## data\_treatment\_actual

June 22, 2021

[]: import sys# make sure all relevant libraries are installed

```
!{sys.executable} -m pip install pandas pyserial pyyaml
[5]: # import all functions used for processing data
     import pandas as pd
     from SerialLogger import write_logs
     from Analyzelogs import load_log, categorize_logfiles, calculate_all, \
     calculate_cost, extract_avg
    Reading content in progress, this may take some time...
    Writing log xor_256 for device Arduino Nano 33 IoT
    Writing log rng_256 for device Arduino Nano 33 IoT
    Done!
[]: # run this cell each time you log primitive(s) to write the .log files
     # from serial communication. store in test_results folder.
     write_logs("test_results/")
[5]: # Categorizing all the expermimental results on cryptographic primitives
     # for ESP32 into pandas dataframes. stored in dicts.
     root_dir = "./test_results"
     esp32_dir = "Adafruit_Feather"
     esp8266_dir = "Adafruit_Huzzah"
     nano_dir = "Arduino_Nano_33_IoT"
     # put all logfiles into dicts by each device, each logfile a pandas dataframe
     # with columnssample and time [microseconds]
     esp32 = categorize_logfiles(f"{root_dir}/{esp32_dir}")
     esp8266 = categorize_logfiles(f"{root_dir}/{esp8266_dir}")
     nano_iot = categorize_logfiles(f"{root_dir}/{nano_dir}")
     time_col = "time [microseconds]"
     esp32\_xor = esp32["xor\_256"][time\_col]
```

