Before you start

This assignment is almost entirely programming with a few written questions that you will answer based on the code you write. Hence, most of the instructions are in the starter code for this assignment. Here, we’ll summarize information that is not in the starter code. In addition to python3, you’ll need to install the python libraries for matplotlib, a graph plotting library. You should be able to install matplotlib by following the instructions here: https://matplotlib.org/users/installing.html.

This assignment will take more time than assignment 1. Please start early and ask for help if you’re stuck. Use github to upload your final assignment submission; we’ll send you github instructions separately. Keep pushing your code to github often in case your computer crashes the night before the assignment is due. When answering the written questions, provide as much detail as you think is appropriate, but err on the side of providing more rather than less detail for a question. We’ll read everything you turn in.

The assignment will also teach you how to develop protocols using a network simulator, a computer program that imitates the behavior of the essential components of a network. For this assignment, we’ll be using a custom simulator developed solely for assignment 2. I’ll list some of the non-obvious details of the simulator here because they might help you when debugging your solutions to assignment 2.

Randomness in the simulator. The simulator uses a random number generator to generate independent and identically distributed (IID) packet losses when packets are dequeued. This is in addition to packet drops if a queue overflows when packets are enqueued. You do not need to understand probability for this assignment, but you should be aware of the fact that randomness causes non-determinism in simulations. For instance, if you run two simulations with identical settings and the same loss rate, you might see different outputs because the sequence of packet drops will be randomly generated. To make such random simulations deterministic, you can use a seed to initialize a random number generator so that it generates the same sequence of random numbers. If you pass the same seed to the simulator in two different simulation runs with the same settings, the output will be the same. Also, if you don’t use random losses in your simulation (i.e., you set the loss ratio to 0), then your output should be deterministic because no other simulator component uses a random number generator.

Sequence numbers. As a matter of convention sequence numbers for packets start from 0 (i.e., the first packet has sequence number 0). Also, we’ll be dealing with providing reliability at the level of packets, not bytes.

Link capacity in the simulator. The link capacity in all our simulations is fixed to 1 packet per simulation tick. This seems like an arbitrary restriction, but it simplifies the implementation of the simulation and the assignment, without losing anything in terms of what you will learn from the assignment. All time-based quantities (RTT, timeout, $RTT_{min}$, queueing delay, etc.) in this assignment are measured in units of simulation ticks, an arbitrary unit of time we use in our simulations.

\[^{1}\text{In other words, every packet is dropped independent of every other packet, but we use the same probability to decide whether to drop a packet or not.}\]
**RTT\(_{\text{min}}\) in the simulator.** Because time is quantized to ticks in our simulation, an \(RTT_{\text{min}}\) of 10 ticks will result in a delay of 11 ticks between when the first packet is sent out in StopAndWait and when the second packet is sent out. This is because the first packet will be sent out in tick 0, its ACK will be received 10 ticks later, and then only in the next tick (tick 11) can the second packet be sent. For this reason, we subtract one from the user-specified \(RTT_{\text{min}}\) before feeding it into our simulator. This is not particularly important, but it’s good to know in case you wonder why there is some odd-looking code in simulator.py.

**Senders and receivers.** Because this is a simulator, we get to simplify as much as we want, while still imitating network components relevant to what we’re trying to study. Hence, the sender and the receiver are really the same Python object called host. This host could belong to one of three classes: StopAndWaitHost, SlidingWindowHost, or AimdHost. All three classes implement two methods send() and recv() corresponding to the sender and receiver respectively.

**Running the simulator.** Running `python3 simulator.py -h` or `python3 simulator.py --help` should give you the usage for the simulator. For instance, the simulator lets you set \(RTT_{\text{min}}\), the limit on the number of packets in the link’s queue, the loss ratio, the window size for the sliding window protocol and so on. The simulator reports the max sequence number that has been received in order at the end of the simulation period. Adding 1 to this gives you the number of packets that have been received in order. Dividing the number of packets by the simulation period will give you the transport-layer throughput.

### 1 Moving averages and retransmission timeouts (20 points)

#### 1.1 Moving averages (5 points)

Implement a simple exponentially weighted moving average (EWMA) filter by translating the equations for the EWMA from lecture 4 into Python code. We have provided starter code in `ewma.py`. You’ll use your implementation to understand how \(\alpha\) (also called the gain of the EWMA) affects how quickly the mean value calculated by the EWMA converges to the true value.

For this, we’ll use a synthetic set of 100 samples. The first 50 samples are 0, and the next 50 are 1. We’ll feed this to an EWMA and see how the mean estimate (also called the smoothed estimate) tracks the actual samples. Run `ewma.py` using two different \(\alpha\) values: one high (0.5) and one low (0.01). How does the value of \(\alpha\) affect convergence? Explain this using the equation for an EWMA filter.

#### 1.2 Retransmission timeouts (15 points)

Complete the TODOs in `timeout_calculator.py`, which is a class used by both the Stop-And-Wait and sliding window protocols for their timeout calculations. Make sure that any computed timeout values always fall between MIN\_TIMEOUT and MAX\_TIMEOUT. You’ll be able to test out your retransmission logic as part of the Stop-And-Wait and sliding window implementations in the next two questions.

