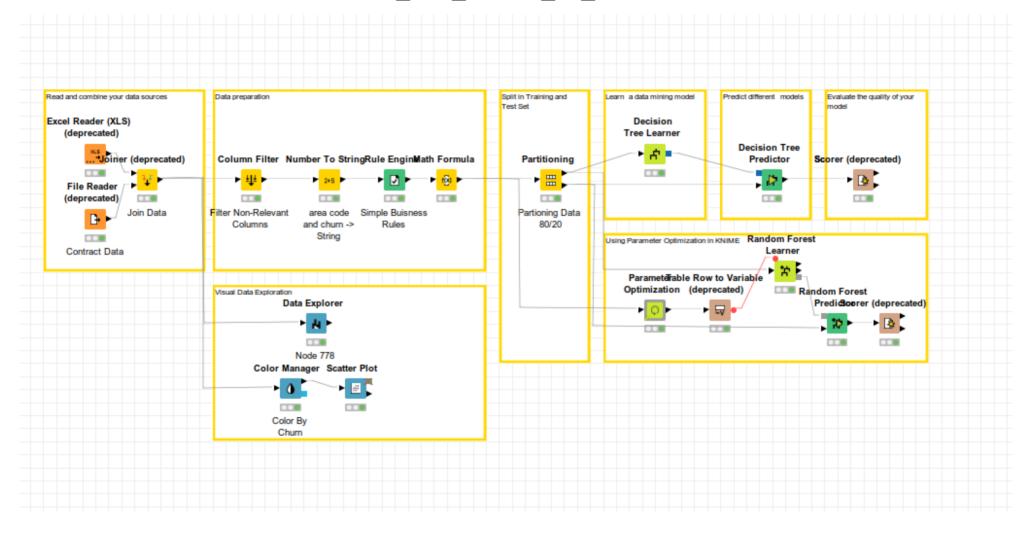
Федеральное государственное бюджетное образовательное учреждение высшего образования «МИРЭА – Российский Технологический Университет»

Отчёт по практической работе №5: «Реализация алгоритма для прогнозирования оттока клиентов»

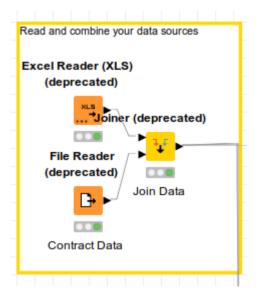
Подготовила студентка группы ББМО-01-21 Концевая Юлия

- ▼ ♠ EXAMPLES (knime@api.hub.knime.com)
 - ▶ 00_Components
 - ▶ 101_Data_Access
 - ▶ 1 02_ETL_Data_Manipulation
 - ▶ 103 Visualization
 - **▼** 04 Analytics
 - ▶ 01_Preprocessing
 - ▶ 02 Statistics
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 - △01_Example_for_Learning_a_Decision_Tree
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DataScience_with_KNIME_an_Introduction