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print(round(esp32_xor.mean()))
    3
[6]: # Constructing the measured primitive table using the mean execution time in \square
     \rightarrow microseconds
     index = ["XOR", "RNG", "BLAKE2s", "SHA256", "SHA3-256", "AES256 Enc.",
      "AES256 Dec.", "AES-256-GCM", "NIST P-256 ECC+DH"]
     primitives = ['xor_256', 'rng_256', 'blake2s', 'sha2_256', 'sha3_256'
     ,'aes256_enc','aes256_dec','aes256_gcm', 'secp256r1_ecc_dh']
     # for each primitive in the list, calculate the avg execution time and put that \Box
     \rightarrow in a list.
     # returns a column of average execution time for each device
     esp32_avgs = extract_avg(esp32, primitives)
     esp8266_avgs = extract_avg(esp8266, primitives)
     nano_avgs = extract_avg(nano_iot, primitives)
     # display values in dataframe table setting to 2 significant digit
     pd.set_option('display.float_format', '{:.2E}'.format)
     # create primitive execution time table, columns for each device average_
      \rightarrow execution time.
     primitives_avg = pd.DataFrame({"ESP32" : esp32_avgs,
     "ESP8266" : esp8266_avgs, "Nano IoT" : nano_avgs}, index=index)
     print(primitives_avg)
     # store in file ./tables/primitives_avg.tex
     primitives_avg.to_latex('./tables/primitives_avg.tex')
                          ESP32 ESP8266 Nano IoT
    XOR.
                       3.00E-06 7.00E-06 1.10E-05
    R.N.G
                      3.10E-05 9.20E-05 1.70E-04
    BLAKE2s
                      1.90E-05 8.30E-05 1.44E-04
    SHA256
                      2.80E-05 1.07E-04 2.61E-04
                      1.90E-04 5.50E-04 1.34E-03
    SHA3-256
    AES256 Enc.
                      1.50E-05 1.79E-04 3.24E-04
    AES256 Dec.
                      9.00E-06 2.44E-04 5.25E-04
                       1.53E-04 1.03E-03 2.57E-03
    AES-256-GCM
    NIST P-256 ECC+DH 1.70E-01 9.78E-01 7.81E-01
[7]: # implemented scheme average execution time
     # index for dataframe/table implemented scheme average excution time
     implemented_schemes_index = ["HashXOR SHA3", "NewHope", "ECIES"]
     # keys to use in dictionary to get right dataframe
```

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implemented_schemes = ["hashxor", "newhope_client", "ecies"]
      # create 3 lists containing avg execution time of lists for each device
      # in form [hashxor_avq, newhope_avq, ecies_avq]
      esp32_avg_schemes = extract_avg(esp32, implemented_schemes)
      esp8266_avg_schemes = extract_avg(esp8266, implemented_schemes)
      nano_avg_schemes = extract_avg(nano_iot, implemented_schemes)
      # assemble dataframe where primitive is index and columns are avg execution,
      → time of schemes by device.
      implemented schemes_avg = pd.DataFrame({ "ESP32" : esp32_avg_schemes,
      "ESP8266" : esp8266_avg_schemes, "Nano IoT" : nano_avg_schemes},
      index=implemented_schemes_index)
      print(implemented_schemes_avg)
       # store processed data in file
      implemented_schemes_avg.to_latex('./tables/implemented_schemes_avg.tex')
                     ESP32 ESP8266 Nano IoT
     HashXOR SHA3 3.11E-04 8.56E-04 2.18E-03
     NewHope
                  1.12E-02
                                NaN 7.73E-02
     ECIES
                  1.70E-01 9.79E-01 7.84E-01
[13]: # theoretical scheme average execution time
      # khan: rand, 3 xor, 4 hash, 1 sc.dc
      # braeken: 7 hash
      # SEL-AKA: rand, 3 hash, sc.en
      # ecies: 1 hash, 1 ecc+dh, 1 aes256/gcm
      # hashxor: 2 rng, 3 xor, 2 distinct hashing functions
      # index to use in assembled dataframe of theoretical scheme aug execution time
      scheme index = ['Khan', "Braeken", "SEL-AKA", "HashXOR B1AKE2s",
      "HashXOR SHA3", "ECIES"]
      # primitives and amounts they're used for each scheme, used to calculate scheme
      # cost by table lookup in avg
      # primitive execution time table
      braeken_prims = [["SHA256", 7]]
      sel_aka_prims = [["RNG", 1], ["SHA256",3], ["AES256 Enc.", 1]]
      khan_prims = [["RNG", 1], ["XOR", 3], ["AES256 Dec.", 1], ["SHA256", 4]]
      hashxor_2 = [["RNG", 2], ["XOR", 3], ["SHA256", 1], ["BLAKE2s", 1]]
      ecies_prims = [["SHA256", 1], ["NIST P-256 ECC+DH", 1], ["AES-256-GCM", 1]]
      hashxor_prims = [["RNG", 2], ["XOR",3], ["SHA256", 1], ["SHA3-256", 1]]
      # scheme costs put in a list so they can be iterated through,
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# same order as index so they're matching
      schemes = [khan_prims, braeken_prims, sel_aka_prims,
       hashxor_2, hashxor_prims, ecies_prims ]
      # use scheme costs and goes through all of them, returns
      # list of any theoretical execution time for the schemes in the same order as_{\sqcup}
      \rightarrow index.
      esp32 schemes = calculate all(primitives avg, schemes, "ESP32")
      esp8266_schemes = calculate_all(primitives_avg, schemes, "ESP8266")
      nano_schemes = calculate_all(primitives_avg, schemes, "Nano IoT")
      # assemble dataframe/table by using columns as costs
      # for each device, and scheme names as index.
      schemes_theoretical = pd.DataFrame({"ESP32" : esp32_schemes,
       "ESP8266" : esp8266_schemes, "Nano IoT" : nano_schemes}, index=scheme_index)
      schemes_theoretical.to_latex("./tables/theoretical_schemes_avg.tex")
      print(schemes_theoretical)
                        ESP32 ESP8266 Nano IoT
     Khan
                     1.61E-04 7.85E-04 1.77E-03
                     1.96E-04 7.49E-04 1.83E-03
     Braeken
     SEL-AKA
                     1.30E-04 5.92E-04 1.28E-03
     HashXOR BlAKE2s 1.18E-04 3.95E-04 7.78E-04
     HashXOR SHA3 2.89E-04 8.62E-04 1.98E-03
     ECIES
                     1.70E-01 9.79E-01 7.84E-01
[73]: | # calculate percentage error between theoretical schemes and implemented schemes
      def percentage_difference(a,b):
          delta = abs(a - b);
          return round((delta/(a+b))*100,2)
      def percentage_error(a, b):
          delta = abs(a - b)
          return round((delta/abs(b))*100,2)
      def percentage_change(a,b):
          delta = a - b
          return round((delta/abs(b))*100, 2)
      # look at error differences between scheme theoretical and implemented
      theoretical = schemes_theoretical.at["ECIES", "ESP32"]*1e6
      implemented = implemented_schemes_avg.at["ECIES", "ESP32"]*1e6
```

```
print("theoretical: ",theoretical)
print("implemented: ",implemented)
error = percentage_error(theoretical, implemented)
difference = percentage_difference(theoretical, implemented);
change = percentage_change(implemented, theoretical)
print(f"error: {error}%\ndifference: {difference}%\nchange: {change}%")
```

theoretical: 170006.0 implemented: 170049.0

error: 0.03% difference: 0.01% change: 0.03%

[]: