

Creating Optimal Model

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Goal of Assignment

Find a model that outputs the most amount of profit

- +1.5\$ for correct prediction 0

*Predict customer wouldn't buy, and actually wouldn't have. **Correctly send customer discount.***

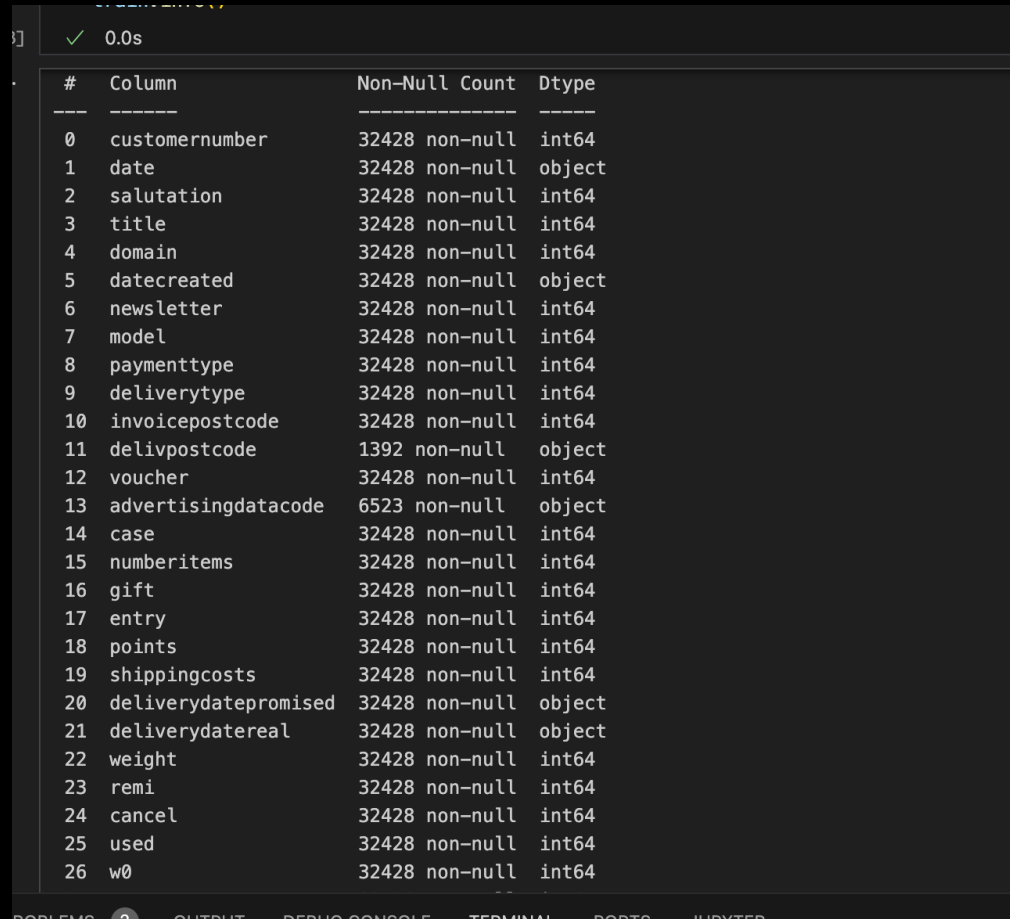
- -5\$ for false prediction 1

*Predict customer wouldn't buy, but actually would have. **Falsely send customer discount.***

Maximize Precision (0) and Recall (1)

Precision more important, as nearly 5x more instances of prediction 0.

Preprocessing



A screenshot of a Jupyter Notebook interface. At the top, a status bar shows a green checkmark and '0.0s'. Below it, a pandas DataFrame summary is displayed. The summary has four columns: '#', 'Column', 'Non-Null Count', and 'Dtype'. It lists 27 features (rows 0 to 26). Most features have a count of 32428 and are marked as 'non-null'. The 'Dtype' column shows various data types including 'int64', 'object', and 'real'. The bottom of the image shows the Jupyter Notebook tabs: 'PROBLEMS', 'OUTPUT', 'DEBUG CONSOLE', 'TERMINAL', 'PORTS', and 'JUPYTER'.