Первый набор данных

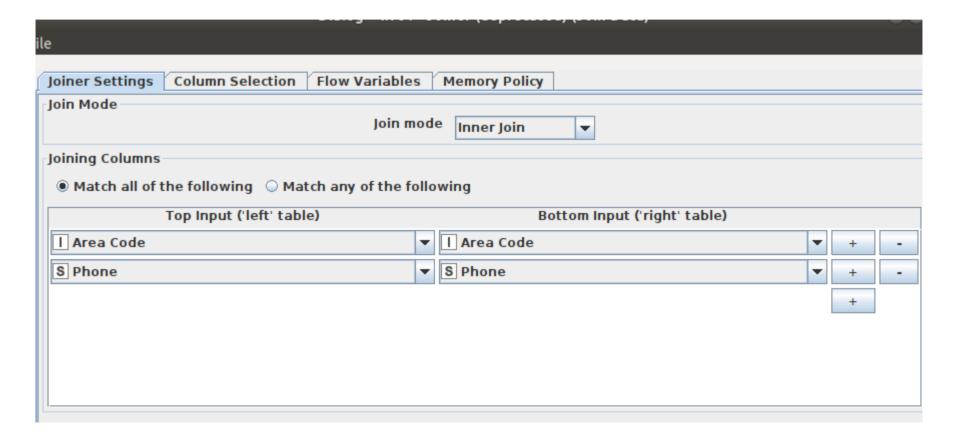


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Row0					10		110									382-4657
Rowl					13.7	1	123	27.47								371-7191
Row2						0	114	41.38								358-1921
Row3						2	71	50.9				7102				375-9999
Row4	_						113	28.34								330-6626
Row5							98	37.98								391-8027
Row6	_					3	88					9.57	_			355-9993
Row7						0	79									329-9001
Row8					8.7	1	97	31.37				9.71				335-4719
Row9	37	258.6	222	326.4	11.2	0	84	43.96	111	18.87	97	14.69	5	3.02	415	330-8173
Row10	0	129.1	228.5	208.8	12.7	4	137	21.95	83	19.42	111	9.4	6	3.43	415	329-6603
Rowll	0	187.7	163.4			0	127	31.91	148	13.89	94	8.82	5	2.46	415	344-9403
Row12	0	128.8	104.9	141.1	11.2	1	96	21.9	71	8.92	128	6.35	2	3.02	408	363-1107
Row13	0	156.6	247.6	192.3	12.3	3	88	26.62	75	21.05	115	8.65	5	3.32	510	394-8006
Rowl 4	0	120.7	307.2	203	13.1	4	70	20.52	76	26.11	99	9.14	6	3.54	415	366-9238
Row15	0	332.9	317.8	160.6	5.4	4	67	56.59	97	27.01	128	7.23	9	1.46	415	351-7269
Rowl6	27	196.4	280.9	89.3	13.8	1	139	33.39	90	23.88	75	4.02	4	3.73	408	350-8884
Row17	0	190.7	218.2	129.6	8.1	3	114	32.42	111	18.55	121	5.83	3	2.19	510	386-2923
Row18	33	189.7	212.8	165.7	10	1	66	32.25	65	18.09	108	7.46	5	2.7	510	356-2992
Rowl9	0	224.4	159.5	192.8	13	1	90	38.15		13.56	74	8.68	2	3.51	415	373-2782
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Row22	0	183	72.9	181.8	9.5	0	112	31.11	99	6.2	78	8.18	19	2.57	415	358-1958
Row23	0				7.7	2	103									350-2565
D24		01.1	245.2	227	10.2	_	0.0	10.70	70	20.04	115	10.07		2.70	E10	242 4606

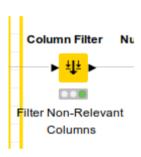
Второй набор данных

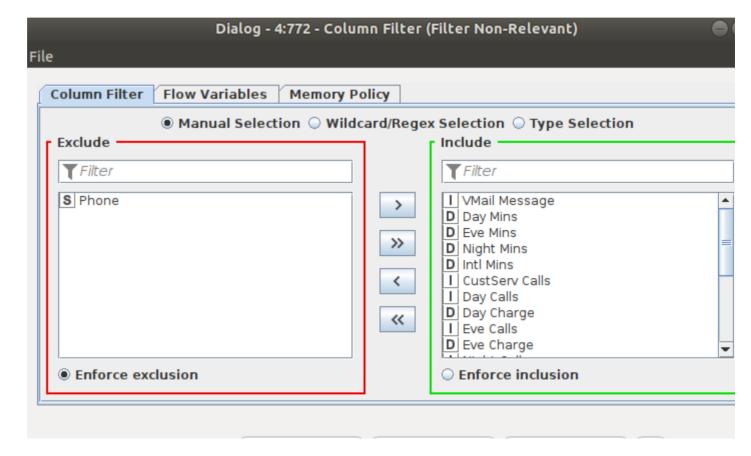
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Row2	137	0	0	0	NJ	415	358-1921
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Row4	75	0	1	0	OK	415	330-6626
Row5	118	0	1	0	AL	510	391-8027
Row6	121	0	0	1	MA	510	355-9993
Row7	147	0	1	0	МО	415	329-9001
Row8	117	0	0	0	LA	408	335-4719
Row9	141	0	1	1	WV	415	330-8173
Rowl0	65	1	0	0	IN	415	329-6603
Rowll	74	0	0	0	RI	415	344-9403
Rowl2	168	0	0	0	IA	408	363-1107
Row13	95	0	0	0	MT	510	394-8006
Rowl4	62	0	0	0	IA	415	366-9238
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Rowl6	85	0	0	1	ID	408	350-8884
Rowl7	93	0	0	0	VT	510	386-2923
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Row19	73	0	0	0	TX	415	373-2782
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Row21	77	1	0	0	СО	408	393-7984
Row22	130	0	0	0	AZ	415	358-1958
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D24	122	0	^	^		F10	242 4000

