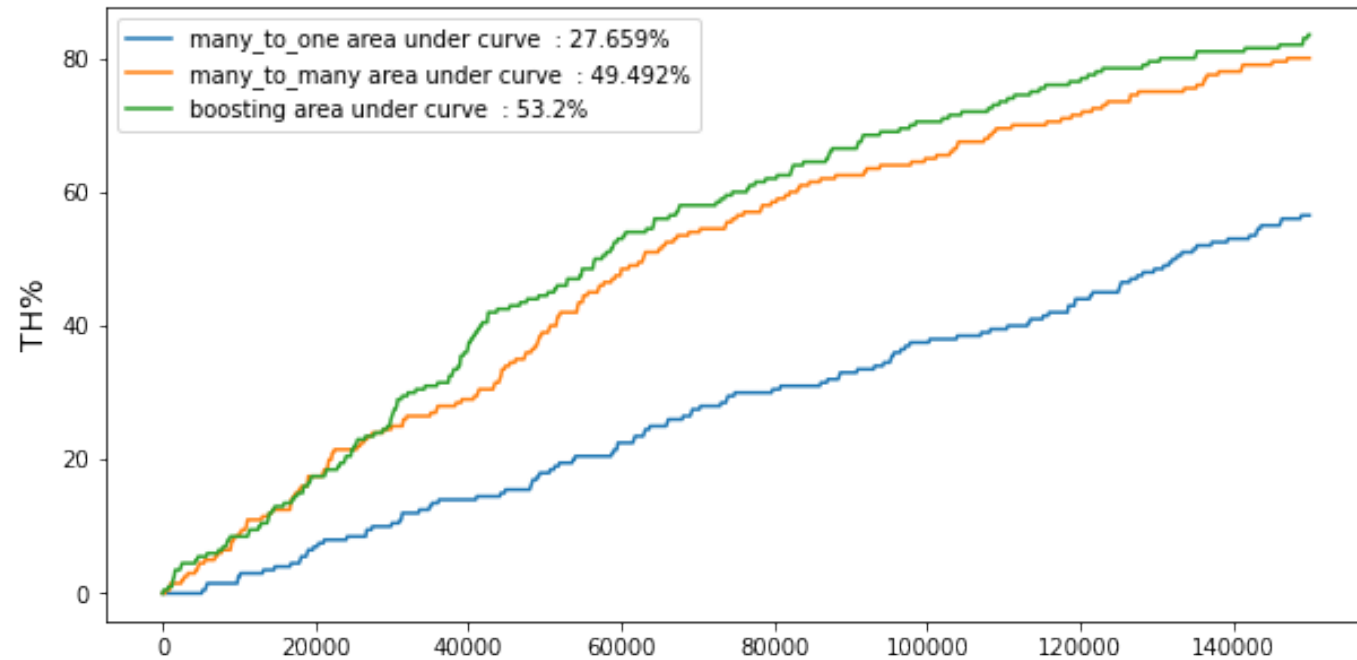


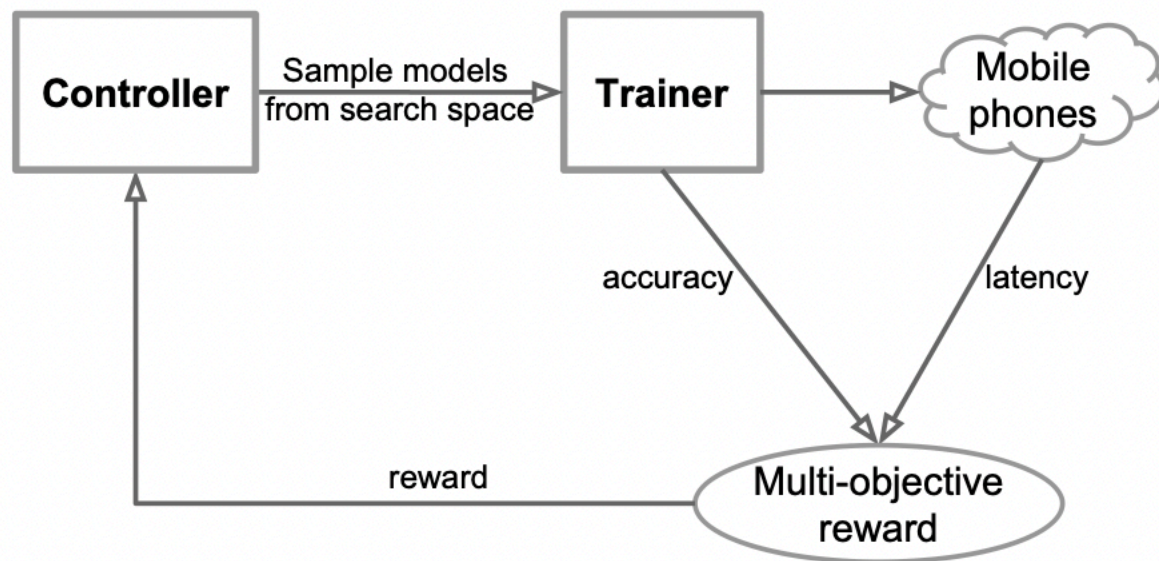
		Predicted Values		
		Setosa	Versicolor	Virginica
Actual Values	Setosa	16 (cell 1)	0 (cell 2)	0 (cell 3)
	Versicolor	0 (cell 4)	17 (cell 5)	1 (cell 6)
	Virginica	0 (cell 7)	0 (cell 8)	11 (cell 9)

$$RMSE = \sqrt{\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

$$MSE = \frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

$$MAE = \frac{1}{n} \sum_{i=1}^n |y_i - \hat{y}_i|$$





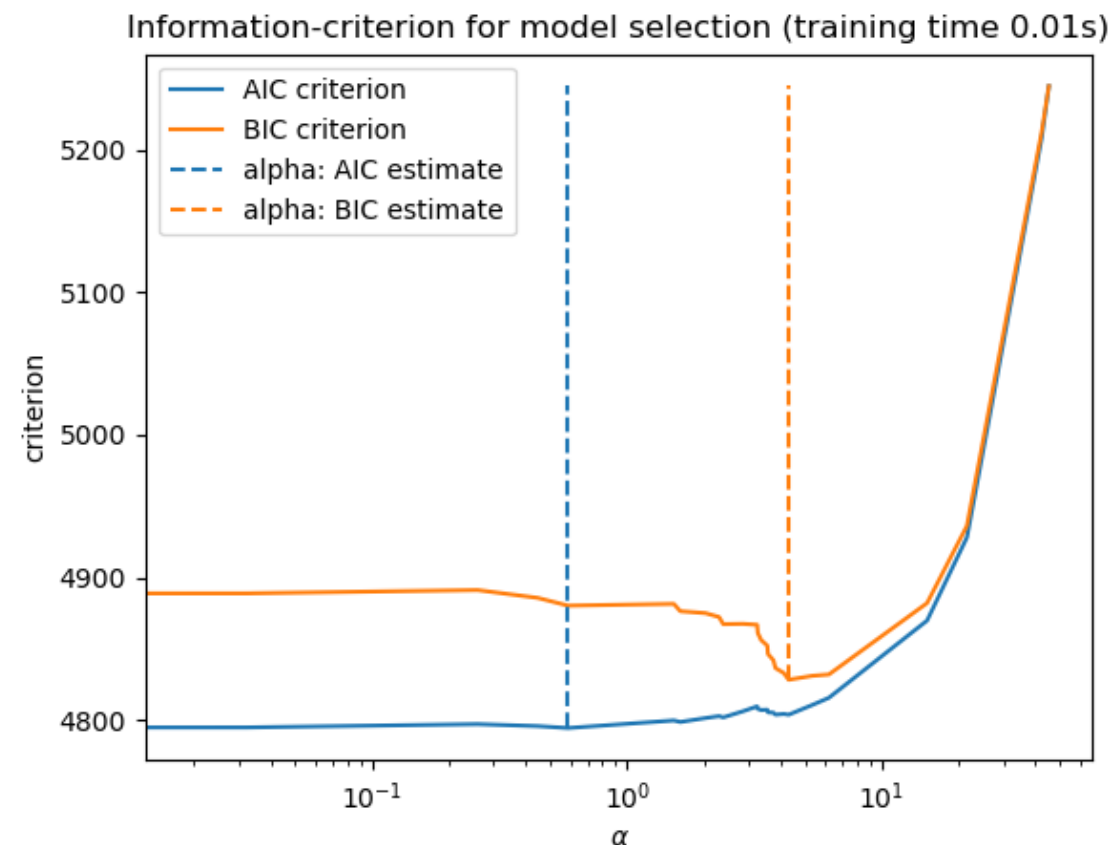
$$\underset{m}{\text{maximize}} \quad ACC(m) \times \left[\frac{LAT(m)}{T} \right]^w$$

Figure 1: An Overview of Platform-Aware Neural Architecture Search for Mobile.

<https://arxiv.org/pdf/1807.11626.pdf>

$$AIC = 2k - 2\ln(L)$$

$$BIC = -2\ln(L) + k\ln(n)$$



<https://machinelearningmastery.com/probabilistic-model-selection-measures/>

https://scikit-learn.org/stable/auto_examples/linear_model/plot_lasso_model_selection.html

Критерий Стьюдента

- Параметрические
- Непараметрические

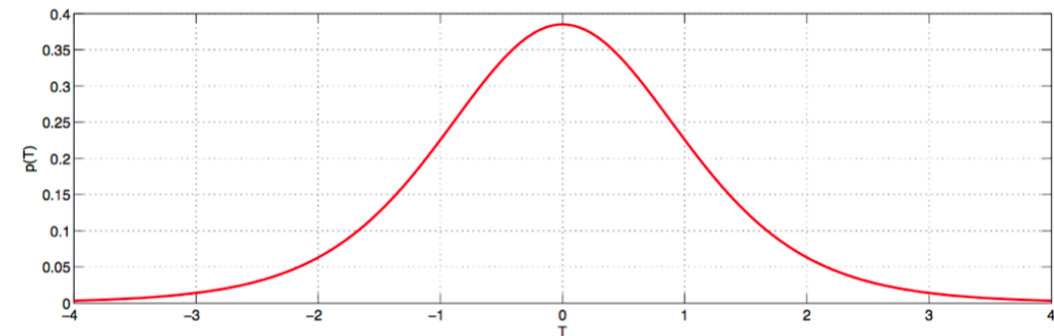
выборки: $X_1^{n_1} = (X_{11}, \dots, X_{1n_1}), X_1 \sim N(\mu_1, \sigma_1^2)$
 $X_2^{n_2} = (X_{21}, \dots, X_{2n_2}), X_2 \sim N(\mu_2, \sigma_2^2)$
 σ_1, σ_2 неизвестны

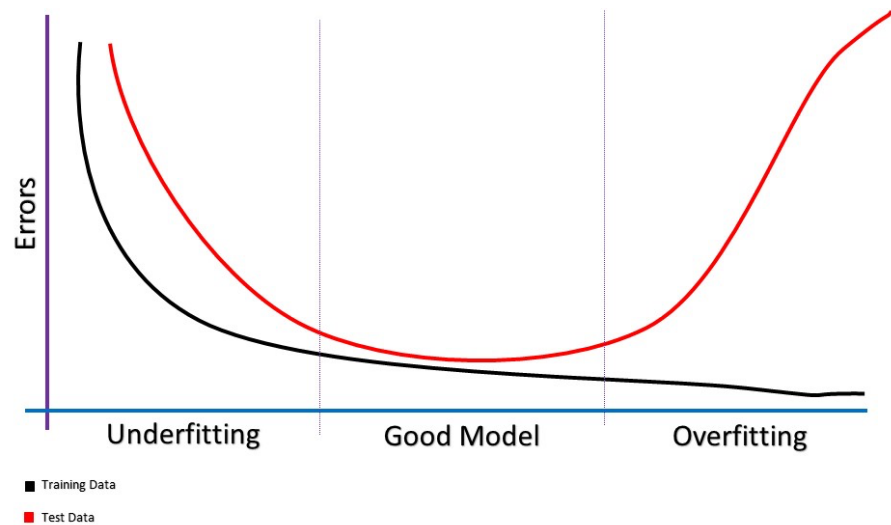
нулевая гипотеза: $H_0: \mu_1 = \mu_2$

альтернатива: $H_1: \mu_1 < \neq > \mu_2$

статистика: $T(X_1^{n_1}, X_2^{n_2}) = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}}$
$$\nu = \frac{\left(\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}\right)^2}{\frac{S_1^4}{n_1^2(n_1-1)} + \frac{S_2^4}{n_2^2(n_2-1)}}$$

нулевое распределение: $\approx St(\nu)$





	Underfitting	Just right	Overfitting
Symptoms	<ul style="list-style-type: none"> • High training error • Training error close to test error • High bias 	<ul style="list-style-type: none"> • Training error slightly lower than test error 	<ul style="list-style-type: none"> • Very low training error • Training error much lower than test error • High variance
Regression illustration			
Classification illustration			
Deep learning illustration			
Possible remedies	<ul style="list-style-type: none"> • Complexify model • Add more features • Train longer 		<ul style="list-style-type: none"> • Perform regularization • Get more data