Prezentacja Projektu I Tunowalność

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Algorytmy

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W naszym projekcie rozważamy algorytmy:

- Random Forest,
- Light GBM,
- XgBoost.

Dla nich optymalizopwałyśmy parametry:

- Random Forest: {'n_estimators', 'min_samples_split', 'min_samples_leaf', 'max_features', 'max_depth'}
- Light GBM: {num_leaves', 'n_estimators', learning_rate'},
- XgBoost: {'subsample', 'min_child_weight', 'max_depth', 'eta','colsample_bytree'}.



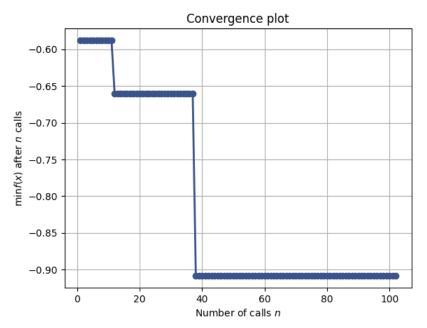
Ramki danych

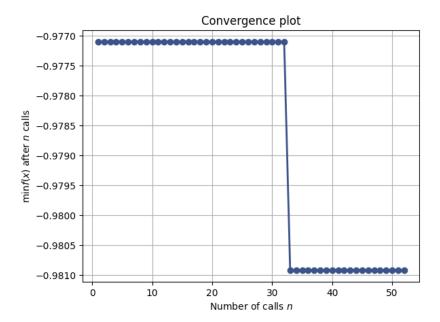
Parametry wybranych przez nas algorytmów optymalizowałyśmy na zbiorach danych:

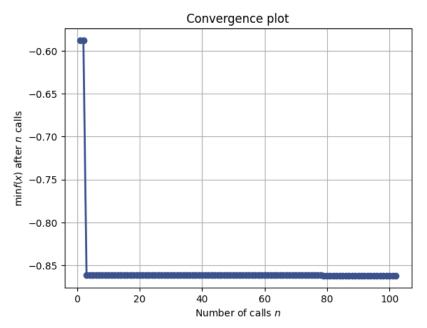
- Titanic,
- SpeedDating,
- Credit-g,
- Diabetes

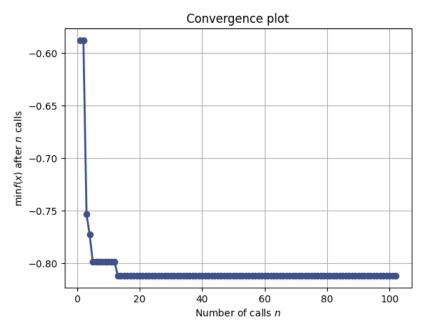
Hiperparametry oraz accuracy score					
Zbiór danych	Metoda samplingu	{'n_estimators', 'min_samples_split', 'min_samples_leaf', 'max_features', 'max_depth'}	Score	Zbieżność	
Titanic	RandomizedSearchCV	{37, 8, 1, 12, 18}	0.7354568238779	-	
	Bayes	{34, 2, 1, 13, 20}	0.9083969465649	38	
SpeedDating	RandomizedSearchCV	{107, 2, 1, 11, 15}	0.8646669337874	-	
	Bayes	{112, 18, 1, 13, 19}	0.8621718377083	3	
Credit-g	RandomizedSearchCV	{45, 4, 3, 2, 8}	0.72	-	
	Bayes	{117, 17, 1, 2, 20}	0.745	21	
Diabetes	RandomizedSearchCV	{50, 6, 7, 4, 10}	0.7720778355325	-	
	Bayes	{119, 17, 1, 2, 20}	0.8116883116883	16	

 $\label{thm:continuous} \mbox{Tabela 1: Optymalne kombinacje hiperparametrów dla zbiorów danych 1-4 korzystając z RandomizedSearchCV i Optymalizacji Bayesowskiej z algorytmem RandomForest.}$





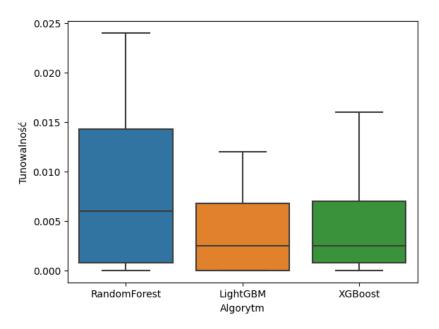




Tunowalność algorytmów

Tunowalność				
Algorytm	Miara Tunowalności	Średnia tunowalność		
RandomForest	$\begin{array}{l} d^{(1)} = R^{(1)}(\theta^*) - R^{(1)}(\theta^{(1)*}) = -0.735 + 0.735 = 0 \\ d^{(2)} = R^{(2)}(\theta^*) - R^{(2)}(\theta^{(2)*}) = -0.864 + 0.865 = 0.001 \\ d^{(3)} = R^{(3)}(\theta^*) - R^{(3)}(\theta^{(3)*}) = -0.696 + 0.72 = 0.024 \\ d^{(4)} = R^{(4)}(\theta^*) - R^{(4)}(\theta^{(4)*}) = -0.761 + 0.772 = 0.011 \end{array}$	0.00900		
LightGBM	$\begin{array}{l} d^{(1)} = R^{(1)}(\theta^*) - R^{(1)}(\theta^{(1)*}) = -0.975 + 0.975 = 0 \\ d^{(2)} = R^{(2)}(\theta^*) - R^{(2)}(\theta^{(2)*}) = -0.863 + 0.875 = 0.012 \\ d^{(3)} = R^{(3)}(\theta^*) - R^{(3)}(\theta^{(3)*}) = -0.708 + 0.708 = 0 \\ d^{(4)} = R^{(4)}(\theta^*) - R^{(4)}(\theta^{(4)*}) = -0.743 + 0.748 = 0.005 \end{array}$	0.00425		
XGBoost	$ \begin{aligned} d^{(1)} &= R^{(1)}(\theta^*) - R^{(1)}(\theta^{(1)*}) = -0.975 + 0.976 = 0.001 \\ d^{(2)} &= R^{(2)}(\theta^*) - R^{(2)}(\theta^{(2)*}) = -0.865 + 0.869 = 0.004 \\ d^{(3)} &= R^{(3)}(\theta^*) - R^{(3)}(\theta^{(3)*}) = -0.698 + 0.698 = 0 \\ d^{(4)} &= R^{(4)}(\theta^*) - R^{(4)}(\theta^{(4)*}) = -0.743 + 0.759 = 0.016 \end{aligned} $	0.00525		

Tabela 4: Tunowalność dla algorytmów.



Dziękujemy za uwagę!