| | Decision Tree: Attack Detection |
|-------------------|---|
| Parameters to | Traditional Machine learning model accuracy: 99.8434% |
| choose: | Running Times seconds: 8524, FPR: 10.04% |
| $\max depth =$ | FedAvg accuracy: 99.8527%, Running Times seconds: 154.0514, FPR: 9.5726% |
| 5, 10, 15, 20, | First round: |
| 30, 40, 50, 100, | IoT device 0 Best Accuracy: 0.998310 FPR: 0.055913 |
| 150, 200 | using 'max depth': 40, 'min samples leaf': 3, 'min samples split': 5 |
| min samples | IoT device 1 Best Accuracy: 0.997699 FPR: 0.085724 |
| split = 0.1, | using 'max depth': 10, 'min samples leaf': 4, 'min samples split': 5 |
| 0.2, 0.3, 0.4, | IoT device 2 Best Accuracy: 0.999689 FPR: 0.028007 |
| 0.5, 0.6, 0.7, | using 'max depth': 10, 'min samples leaf': 1, 'min samples split': 0.1 |
| 0.8, 0.9, 1.0, 2, | IoT device 3 Best Accuracy: 0.999099 FPR: 0.083797 |
| 3, 4, 5, 6, 7 | using 'max depth': 100, 'min samples leaf': 3, 'min samples split': 3 |
| min samples | IoT device 4 Best Accuracy: 0.999220 FPR: 0.073456 |
| leaf = 1, 2, 3, | using 'max depth': 50, 'min samples leaf': 1, 'min samples split': 7 |
| 4, 5 | IoT device 5 Best Accuracy: 0.999882 FPR: 0.083333 |
| | using 'max depth': 5, 'min samples leaf': 1, 'min samples split': 3 |
| | IoT device 6 Best Accuracy: 0.999091 FPR: 0.033563 |
| | using 'max depth': 10, 'min samples leaf': 1, 'min samples split': 7 |
| | IoT device 7 Best Accuracy: 0.999201 FPR: 0.021190 |
| | using 'max depth': 5, 'min samples leaf': 1, 'min samples split': 3 |
| | IoT device 8 Best Accuracy: 0.998341 FPR: 0.139175 |
| | using 'max depth': 150, 'min samples leaf': 2, 'min samples split': 7 IoT device 9 Best Accuracy: 0.998414 FPR: 0.060517 |
| | using 'max depth': 10, 'min samples leaf': 4, 'min samples split': 7 |
| | Second round: |
| | Aggregate updates with mean parameters: 'max depth': 39, 'min samples leaf': 2, |
| | 'min samples split': 5 IoT Device 0 Accuracy: 0.997822 FPR: 0.099087 IoT Device 1 Accuracy: 0.997357 |
| | FPR: 0.110860 |
| | IoT Device 2 Accuracy: 0.999414 FPR: 0.111226 IoT Device 3 Accuracy: 0.998819 |
| | FPR: 0.111539 |
| | IoT Device 4 Accuracy: 0.999023 FPR: 0.134396 IoT Device 5 Accuracy: 0.999825 |
| | FPR: 0.000000 |
| | IoT Device 6 Accuracy: 0.998789 FPR: 0.066953 IoT Device 7 Accuracy: 0.998868 FPR: 0.062750 |
| | IoT Device 8 Accuracy: 0.998145 FPR: 0.159937 IoT Device 9 Accuracy: 0.997864 |
| | FPR: 0.100517 |
| | FedGroup accuracy: 99.8661%, Running Times seconds: 153.9543, FPR: 7.7021% |
| | First round: same as the first round of FedAvg |
| | Second round: |
| | Camera group: IoT Device 0 and 4, |
| | Accuracy: 99.8509% using 'max depth': 45, 'min samples leaf': 2, 'min samples split': |
| | 6 |
| | Appliances group: IoT Device 9, |
| | Accuracy: 99.8310% using 'max depth': 10, 'min samples leaf': 4, 'min samples split': |
| | 7 |
| | Controller group: IoT Device 2, |
| | Accuracy: 99.9450% using 'max depth': 10, 'min samples leaf': 1, 'min samples split': |
| | 0.1 |
| | Energy group: IoT Device 1, 3, 5, 6, 7, and 8, |
| | Accuracy: 99.8638% using 'max depth': 47, 'min samples leaf': 2, 'min samples split': |
| | 5 |
| | |