#	Column	Non-Null Count	Dtype
0	customernumber	32428 non-null	int64
1	date	32428 non-null	object
2	salutation	32428 non-null	int64
3	title	32428 non-null	int64
4	domain	32428 non-null	int64
5	datecreated	32428 non-null	object
6	newsletter	32428 non-null	int64
7	model	32428 non-null	int64
8	paymenttype	32428 non-null	int64
9	deliverytype	32428 non-null	int64
10	invoicepostcode	32428 non-null	int64
11	delivpostcode	1392 non-null	object
12	voucher	32428 non-null	int64
13	advertisingdatacode	6523 non-null	object
14	case	32428 non-null	int64
15	numberitems	32428 non-null	int64
16	gift	32428 non-null	int64
17	entry	32428 non-null	int64
18	points	32428 non-null	int64
19	shippingcosts	32428 non-null	int64
20	deliverydatepromised	32428 non-null	object
21	deliverydatereal	32428 non-null	object
22	weight	32428 non-null	int64
23	remi	32428 non-null	int64
24	cancel	32428 non-null	int64
25	used	32428 non-null	int64
26	w0	32428 non-null	int64

- Drop features for various reasons (too few values, unmeaningful, no predictiveness, data type, ...)
- Created dummy variables
- Splitting training / test

Different Models

Classifiers for all

- Cross-Validation on many parameters for: xgBoost / RandomForest / KNN
- SVC
- Checked classification report on all variations
- Focus: Maximize Precision (0) and Recall (1)

The best model

- RandomForestClassifier
- Class_weight='balanced'
- max_depth=9, min_samples_split=12, min_samples_leaf=17, max_features=5, n_estimators=37, random_state=71

Without Class_weight='balanced'

	precision	recall	f1-score	support
0	0.82	1.00	0.90	5292
1	0.00	0.00	0.00	1194
accuracy			0.82	6486
macro avg	0.41	0.50	0.45	6486
weighted avg	0.67	0.82	0.73	6486

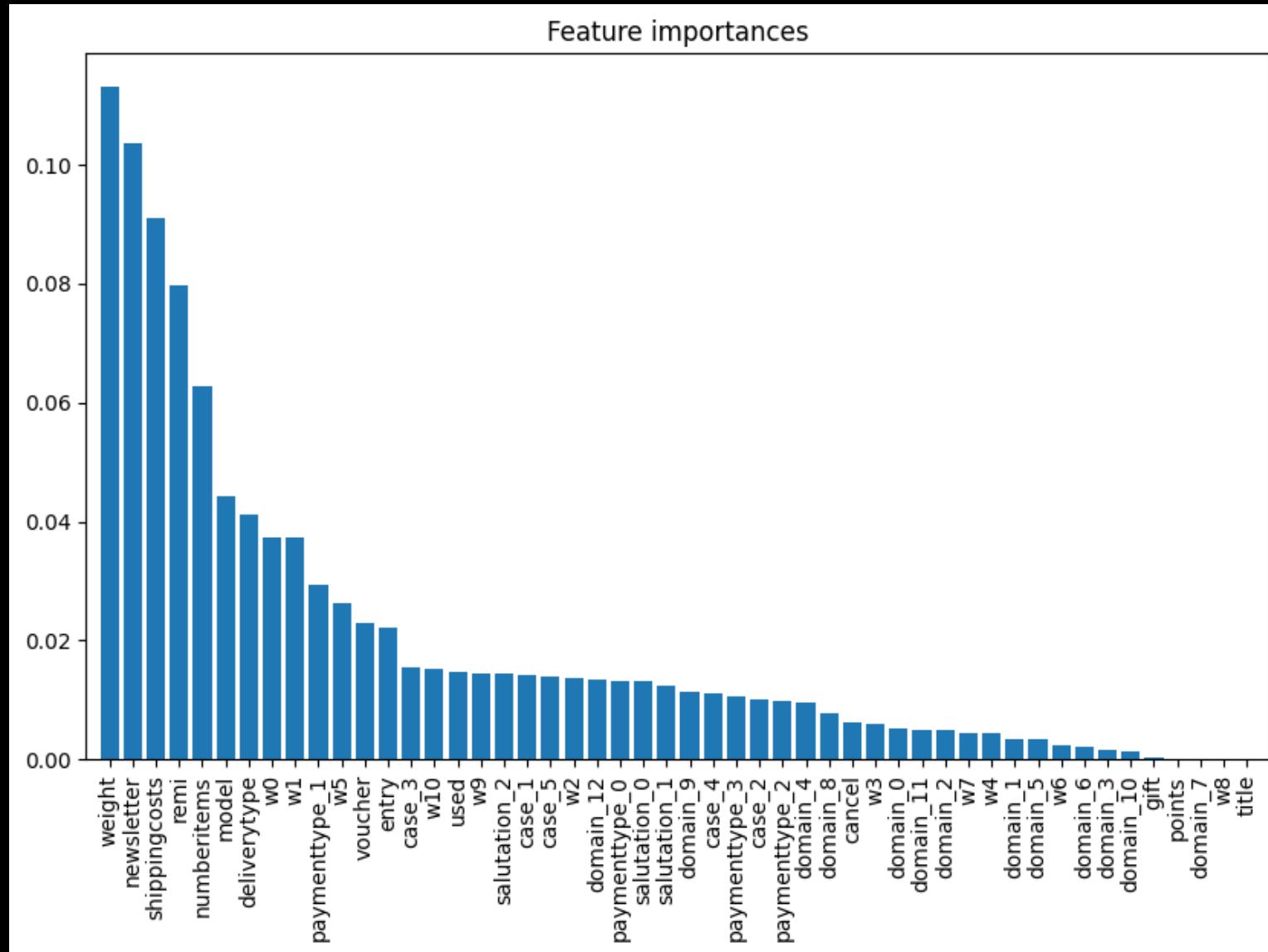
With Class_weight='balanced'

	precision	recall	f1-score	support
0	0.86	0.61	0.72	5256
1	0.26	0.59	0.36	1230
accuracy			0.61	6486
macro avg	0.56	0.60	0.54	6486
weighted avg	0.75	0.61	0.65	6486

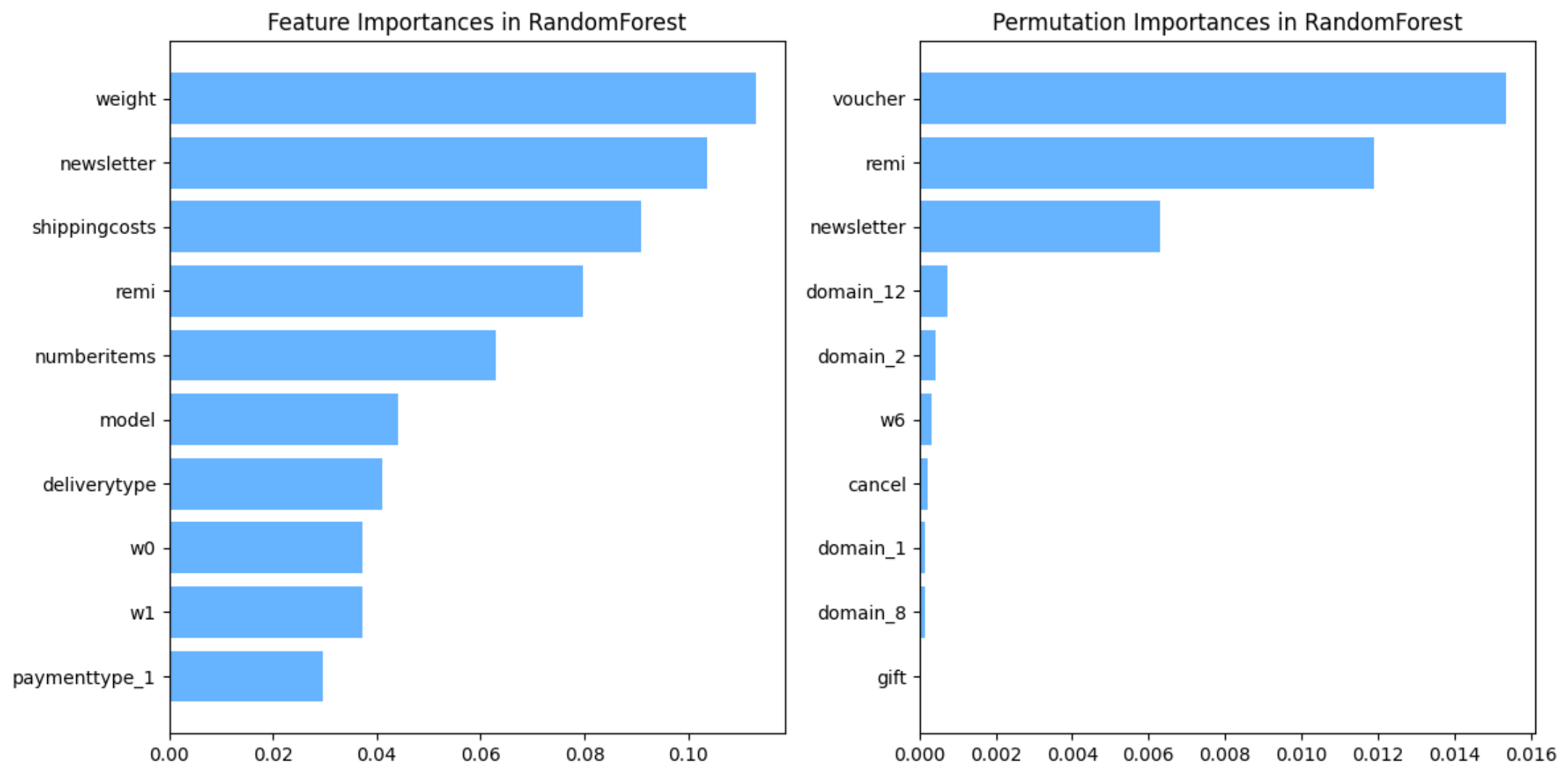
Model: xgboost

	precision	recall	f1-score	support
0	0.85	0.62	0.72	5256
1	0.25	0.53	0.34	1230
accuracy			0.61	6486
macro avg	0.55	0.58	0.53	6486
weighted avg	0.74	0.61	0.65	6486

