

Very useful application of Viterbi Algorithm for Hidden Markov Models.

Sequence of states at discrete time steps. The hidden part means we can't see the states directly. Only can see observations. We typically assume we know how the system transitions from one state to next, plus what observations are likely given each state.

The Problem

- Have: Sequence of observations {o₁, o₂, ..., o_n}
- Want: Sequence of states {s₁, s₂, ..., s_n}
- Viterbi Algorithm: Find most-likely explanation (sequence of states) for observations about a Hidden Markov Model



Example: Office Lighting

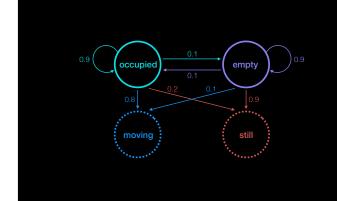


Example: Office Lighting



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Example: Office Lighting

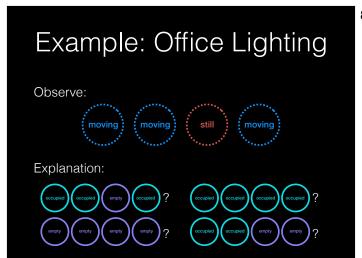






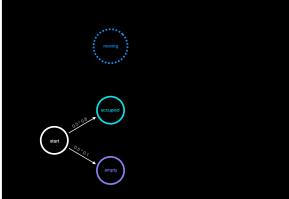






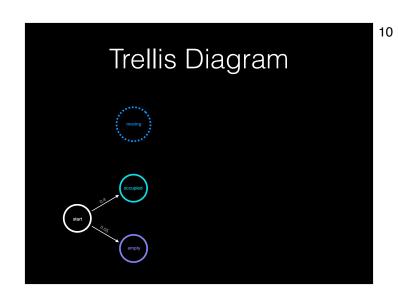
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Trellis Diagram

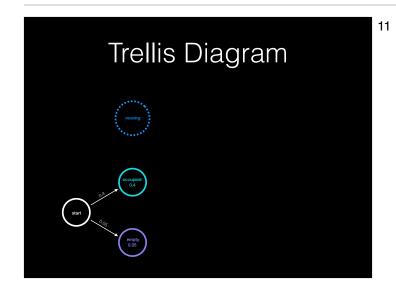


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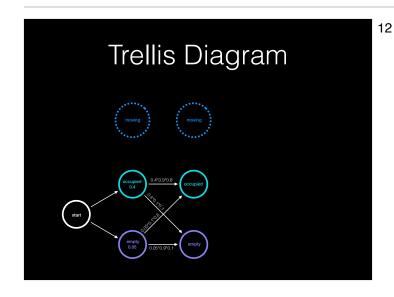
In order to visualize how the Viterbi Algorithm works, we'll make use of something called a trellis diagram. Observations on the top. States and transitions are below that. Here we're multiplying the probability of each state (50/50) times P(our_observation | state).



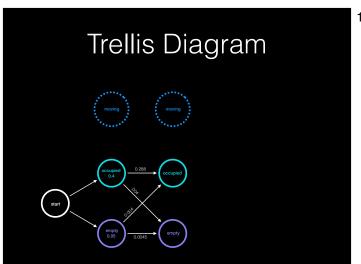
More specifically, we're using Bayes' Rule. Usually when you do this, you normalize by the probability of the observation, but here we only care about the maximum, so I'm skipping that part.

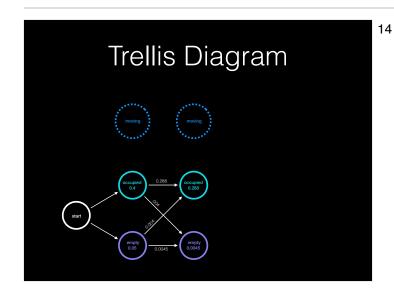


We fill in the states with their probabilities, and then we move on to the next time step.

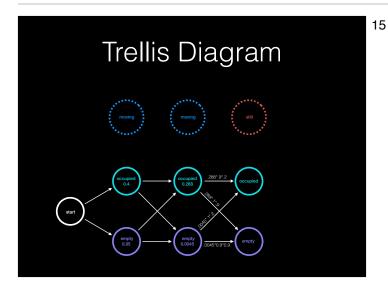


We have the same observation again, so, starting with the probabilities we just calculated, we multiply by the probability of each state transition, times the probability of the observation given the state we end up in.