Table A.6: Decision Tree: Attack Detection

| | T '' D ' Att 1 D t t' |
|-----------------------------|--|
| Parameters to | Logistic Regression: Attack Detection Traditional Machine learning model accuracy: 99.7563% |
| choose: | Running Times seconds: 21375.6019, FPR: 24.4798% |
| C = 0.01, 0.1, | FedAvg accuracy: 99.7654%, Running Times seconds: 2912.2639, FPR: 20.2837% |
| C = 0.01, 0.1, $1, 10, 100$ | First round: |
| 1, 10, 100 | IoT device 0 Best Accuracy: 0.996143 FPR: 0.234854 using 'C': 0.1 |
| | IoT device 1 Best Accuracy: 0.995844 FPR: 0.110688 using 'C': 1 |
| | IoT device 2 Best Accuracy: 0.998931 FPR: 0.166839 using 'C': 10 |
| | IoT device 3 Best Accuracy: 0.997947 FPR: 0.296582 using 'C': 1 |
| | IoT device 4 Best Accuracy: 0.997942 FPR: 0.280559 using 'C': 0.01 |
| | IoT device 5 Best Accuracy: 0.999851 FPR: 0.083333 using 'C': 0.1 |
| | IoT device 6 Best Accuracy: 0.998057 FPR: 0.250115 using 'C': 1 |
| | IoT device 7 Best Accuracy: 0.998905 FPR: 0.083619 using 'C': 0.1 |
| | IoT device 8 Best Accuracy: 0.996711 FPR: 0.333476 using 'C': 0.1 |
| | IoT device 9 Best Accuracy: 0.996908 FPR: 0.180287 using 'C': 0.01 |
| | Second round: |
| | Aggregate updates with mean parameters: using 'C': 1.34 |
| | IoT Device 0 Accuracy: 0.996020 FPR: 0.110572 IoT Device 1 Accuracy: 0.995777 |
| | FPR: 0.110860 |
| | IoT Device 2 Accuracy: 0.998877 FPR: 0.166839 IoT Device 3 Accuracy: 0.998006 |
| | FPR: 0.296582 |
| | IoT Device 4 Accuracy: 0.997897 FPR: 0.280559 IoT Device 5 Accuracy: 0.999851 |
| | FPR: 0.083333 |
| | IoT Device 6 Accuracy: 0.998057 FPR: 0.250115 IoT Device 7 Accuracy: 0.998885 |
| | FPR: 0.073202 |
| | IoT Device 8 Accuracy: 0.996688 FPR: 0.333512 IoT Device 9 Accuracy: 0.996610 |
| | FPR: 0.180287 |
| | FedGroup accuracy: 99.7691%, Running Times seconds: 2998.8492, FPR: |
| | 20.1825% |
| | First round: same as the first round of FedAvg |
| | Second round: |
| | Camera group: IoT Device 0 and 4, Accuracy: 99.7093% using 'C': 0.055 |
| | Appliances group: IoT Device 9, Accuracy: 99.7175% using 'C': 0.01 |
| | Controller group: IoT Device 2, Accuracy: 99.9028% using 'C': 10 |
| | Energy group : IoT Device 1, 3, 5, 6, 7, and 8, Accuracy: 99.7753% using 'C': 0.055 |