Why do we need a minimum value of the retransmission timeout? Why do we need a maximum value of the retransmission timeout? These questions might be easier to answer once you have incorporated the retransmission logic into both the Stop-And-Wait and sliding window protocols. To answer these two questions, you could try disabling either the MIN or the MAX\_TIMEOUTs and see (1) what the effect on throughput is and (2) how it affects the likelihood of congestion collapse.

### 2 The Stop-And-Wait protocol (20 points)

Implement the Stop-And-Wait protocol using the starter code provided in `stop_and_wait_host.py`. Run the protocol using `simulator.py`. Carry out and report on the results of the following experiments. Use the simulator’s default large queue size limits (1M packets) for this experiment.
1. (7 points) Make sure your implementation works. This means checking that a packet is being sent out every $RTT_{min}$ ticks. Print out a line to the terminal every time you send out a packet, and attach this output with your submission.

2. (6 points) Run the Stop-And-Wait protocol for several different values of $RTT_{min}$. Check that the throughput tracks $\frac{1}{RTT_{min}}$ equation we derived in class. Plot the throughput you get and compare it with the equation we derived in class. You can use matplotlib for this.

3. (7 points) Introduce a small amount of IID loss (about 1%). Check (1) if the throughput continues to be close to the value predicted by the equation and (2) that the protocol continues to function correctly despite the loss of packets. If the throughput is much less than the value predicted by the equation, explain why, by looking at when packets are originally transmitted and when they are retransmitted. Does the divergence between the simulation’s throughput and the equation’s predicted throughput increase or decrease as $RTT_{min}$ increases? Why?

3 The sliding window protocol (20 points)

Implement the sliding window protocol using the starter code provided in sliding_window_host.py. Again, run the protocol using simulator.py. Use the simulator’s default large queue size limits for this experiment. Answer the same three questions as the Stop-And-Wait protocol in the previous section, but remember to vary the window size as well in addition to $RTT_{min}$. When a small amount of loss (1%) is introduced, how does the divergence between the simulation’s throughput and the equation’s predicted throughput vary now as a function of both $RTT_{min}$ and the window size. The point breakdown is the same as the previous question.

4 Congestion collapse (20 points)

We’ll now simulate congestion collapse using our simulator. For this, use the sliding window protocol. Calculate the bandwidth-delay product (BDP) as the product of the link capacity and the minimum round-trip time. Vary the window size from 1 to the BDP. Check that the transport-layer throughput matches up with the $\frac{W}{RTT_{min}}$ equation, similar to the previous question. Now vary the window size beyond the BDP all the way until 100.BDP. Plot the transport-layer throughput as a function of the window size. Can you see the congestion collapse pattern that we discussed in class (i.e., utility goes down as offered load increases)? What is the reason for this congestion collapse? Can you provide evidence for this by looking at the number of retransmissions and the number of original packets being delivered by the link?

For this assignment, to keep the simulations short, pick an $RTT_{min}$ of around 10 so that the BDP is around 10 (recall link capacity is 1). This will allow you to vary window sizes from 1 to 1000 and still complete each simulation in 100000 simulation ticks or so. Large window sizes will need a longer simulation time before the simulation settles into steady state. Do not introduce any IID loss (i.e., set loss ratio to 0) for this experiment. You could also try modifying MIN_TIMEOUT and MAX_TIMEOUT to observe a sharper version of the congestion collapse, which may also help you answer §1.2. Finally, you could also try reducing the queue size limit to observe a sharper version of the congestion collapse.

For this question, you get 10 points if you can demonstrate one set of simulation settings that leads to a congestion collapse. Write down what these settings were in your assignment submission. To demonstrate a congestion collapse, you must provide a plot of transport-layer throughput vs. window size that resembles the classic congestion collapse curve we discussed in class. You get the next 10 points if you can explain why this congestion collapse occurs and provide quantitative justification for your explanation using the number of original and retransmitted packets from your simulations.
5 AIMD (20 points)

In the final part of this assignment, you will implement the AIMD algorithm that fixes congestion collapse. Use the file aimd_host.py, which shares a considerable amount of code with sliding_window_host.py. So if you have completed sliding_window_host.py, most of your work for AIMD is already done. The only major new parts of AIMD are implementing the Additive Increase and Multiplicative Decrease rules. Answer the following questions. If you need a concrete value of $RTT_{\text{min}}$, you can set it to 10 ticks for this experiment, but feel free to use your own value of $RTT_{\text{min}}$ if you wish.

1. (5 points) First make sure the AIMD algorithm is implemented as per the instructions in the TODOs.

2. (5 points) Why do we wait for an RTT before we decrease the window a second time using multiplicative decrease? What would happen if we didn’t wait?

3. (5 points) Set the queue size limit to something small, like about half the BDP. Use matplotlib (or a program of your choice) to plot the evolution of the window size over time. You can plot the window size by printing out the window size and the tick number every time the send() function is called. Attach the plot with your submission. Do you see the additive increase, multiplicative decrease, sawtooth pattern that we discussed in the lecture? What is the period of the sawtooth in relation to $RTT_{\text{min}}$? What is the throughput of AIMD in this case?

4. (3 points) Increase the queue size limit from half the BDP to 1 BDP and then 2 and 3 BDP. What happens to the throughput of AIMD? Why?

5. (2 points) AIMD needs a certain amount of queue capacity so that it achieves throughput close to the link’s capacity. What is the purpose of this queue?