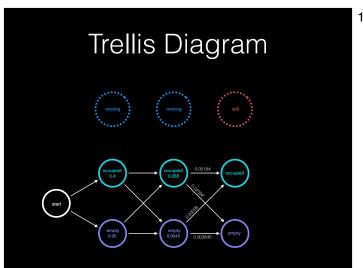


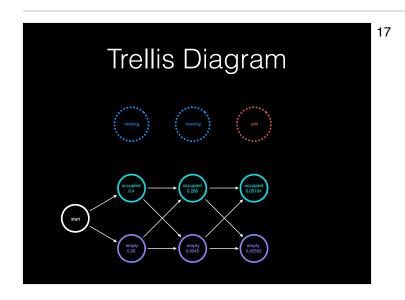


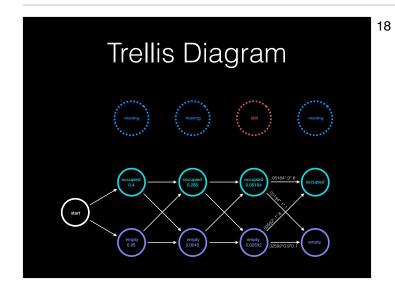
And then we update our state probabilities again.

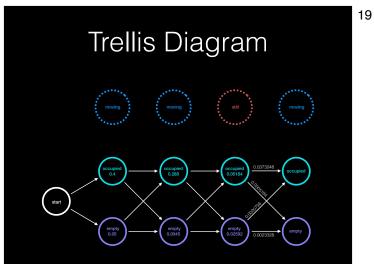


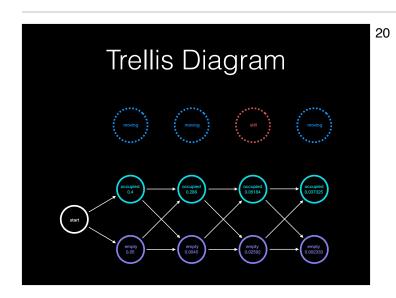
In the next time step we get a different observation, so the calculations change slightly



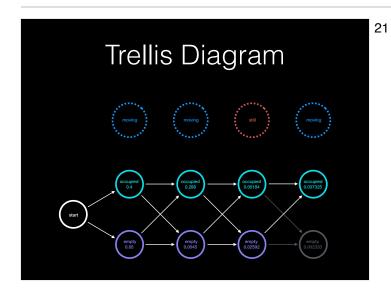




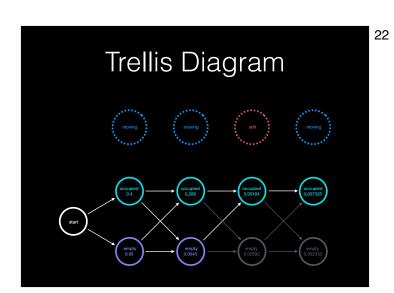




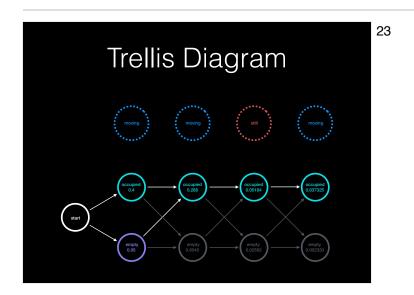
And finally we finish our trellis diagram. Now we look at the last two states and see which has the higher probability. That's where we end up.



We cross out the other state, and move backwards.



At each time step, we pick the state with the highest probability

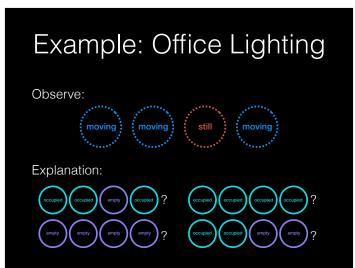


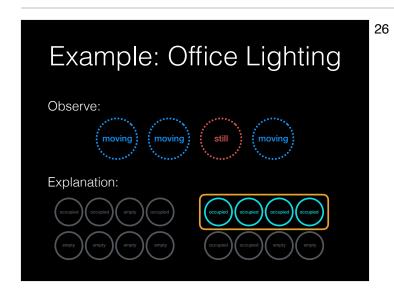
Trellis Diagram

moving only occupied o

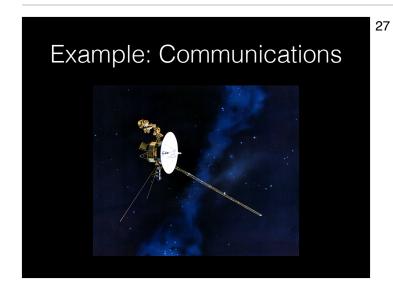
Until finally we have our entire path. This is called the Viterbi path, and it's the most likely explanation for the observations.