Table A.7: Logistic Regression: Attack Detection

| | Ensemble Learning: Attack Detection |
|----------------|---|
| Parameters to | Traditional Machine learning model accuracy: 99.8490% |
| choose: | Running Times seconds: 33940.1807, FPR: 9.6031% |
| knn n | FedAvg accuracy: 99.9185% |
| neighbors': 1, | Running Times seconds: 2390, FPR: 9.03% |
| 2, 3, 4, 5, | First round: Fine-tuning 'knn n neighbors' |
| | Second round: Aggregate updates with mean parameters |
| | IoT Device 0 Accuracy: 0.998792 IoT Device 1 Accuracy: 0.998542 |
| | IoT Device 2 Accuracy: 0.999314 IoT Device 3 Accuracy: 0.999680 |
| | IoT Device 4 Accuracy: 0.999538 IoT Device 5 Accuracy: 0.999857 |
| | IoT Device 6 Accuracy: 0.999085 IoT Device 7 Accuracy: 0.999343 |
| | IoT Device 8 Accuracy: 0.999165 IoT Device 9 Accuracy: 0.998514 |
| | FedGroup accuracy: 99.9185% |
| | Running Times seconds: 2143, FPR: 9.43% |
| | First round: same as the first round of FedAvg_EL |
| | Second round: Aggregate updates with mean parameters based on groups |
| | Camera group: IoT Device 0 and 4, Accuracy: 99.91205% |
| | Appliances group: IoT Device 9, Accuracy: 99.8514% |
| | Controller group: IoT Device 2, Accuracy: 99.9314% |
| | Energy group : IoT Device 1, 3, 5, 6, 7, and 8, Accuracy: 99.9278% |

Table A.8: Ensemble Learning: Attack Detection

Appendix A.2. Appendix B: Results of models: Attack Type Detection

| | Decision Tree: Attack Type Detection |
|----------------------|---|
| Parameters to | Traditional Machine learning model accuracy: 88.4082% |
| choose: | Running Times seconds: 35.0063, FPR: 0.2651% |
| $\max depth =$ | FedAvg accuracy: 93.8953%, Running Times seconds: 10.9335, FPR: 0.3450% |
| 5, 10, 15, 20, | First round: Hyperparameters and Fine-tuning |
| 30, 40, 50, 100, | Second round: |
| 150, 200 | Aggregate updates with mean parameters: 'max depth': 81, 'min samples leaf': 2, |
| min samples | 'min samples split': 3 |
| split = 0.1, | IoT Device 0 Accuracy: 0.885343 FPR: 0.001042 IoT Device 1 Accuracy: 0.917624 |
| 0.2, 0.3, 0.4, | FPR: 0.001149 |
| 0.5, 0.6, 0.7, | IoT Device 2 Accuracy: 0.956254 FPR: 0.006944 IoT Device 3 Accuracy: 0.919554 |
| 0.8, 0.9, 1.0, 2, | FPR: 0.001868 |
| 3, 4, 5, 6, 7 | IoT Device 4 Accuracy: 0.875921 FPR: 0.010263 IoT Device 5 Accuracy: 0.914667 |
| min samples | FPR: 0.000000 |
| leaf = 1, 2, 3, | IoT Device 6 Accuracy: 0.935000 FPR: 0.004762 IoT Device 7 Accuracy: 0.914859 |
| 4, 5 | FPR: 0.001208 |
| | IoT Device 8 Accuracy: 0.855905 FPR: 0.003824 IoT Device 9 Accuracy: 0.888499 |
| | FPR: 0.003437 |
| | FedGroup accuracy: 94.8587%, Running Times seconds: 10.9313, FPR: 0.2663% |
| | <u>First round:</u> same as the first round of FedAvg |
| | Second round: |
| | Camera group: IoT Device 0 and 4, Accuracy: 91.4581% |
| | using 'max depth': 65, 'min samples leaf': 2, 'min samples split': 3 |
| | Appliances group: IoT Device 9, Accuracy: 91.7320% |
| | using 'max depth': 40, 'min samples leaf': 2, 'min samples split': 3 |
| | Controller group: IoT Device 2, Accuracy: 100.0000% |
| | using 'max depth': 10, 'min samples leaf': 1, 'min samples split': 3 |
| | Energy group : IoT Device 1, 3, 5, 6, 7, and 8, Accuracy: 94.8587% |
| | using 'max depth': 106, 'min samples leaf': 2, 'min samples split': 4 |

Table A.9: Decision Tree: Attack Type Detection

| | Logistic Regression: Attack Type Detection |
|----------------|---|
| Parameters to | Traditional Machine learning model accuracy: 39.7257% |
| choose: | Running Times seconds: 5442.6151, FPR: 1.3722% |
| C = 0.01, 0.1, | FedAvg accuracy: 49.5542%, Running Times seconds: 183.2524, FPR: 2.7627% |
| 1, 10, 100 | First round: |
| | IoT device 0 Best Accuracy: 0.475672 FPR: 0.012224 using 'C': 100 |
| | IoT device 1 Best Accuracy: 0.377739 FPR: 0.016092 using 'C': 1 |
| | IoT device 2 Best Accuracy: 0.531175 FPR: 0.048611 using 'C': 0.01 |
| | IoT device 3 Best Accuracy: 0.360421 FPR: 0.032711 using 'C': 100 |
| | IoT device 4 Best Accuracy: 0.451356 FPR: 0.040175 using 'C': 0.01 |
| | IoT device 5 Best Accuracy: 0.784667 FPR: 0.000000 using 'C': 10 |
| | IoT device 6 Best Accuracy: 0.421944 FPR: 0.042857 using 'C': 10 |
| | IoT device 7 Best Accuracy: 0.282076 FPR: 0.034695 using 'C': 1 |
| | IoT device 8 Best Accuracy: 0.486789 FPR: 0.015952 using 'C': 0.01 |
| | IoT device 9 Best Accuracy: 0.496668 FPR: 0.180287 using 'C': 0.01 |
| | Second round: |
| | Aggregate updates with mean parameters: using 'C': 22 |
| | IoT Device 0 Accuracy: 0.477079 FPR: 0.012224 IoT Device 1 Accuracy: 0.369502 FPR: 0.018391 |
| | IoT Device 2 Accuracy: 0.461066 FPR: 0.055556 IoT Device 3 Accuracy: 0.350340 FPR: 0.032784 |
| | IoT Device 4 Accuracy: 0.418393 FPR: 0.041842 IoT Device 5 Accuracy: 0.784667 FPR: 0.000000 |
| | IoT Device 6 Accuracy: 0.420278 FPR: 0.042857 IoT Device 7 Accuracy: 0.255525 FPR: 0.034695 |
| | IoT Device 8 Accuracy: 0.375752 FPR: 0.022563 IoT Device 9 Accuracy: 0.414895 FPR: 0.015360 |
| | FedGroup accuracy: 52.0105%, Running Times seconds: 198.7906, FPR: |
| | 2.6341% |
| | First round: same as the first round of FedAvg |
| | Second round: |
| | Camera group: IoT Device 0 and 4, Accuracy: 48.4768% using 'C': 50 |
| | Appliances group: IoT Device 9, Accuracy: 67.8665% using 'C': 0.01 |
| | Controller group: IoT Device 2, Accuracy: 57.8307% using 'C': 0.01 |
| | Energy group : IoT Device 1, 3, 5, 6, 7, and 8, Accuracy: 49.5757% using 'C': 20 |
| | 3. 3 1 |