Объединение данных

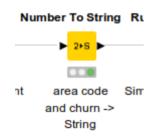


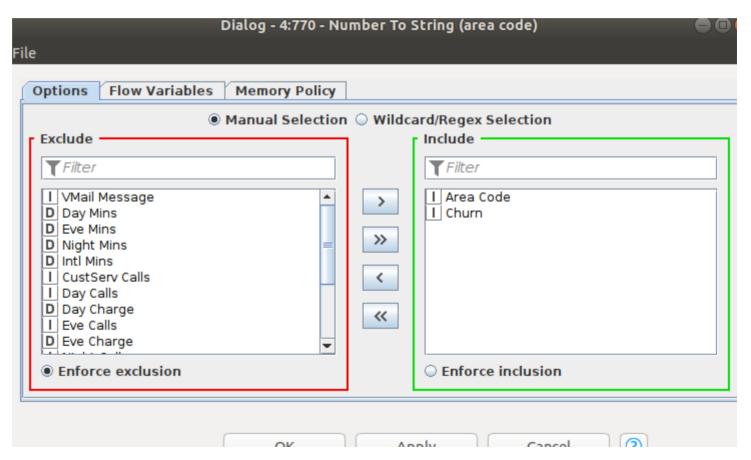
Фильтрация ненужных значений



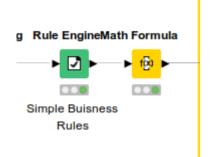


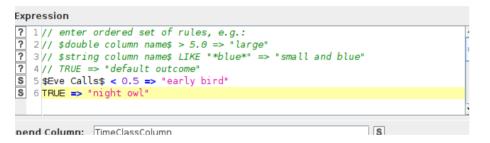
Преобразование типов данных





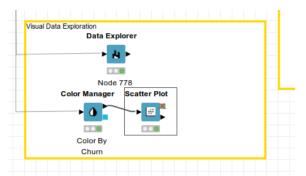
Предварительная обработка данных



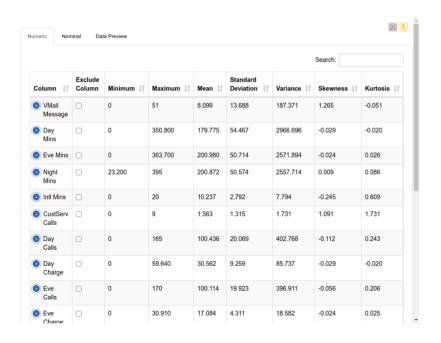


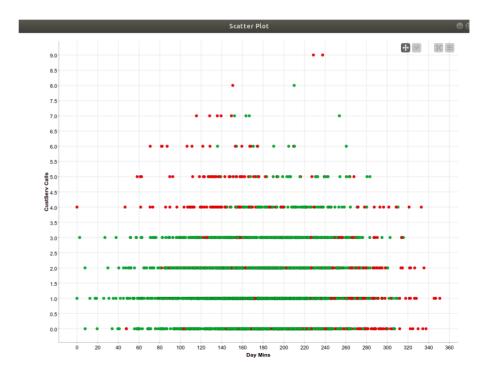


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0	0	1	OH	611.5	night owl
0	0	0	NJ	527.2	night owl
0	1	0	ОН	558.2	night owl
0	1	0	OK	501.9	night owl
0	1	0	AL	647.9	night owl
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0	0	0	LA	751.9	night owl
0	1	1	WV	807	night owl
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0	0	0	RI	547.1	night owl
0	0	0	IA	374.8	night owl
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0	0	0	VT	538.5	night owl
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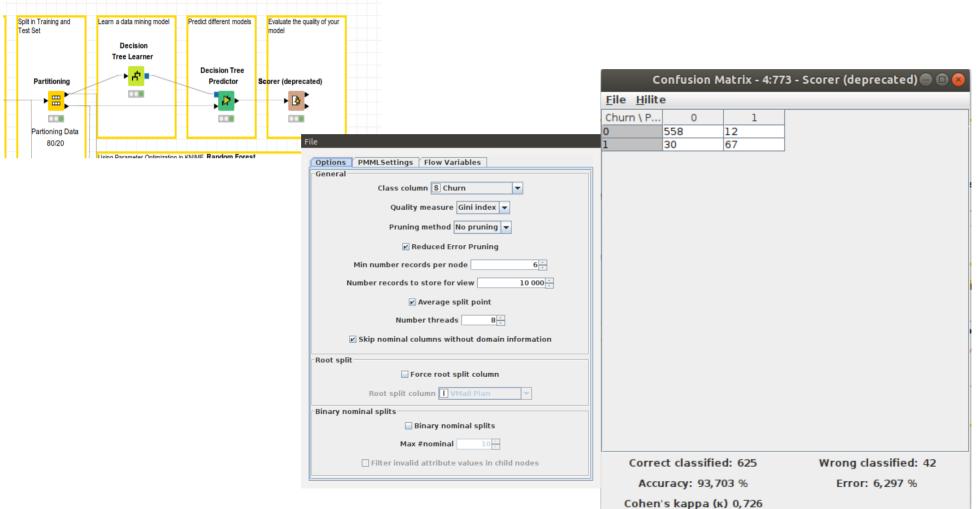


Предварительная оценка данных

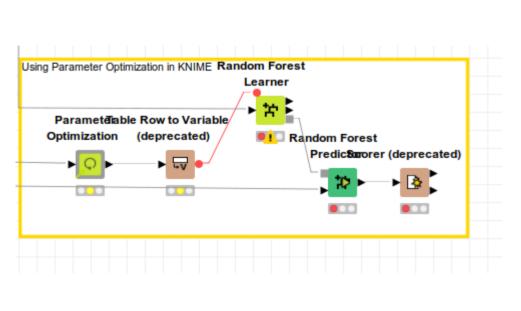




Построение модели и оценка точности



Оптимизация гиперпараметров



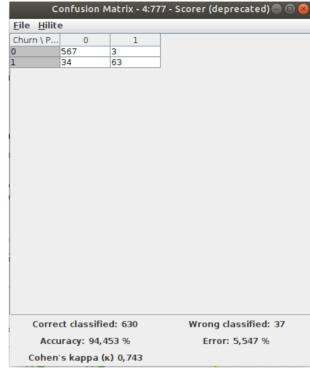


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Спасибо за внимание!