Use of machine-learning algorithms to predict urine culture results in cats

Materials and methods

Feature preprocessing

Patient data, including sex, age, color, clarity, USG, and ph, were utilized as features. Numeric features: protein, glucose, ketones, hemoprotein, bilirubin, WBC, and RBC were accounted for to target positive culture. Ignore_features (visit number; BW) were excluded from training as its values contain NaN values and strings. A positive culture was encoded using the one-hot encoding (0 and 1). Cat sex was splitted into 4 different groups.

Machine-learning model training

Two Feline UTI prediction models were trained. The ExtraTreesClassifier (ET) model was trained with all patient features excluding Sed Bac Quantity (SBQ) and Sed Bac Type (SBT). The second model, ExtraTreesClassifier (ET) trained with excluded features SBQ; SBT. The first model excluded SBQ and SBT due to limited resources to obtain data from the veterinary lab. The second model was trained to analyze AUC performance by just training the excluded data.

	Accuracy	AUC	Recall	Prec.	F1	Карра	мсс		Accuracy	AUC	Recall	Prec.	F1	Kappa	M
Fold								Fold							
0	0.8662	0.7313	0.3846	0.7692	0.5128	0.4451	0.4811	0	0.8908	0.7987	0.5192	0.8182	0.6353	0.5749	0.59
1	0.8944	0.8335	0.5577	0.8056	0.6591	0.5990	0.6132	1	0.9014	0.8077	0.6346	0.7857	0.7021	0.6439	0.64
2	0.9085	0.8433	0.6346	0.8250	0.7174	0.6639	0.6720	2	0.8873	0.8375	0.5769	0.7500	0.6522	0.5863	0.59
3	0.8768	0.8497	0.5192	0.7297	0.6067	0.5361	0.5470	3	0.8838	0.8003	0.5769	0.7317	0.6452	0.5768	0.58
4	0.8662	0.7933	0.4615	0.7059	0.5581	0.4833	0.4985	4	0.8662	0.8278	0.5577	0.6591	0.6042	0.5243	0.52
5	0.8627	0.8236	0.4423	0.6970	0.5412	0.4651	0.4818	5	0.8944	0.8579	0.6346	0.7500	0.6875	0.6245	0.62
6	0.8905	0.8100	0.5490	0.7778	0.6437	0.5812	0.5935	6	0.8410	0.7543	0.4706	0.5714	0.5161	0.4221	0.42
7	0.8834	0.7787	0.5192	0.7714	0.6207	0.5549	0.5701	7	0.8657	0.7472	0.4231	0.7333	0.5366	0.4646	0.48
8	0.8587	0.7327	0.3654	0.7308	0.4872	0.4156	0.4493	8	0.8481	0.7878	0.4808	0.6098	0.5376	0.4482	0.45
9	0.8905	0.8524	0.5769	0.7692	0.6593	0.5957	0.6044	9	0.8516	0.7454	0.4038	0.6562	0.5000	0.4186	0.43
Mean	0.8798	0.8048	0.5011	0.7582	0.6006	0.5340	0.5511	Mean	0.8730	0.7965	0.5278	0.7065	0.6017	0.5284	0.53
Std	0.0155	0.0429	0.0817	0.0394	0.0698	0.0754	0.0681	Std	0.0203	0.0367	0.0777	0.0750	0.0700	0.0802	0.07
First model (Data_limted)								S	Second	mode	el (SBC	Q_SBT	Γ)		

Figure 1. Data table illustrating AUC performances between two different models.

The ET model (Fig.1) utilized the training set data, allowing us to determine how numerical features affect the prediction performance for Feline UTI. The model trained using a repeated 10-folds for each of the 10 candidates, totaling 100 times to produce the average confusion matrix to tune the hyperparameters, using the built-in function:

Tuned et = tune model(et) (Python3; ipykernel).

The boundary plot (Fig. 2) visualizes how plots spread for trained set data. Through visualization, we can analyze that all trained data sets were spread out for the first model (Data_limited), and the decision boundary will be in the negative linear slope form. While the second model (SBQ, SBT) reflects skewed trained data sets on the right-hand side of the plot, it will include the decision boundary in the form of a curve hyper-plane to fit the dimension of the space.

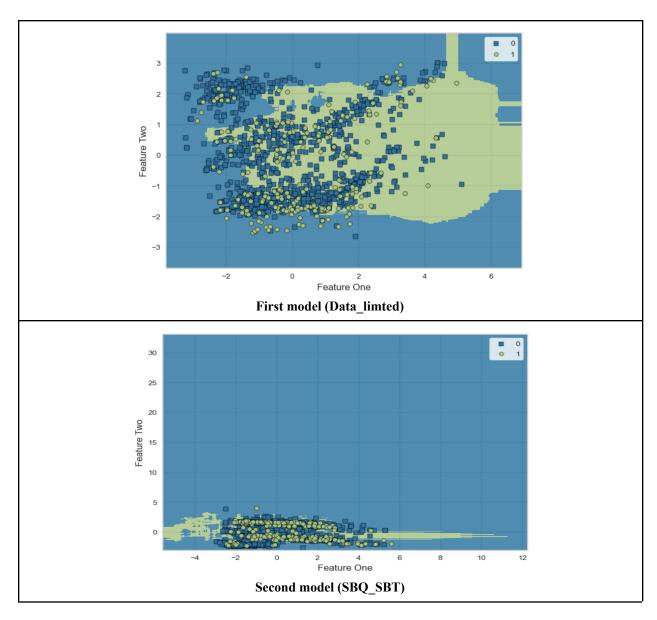


Figure 2. Data table illustrating AUC performances between two different models.