Table A.10: Logistic Regression: Attack Type Detection

| | Ensemble Learning: Attack Detection |
|-----------------|--|
| Parameters to | Traditional Machine learning model accuracy: 99.8490% |
| choose: | Running Times seconds: 33940.1807, FPR: 9.6031% |
| 'cart max | FedAvg accuracy: 99.4963% |
| depth': 10, 20, | Running Times seconds: 370.6777, FPR: 0.0286% |
| 30, 40 | First round: Using Samsung Smart Cam to adjust the level-0 model in the Stacking |
| 'cart min | Ensemble learning then Fine-tuning the parameters |
| samples leaf': | Second round: |
| 1, 2, 3 | Aggregate updates with mean parameters: using knn n neighbors = 2, knn p = 1, |
| 'cart min | cart max depth = 10 , cart min samples split = 0.4 , cart min samples leaf = 1 |
| samples split': | IoT Device 0 Accuracy: 0.989538 FPR: 0.000000 IoT Device 1 Accuracy: 1.000000 |
| 0.1, 0.2, 0.3, | FPR: 0.000000 |
| 0.4, 0.5, 0.6, | IoT Device 2 Accuracy: 1.000000 FPR: 0.000000 IoT Device 3 Accuracy: 1.000000 |
| 0.7, 0.8, 0.9, | FPR: 0.000000 |
| 1.0 | IoT Device 4 Accuracy: 1.000000 FPR: 0.000000 IoT Device 5 Accuracy: 1.000000 |
| 'knn n | FPR: 0.000000 |
| neighbors': 1, | IoT Device 6 Accuracy: 1.000000 FPR: 0.002381 IoT Device 7 Accuracy: 1.000000 |
| 2, 3 | FPR: 0.000000 |
| 'knn p': 1, 2 | IoT Device 8 Accuracy: 0.983218 FPR: 0.000476 IoT Device 9 Accuracy: 0.945748 |
| | FPR: 0.000000 |
| | FedGroup accuracy: 99.6444% |
| | Running Times seconds: 341.4309, FPR: 0.0238% |
| | First round: same as the first round of FedAvg |
| | Second round: |
| | Camera group: IoT Device 0 and 4, Accuracy: 100.0000% |
| | using knn n neighbors $= 2$, knn p $= 1$, cart max depth $= 10$, cart min samples split |
| | = 0.1, cart min samples leaf $= 1$ |
| | Appliances group: IoT Device 9, Accuracy: 100.0000% |
| | using knn n neighbors $= 3$, knn p $= 2$, cart max depth $= 10$, cart min samples split |
| | = 1.0, cart min samples leaf $= 1$ |
| | Controller group: IoT Device 2, Accuracy: 100.0000% |
| | using knn n neighbors = 1, knn p = 1, cart max depth = 10, cart min samples split |
| | = 0.1, cart min samples leaf $= 1$ |
| | Energy group : IoT Device 1, 3, 5, 6, 7, and 8, Accuracy: 99.4074% |
| | using knn n neighbors = 3, knn p = 1, cart max depth = 10, cart min samples split |
| | = 0.1, cart min samples leaf $= 1$ |