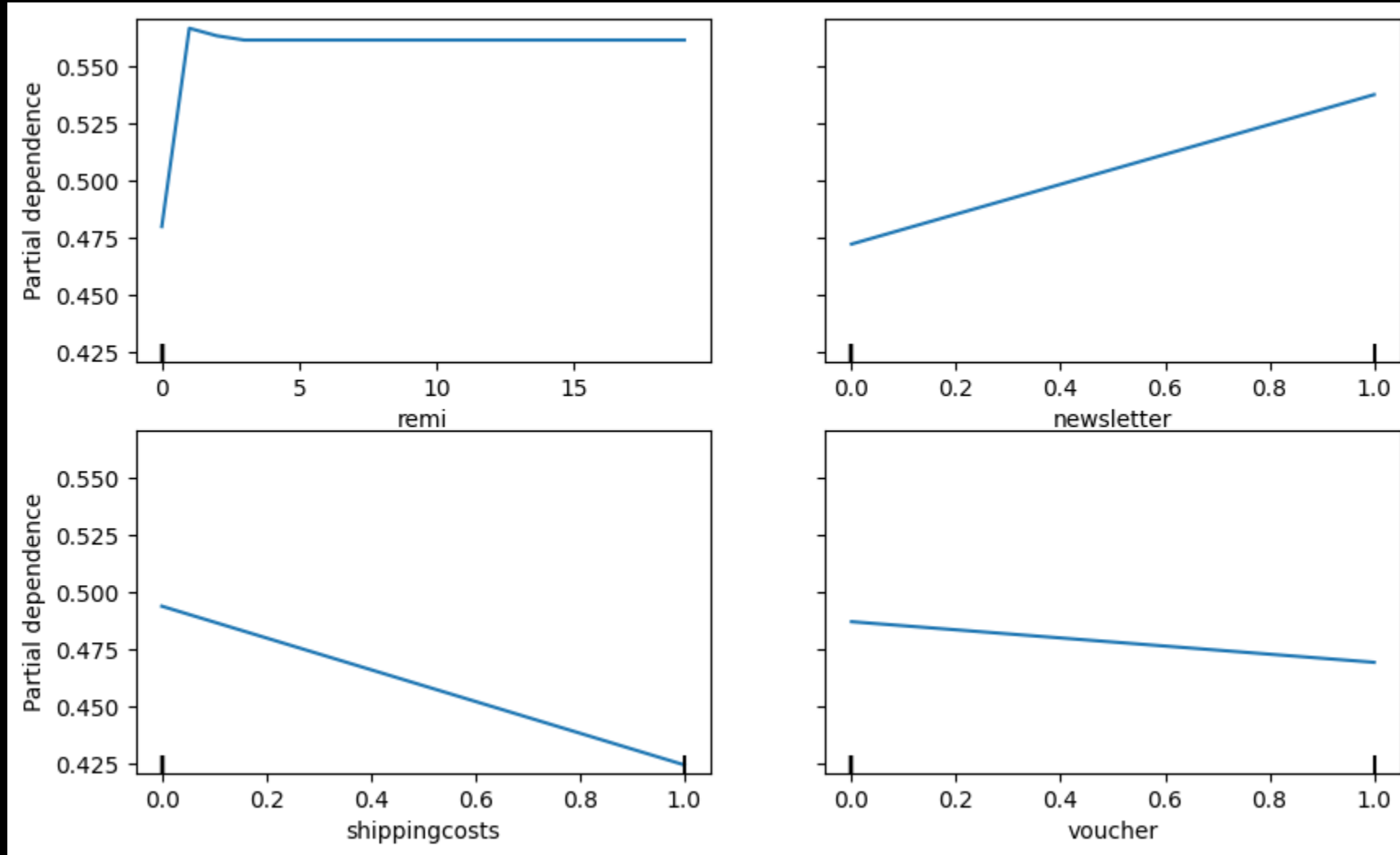
Global Interpretation: Feature importances



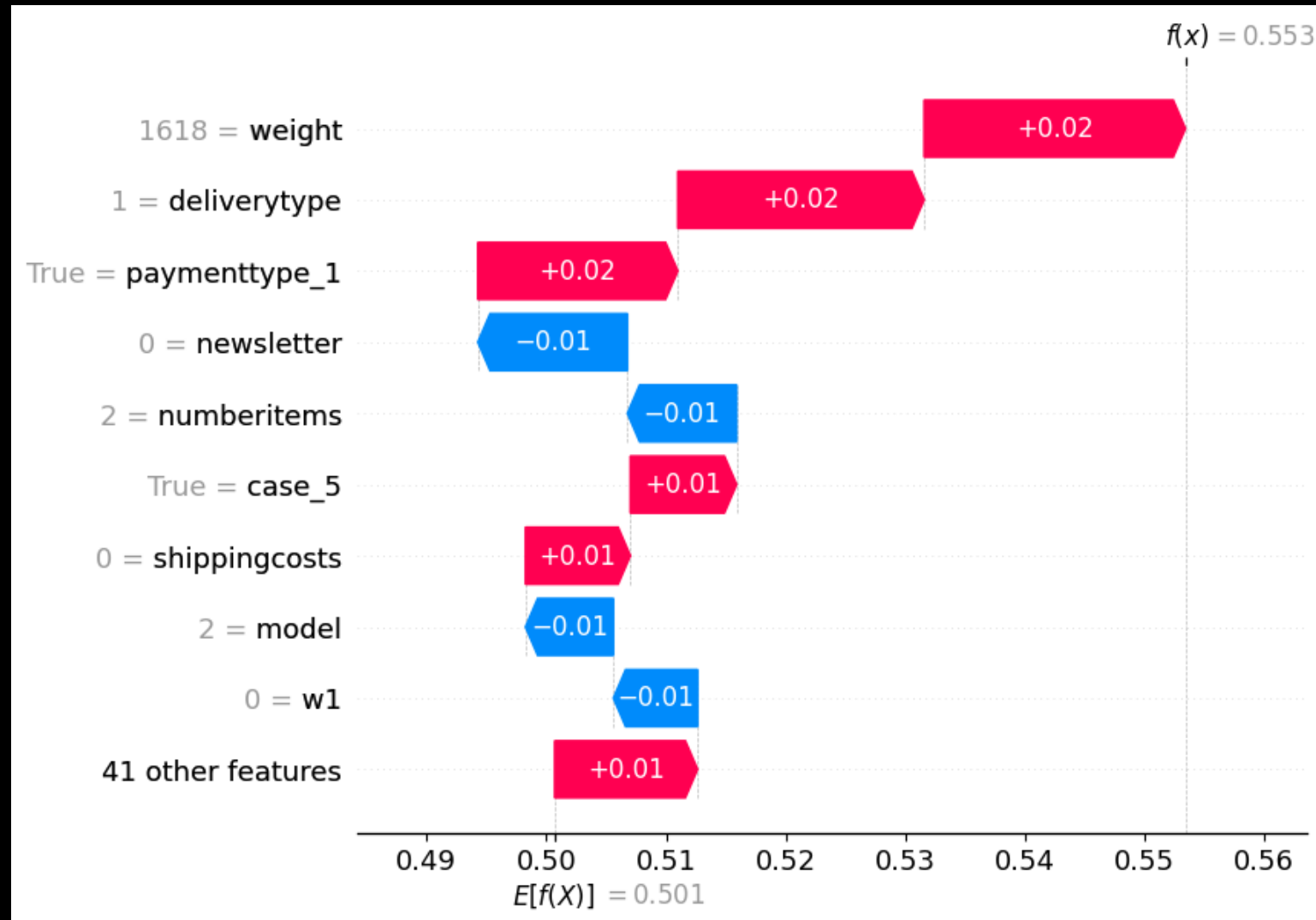
Global Interpretation: Permutation Importances



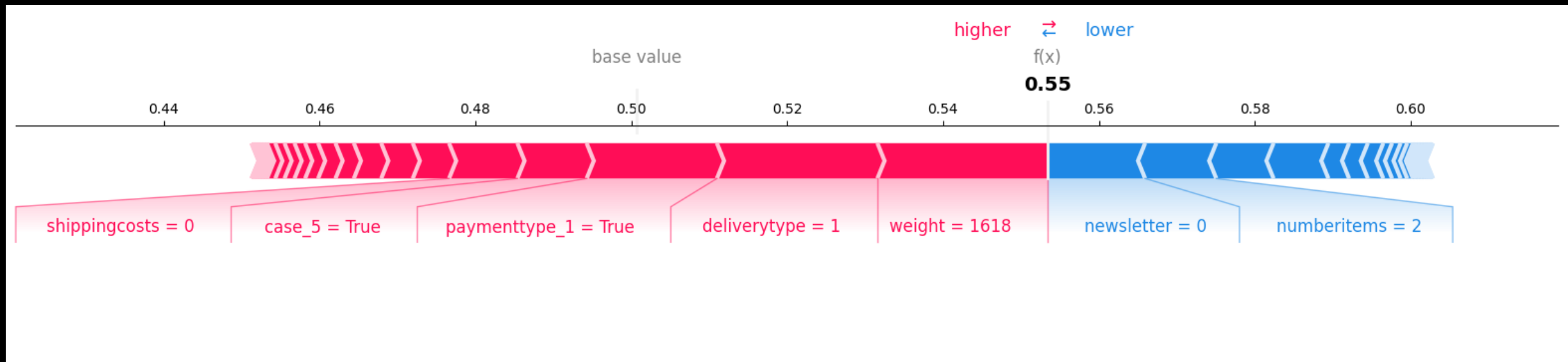
Global Interpretation: Partial Dependence Plots



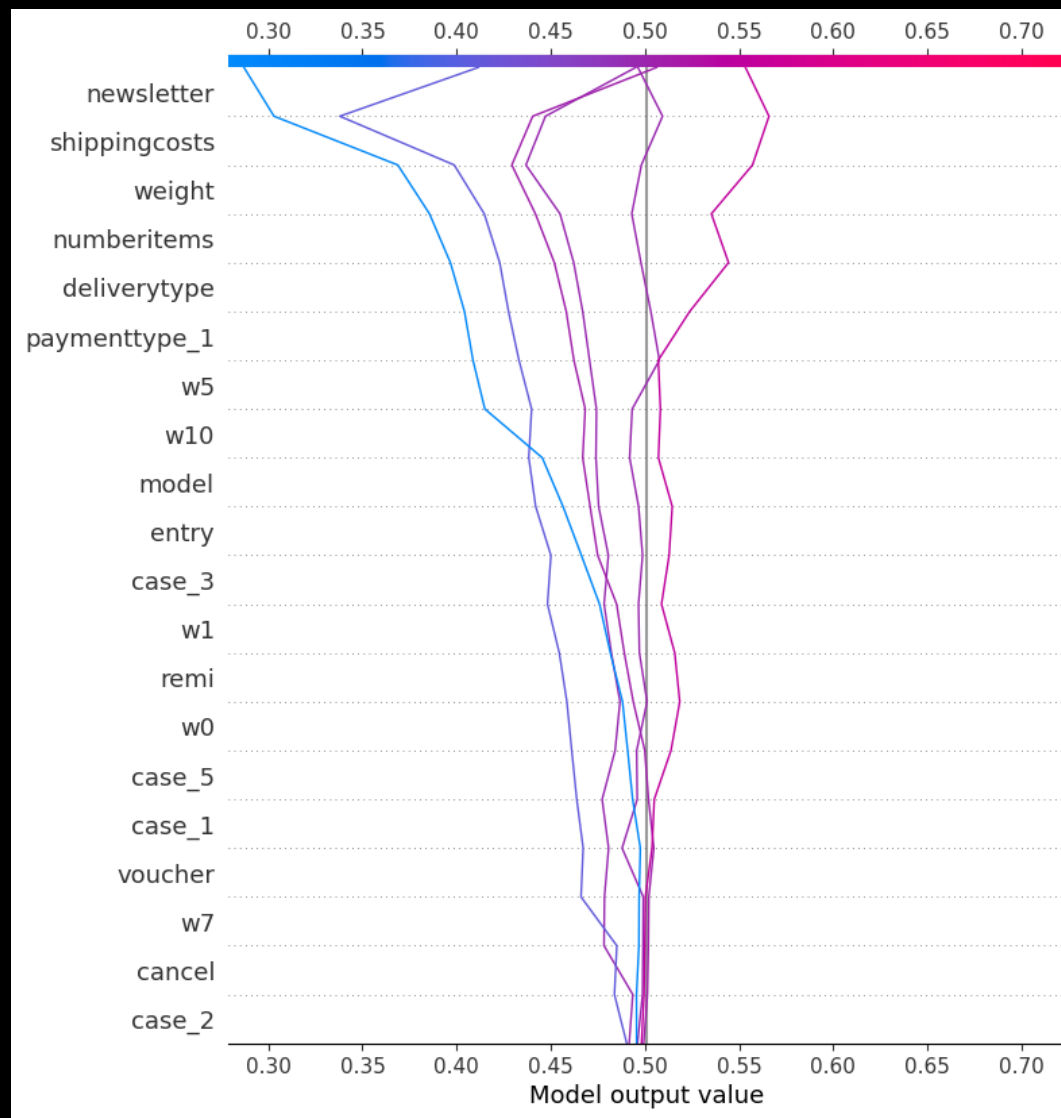
Local interpretation: Waterfall plot of SHAP values for the second row of data points