Machine-learning model performance evaluation

The test set included [types of cats]. The hyperparameters optimized on the training set utilized during ET (with all features excluding sed features) and ET (with only sed features) model testing. Model prediction results reported sensitivity and specificity. Predictions made utilized as a binary classifier, a receiver operating characteristic (ROC) plot was generated, and the area under the curve (AUC) was calculated.

Results

During the study period, 6,732 unique cats were evaluated. Of these cats, 5,850 were tested for Feline UTI prediction: 4,634 in the training set, and 1,216 in the test set. Under the category of "Positive culture", there were a total of 4,052. Between collected data, 3,310 of the training sets were under 0 while 742 remaining data sets were set to 1; 0 meaning negative and 1 meaning positive.

The (Fig. 3) displays the summary of collected statistics for the training and test set. Numeric variables (mean, standard deviation): protein (67.129, 101.355), glucose (88.968, 265.289), ketones (0.513, 5.761), hemoprotein (136.512, 113.445), bilirubin (0.178, 0.821), WBC (11.320, 23.903), RBC (39.359, 43.832). Categorical variables resulted in the count of 4052.

The data was measured using the built-in function:

.describe() (Python3; ipykernel).

describe - 0 and 1 / two data tables for comparison

P	ositive culture	sex	Age	BW	USG	ph	Protein	Glucose	Ketones	Bilirubin	Hemoprotein	WBC	RB
ount	4052.000000	4052.000000	4052.000000	3847.000000	4052.000000	4052.000000	4052.000000	4052.000000	4052.000000	4052.000000	4052.000000	4052.000000	4052.00000
nean	0.183119	2.908934	10.279640	4.686750	1.024500	6.310711	67.128578	88.968411	0.513327	0.178677	136.511599	11.320089	39.35908
std	0.386812	1.100862	5.267742	1.625702	0.147864	0.838740	101.355252	265.288806	5.760639	0.821015	113.445203	23.902521	43.83219
min	0.000000	-9.000000	0.000000	0.000800	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.00000
25%	0.000000	2.000000	6.100000	3.600000	1.012000	6.000000	25.000000	0.000000	0.000000	0.000000	10.000000	3.000000	2.00000
50%	0.000000	2.000000	10.900000	4.500000	1.017000	6.000000	25.000000	0.000000	0.000000	0.000000	150.000000	3.000000	12.00000
75%	0.000000	4.000000	14.400000	5.600000	1.031000	7.000000	75.000000	0.000000	0.000000	0.000000	250.000000	5.000000	100.0000
max	1.000000	4.000000	24.600000	34.000000	Positiv	e culture Se	(Data_li	y Sed Bac Ty		6.000000	250.000000	100.000000	100.0000
max	1.000000	4.000000	24.600000	34.000000	Firs	t model	(Data_li	mted)		6.000000	250.000000	100.000000	100.0000
max	1.000000	4.000000	24.600000	c	Positiv	t model	(Data_li	(mted) y Sed Bac Ty 0 4052.0000	'pe 00	6.000000	250.000000	100.000000	100.0000
max	1.000000	4.000000	24.600000	c	Positiv	t model e culture See 2.000000 0.183119	(Data_li	w Sed Bac Ty 0 4052.0000 3 0.2981	rpe 1000	6.000000	250.000000	100.000000	100.0000
max	1.000000	4.00000	24.600000	c	Positivount 405	t model e culture Sec 2.000000 0.183119 0.386812	(Data_li d Bac Quantity 4052.000000 0.381293 0.925768	(mted) y Sed Bac Ty 0 4052.0000 3 0.2981 9 0.7531	7 pe 1000 124	6.00000	250.000000	100.00000	100.0000
max	1.000000	4.00000	24.600000	c n	Positiv ount 405:	t model e culture Sec 2.000000 0.183119 0.386812 0.000000	(Data_li d Bac Quantity 4052.000000 0.381293 0.925768 0.000000	sed Bac Ty 0 4052.0000 3 0.2981 9 0.7531 0 0.0000	100 124 150	6.00000	250.000000	100.00000	100.0000
max	1.000000	4.000000	24.600000	c n	Positiv ount 405:	t model e culture Sec 2.000000 0.183119 0.386812	(Data_li d Bac Quantity 4052.000000 0.381293 0.925768	sed Bac Ty 0 4052.0000 3 0.2981 9 0.7531 0 0.0000	100 124 150	6.00000	250.000000	100.00000	100.0000
max	1.000000	4.000000	24.600000	e n	Positiv Positiv 405: nean std min 225%	t model e culture Sec 2.000000 0.183119 0.386812 0.000000	(Data_li d Bac Quantity 4052.000000 0.381293 0.925768 0.000000	y Sed Bac Ty 0 4052.0000 3 0.2981 0 0.0000 0 0.0000	rpe 000 124 150 000	6.00000	250.000000	100.00000	100.0000
max	1.000000	4.000000	24.600000	c	Positivo unt 405: min 6025% 650% 660	t model e culture Sec. 2.000000 0.183119 0.386812 0.000000	(Data_li d Bac Quantity 4052.000000 0.381293 0.925768 0.000000 0.0000000	y Sed Bac Ty 0 4052.0000 8 0.2981 9 0.7531 0.0000 0 0.0000	100 100 124 150 100 100	6.00000	250.000000	100.00000	100.0000

Figure 3. Data table describing the summary of statistics.

Machine-learning model performance

The ET model (Fig. 1) displays an AUC performance of 80.48% (first model) compared to the second ET model 79.65%. The ET first model showed slightly higher accuracy than the second model by 0.68%. The first ET model performed better on AUC, Accuracy than the second model of ET.