Table A.11: Ensemble Learning: Attack Detection

| | Ensemble Learning: Attack Type Detection Detail 1 |
|-------------------|---|
| Parameters to | FedAvg_EL accuracy: 99.8941% |
| choose: | Running Times seconds: 4447.7158, FPR: 4.7904% |
| 'knn n | <u>First round:</u> |
| neighbors': 1, | IoT Device 0 Best: 0.998792 using |
| 2, 3, 4, 5, 10, | Direct or Reflection: 'bayes var smoothing': 1e-13, 'cart criterion': 'gini', 'knn n |
| 15, 19 | neighbors': 2; Type of attack: 'bayes var smoothing': 1e-10, 'cart criterion': 'gini', |
| 'cart criterion': | 'knn n neighbors': 3; Rate of attack: 'bayes var smoothing': 1e-13, 'cart criterion': |
| 'gini', 'entropy' | 'gini', 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-13, 'cart |
| 'bayes var | criterion': 'entropy', 'knn n neighbors': 4. |
| smoothing': | IoT Device 1 Best: 0.998542 using |
| 1e-7, 1e-08, | Direct or Reflection: 'bayes var smoothing': 1e-13, 'cart criterion': 'gini', 'knn n |
| 1e-09, 1e-10, | neighbors': 3; Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', |
| 1e-11, 1e-12, | 'knn n neighbors': 2; Rate of attack: 'bayes var smoothing': 1e-14, 'cart criterion': |
| 1e-13, 1e-14 | 'gini', 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-09, 'cart |
| | criterion': 'entropy', 'knn n neighbors': 3. |
| | IoT Device 2 Best: 0.999314 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-07, 'cart criterion': 'entropy', 'knn |
| | n neighbors': 10; Type of attack: 'bayes var smoothing': 1e-11, 'cart criterion': |
| | 'entropy', 'knn n neighbors': 4; Rate of attack: 'bayes var smoothing': 1e-13, 'cart |
| | criterion': 'entropy', 'knn n neighbors': 1; Layer of attack: 'bayes var smoothing': |
| | 1e-12, 'cart criterion': 'entropy', 'knn n neighbors': 5. |
| | IoT Device 3 Best: 0.999680 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-10, 'cart criterion': 'entropy', 'knn n |
| | neighbors': 3; Type of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'gini', |
| | 'knn n neighbors': 5; Rate of attack: 'bayes var smoothing': 1e-7, 'cart criterion': |
| | 'gini', 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-14, 'cart |
| | criterion': 'gini', 'knn n neighbors': 4. |
| | IoT Device 4 Best: 0.999538 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-14, 'cart criterion': 'gini', 'knn n |
| | neighbors': 3; Type of attack: 'bayes var smoothing': 1e-14, 'cart criterion': 'entropy', |
| | 'knn n neighbors': 4; Rate of attack: 'bayes var smoothing': 1e-9, 'cart criterion': |
| | 'gini', 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-11, 'cart |
| | criterion': 'gini', 'knn n neighbors': 4. |
| | IoT Device 5 Best: 0.999857 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', 'knn n |
| | neighbors': 15; Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', |
| | 'knn n neighbors': 10; Rate of attack: 'bayes var smoothing': 1e-11, 'cart criterion': |
| | 'gini', 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-12, 'cart |
| | criterion': 'entropy', 'knn n neighbors': 10. |
| | IoT Device 6 Best: 0.999085 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-8, 'cart criterion': 'entropy', 'knn n |
| | neighbors': 2; Type of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'entropy', |
| | 'knn n neighbors': 3; Rate of attack: 'bayes var smoothing': 1e-8, 'cart criterion': |
| | 'gini', 'knn n neighbors': 2; Layer of attack: 'bayes var smoothing': 1e-14, 'cart |
| | criterion': 'gini', 'knn n neighbors': 3. |
| | IoT Device 7 Best: 0.999343 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-11, 'cart criterion': 'gini', 'knn n |
| | neighbors': 4; Type of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'entropy', |
| | 'knn n neighbors': 4; Rate of attack: 'bayes var smoothing': 1e-8, 'cart criterion': |
| | 'gini', 'knn n neighbors': 4; Layer of attack: 'bayes var smoothing': 1e-14, 'cart |
| | criterion': 'entropy', 'knn n neighbors': 1. |

Table A.12: Ensemble Learning: Attack Type Detection Detail $\boldsymbol{1}$

| | Ensemble Learning: Attack Type Detection Detail 2 |
|-------------------|---|
| Parameters to | FedAvg_EL accuracy: 99.8941% |
| choose: | Running Times seconds: 4447.7158, FPR: 4.7904% |
| 'knn n | First round: |
| neighbors': 1, | IoT Device 8 Best: 0.999165 using |
| 2, 3, 4, 5, 10, | Direct or Reflection: 'bayes var smoothing': 1e-7, 'cart criterion': 'entropy', 'knn n |
| 15, 19 | neighbors': 2; Type of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'entropy', |
| 'cart criterion': | 'knn n neighbors': 2; Rate of attack: 'bayes var smoothing': 1e-12, 'cart criterion': |
| 'gini', 'entropy' | 'entropy', 'knn n neighbors': 5; Layer of attack: 'bayes var smoothing': 1e-13, 'cart |
| 'bayes var | criterion': 'entropy', 'knn n neighbors': 2. |
| smoothing': | IoT Device 9 Best: 0.998514 using |
| 1e-7, 1e-08, | Direct or Reflection: 'bayes var smoothing': 1e-14, 'cart criterion': 'entropy', 'knn n |
| 1e-09, 1e-10, | neighbors': 2; Type of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'entropy', |
| 1e-11, 1e-12, | 'knn n neighbors': 3; Rate of attack: 'bayes var smoothing': 1e-13, 'cart criterion': |
| 1e-13, 1e-14 | entropy', 'knn n neighbors': 2; Layer of attack: 'bayes var smoothing': 1e-8, 'cart |
| | criterion': 'entropy', 'knn n neighbors': 3. |
| | Second round: using Aggregate updates with mean parameters |
| | Direct or Reflection: 'baye var smoothing': 1e-11, 'cart criterion': 'gini, 'knn n neigh- |
| | bors': 5; Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'entropy', 'knn |
| | n neighbors': 4; Rate of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'gini', |
| | 'knn n neighbors': 3; Layer of attack: 'bayes var smoothing': 1e-12, 'cart criterion': |
| | 'entropy', 'knn n neighbors': 4. |
| | IoT Device 0: 0.998810, IoT Device 1: 0.998828, IoT Device 2: 0.999343, IoT Device |
| | 3: 0.999609, IoT Device 4: 0.999538, IoT Device 5: 0.999914, IoT Device 6: 0.998971, |
| | IoT Device 7: 0.999165, IoT Device 8: 0.999201, IoT Device 9: 0.998485 |

Table A.13: Ensemble Learning: Attack Type Detection Detail 2 $\,$

| | Ensemble Learning: Attack Type Detection Detail 2 |
|---------------------------------|--|
| Parameters to | FedGroup accuracy: 99.8878% |
| choose: | Running Times seconds: 4431.1258, FPR: 5.2349% |
| knn n | First round: same as the first round of FedAvg_EL |
| neighbors': 1, | Second round: |
| 2, 3, 4, 5, 10, | Camera group: IoT Device 0 and 4, Accuracy: 0.9991205 using |
| 15, 19 'cart criterion': | Direct or Reflection: 'bayes var smoothing': 1e-14, 'cart criterion': 'gini', 'knn n neighbors': 3 |
| 'gini', 'entropy' 'bayes var | Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', 'knn n neighbors': 4 |
| smoothing': | Rate of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'gini', 'knn n neighbors': |
| 1e-7, 1e-08, 1e-09, 1e-10, | 3 Layer of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', 'knn n neigh- |
| 1e-11, 1e-12, | bors': 4 |
| 1e-13, 1e-14 | Appliances group: IoT Device 9, Accuracy: 0.998514 using |
| , | Direct or Reflection: 'bayes var smoothing': 1e-14, 'cart criterion': 'entropy', 'knn n neighbors': 2 |
| | Type of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'entropy', 'knn n |
| | neighbors': 3 |
| | Rate of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'entropy', 'knn n neighbors': 2 |
| | Layer of attack: 'bayes var smoothing': 1e-8, 'cart criterion': 'entropy', 'knn n neigh- |
| | bors': 3 |
| | Controller group: IoT Device 2, Accuracy: 0.999314 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-07, 'cart criterion': 'entropy', 'knn n neighbors': 10 |
| | Type of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'entropy', 'knn n neighbors': 4 |
| | Rate of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'entropy', 'knn n neigh- |
| | bors': 1 |
| | Layer of attack: bayes var smoothing: 1e-12, cart criterion: entropy, kin in neighbors': 5 |
| | Energy group: IoT Device 1,3,5,6,7 and 8, Accuracy: 0.99927767 using |
| | Direct or Reflection: 'bayes var smoothing': 1e-10, 'cart criterion': 'gini', 'knn n |
| | |
| | Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', 'knn n neighbors': 3 |
| | Rate of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'gini', 'knn n neighbors': |
| | Layer of attack: 'bayes var smoothing': 1e-13, 'cart criterion': 'gini', 'knn n neigh- |
| | Layer of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'entropy', 'knn neighbors': 5 Energy group: IoT Device 1,3,5,6,7 and 8, Accuracy: 0.99927767 using Direct or Reflection: 'bayes var smoothing': 1e-10, 'cart criterion': 'gini', 'knn neighbors': 5 Type of attack: 'bayes var smoothing': 1e-12, 'cart criterion': 'gini', 'knn n neighbors 3 Rate of attack: 'bayes var smoothing': 1e-11, 'cart criterion': 'gini', 'knn n neighbors 4 |

Table A.14: Ensemble Learning: Attack Type Detection Detail $3\,$