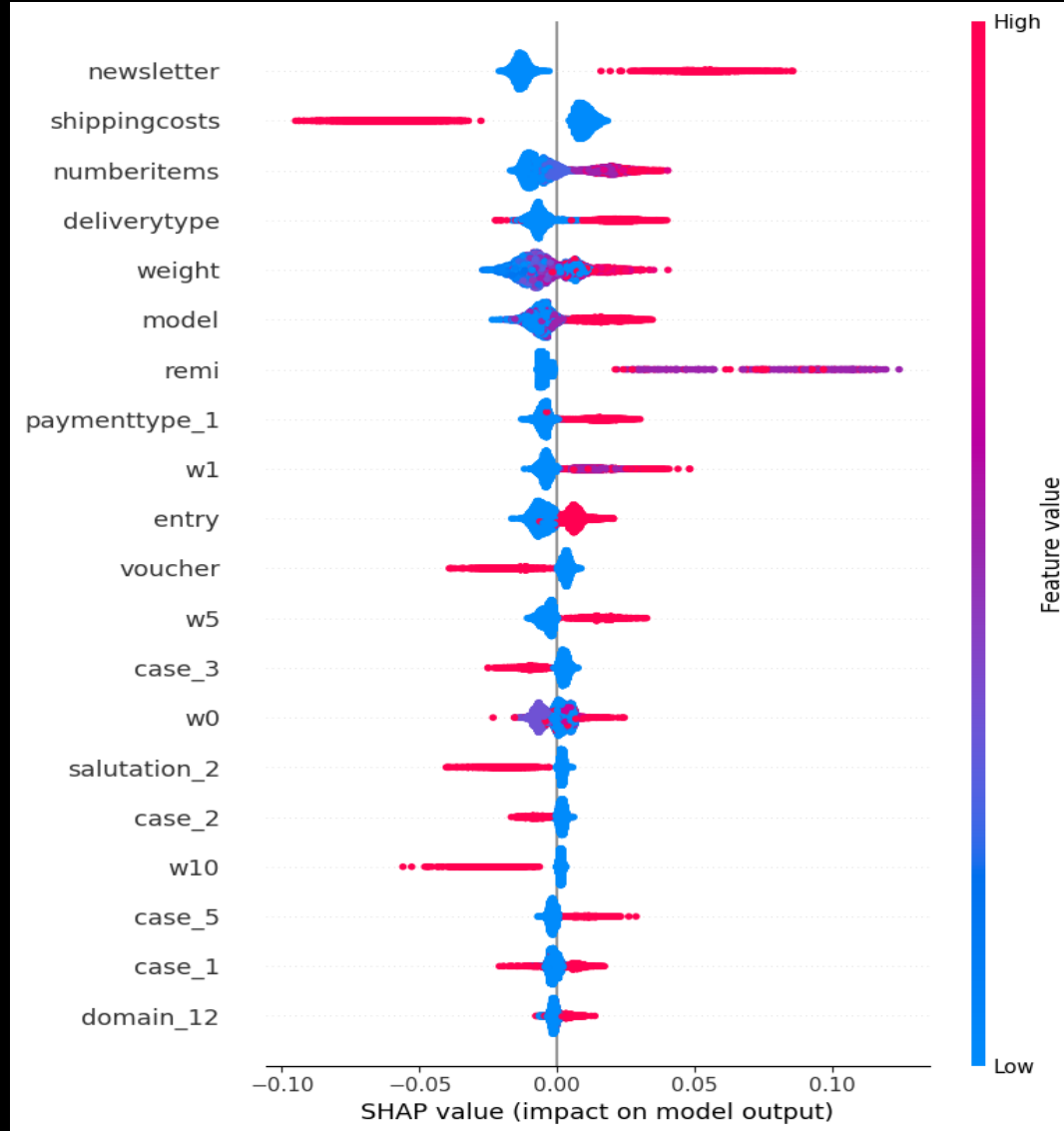
Local interpretation: SHAP power diagram for data point with index 18518



Local interpretation



Global and Local Interpretation



Conclusion and Profit

- Calculation:

$$(\text{Actual_amount_of_0s} * \text{Precision_0} * 1.5) + \\ (\text{Actual_amount_of_1s} * (1 - \text{Recall_1}) * (-5))$$

$$= \underline{\mathbf{4289.01\$}} (y_{\text{test}})$$