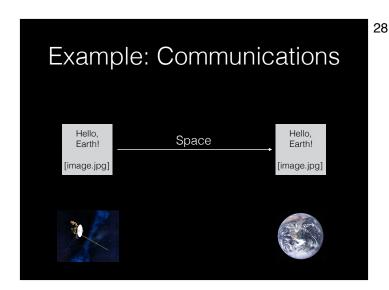




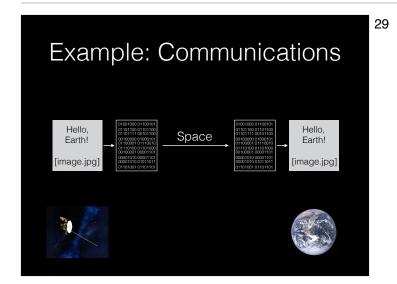
So basically, if the lab had a lighting controller like this, maybe the lights would stop randomly shutting off while I'm working causing me to wave my arms around frantically.



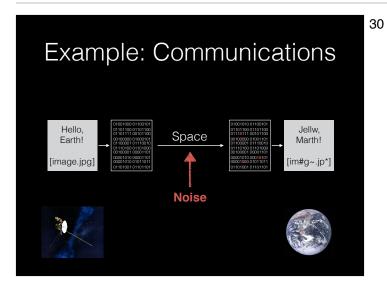
Another great use for the Viterbi Algorithm is in communications. This is a picture of Voyager 1 – one of several deep-space probes that used the algorithm for communications to and from Earth



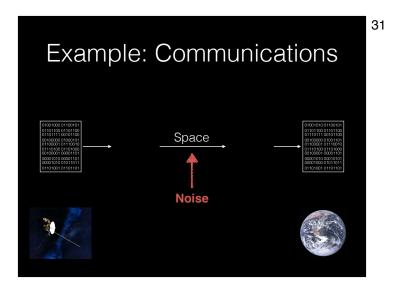
Typically communications works like this. You have some message, and you want to send through space back to Earth.

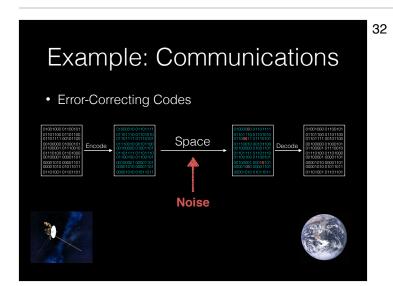


Normally we first convert the message to binary

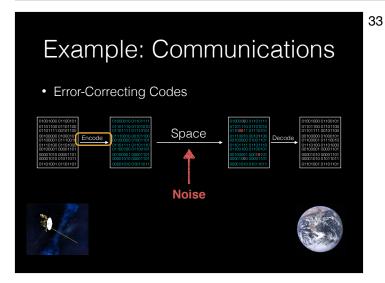


The problem is that usually the message gets corrupted by noise along the way

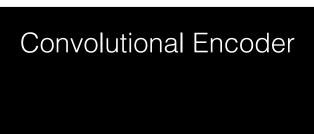




To solve this problem, we'll use something called an error correcting code. Basically, instead of transmitting the white bits, we send them through an encoder to get the blue bits. The blue bits contain extra information that helps recover the message.



I'll get to the decoder in a minute, but first I'll explain what's going on in the encoder.



110001010101010000...

The encoder operates on groups of 3 bits at a time.

Output: 110001

Convolutional Encoder

Input: 110001010101010000...

Output: 110001

Convolutional Encoder

Input: 110001010101010000...

10

Output: 110001

36

35



You slide this 3-bit window over by 1, and it produces 2 output bits

Convolutional Encoder

Input: 110001010101010000...

Output: 11000110

38

Convolutional Encoder

Input: 110001010101010000...

Output: 11000110

39

Convolutional Encoder

Input: 110001010101010000...

Output: 11000110



Input: 110001010101010000...

Output: 1100011000

Slide it over by one more bit, and you get another 2 bits out. This is a rate 1/2 encoder, because it produces twice as many bits of output as it gets input.

Convolutional Encoder

Input: 110001010101010000...

Output: 1100011000

41

40

Convolutional Encoder

Input: 110001010101010000...

11

Output: 1100011000

Convolutional Encoder

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Input: 110001010101010000...
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Output: 110001100011

Convolutional Encoder

Input: 110001010101010000...

Output: 110001100011

Convolutional Encoder

Input: 110001010101010000...

1

Output: 110001100011

44

Convolutional Encoder

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110001010101010000...
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Output: 11000110001111

Convolutional Encoder

110001010101010000...

Output: 11000110001111

Convolutional Encoder

110001010101010000...

Output: 110001100011111

48

50

51

Convolutional Encoder

Input: 11000 1010101010000...

Output: 1100011000111101

Convolutional Encoder

Input: 110001010101010000...

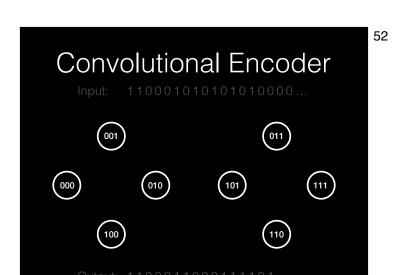
Output: 1100011000111101

If we move this stuff out of the way, we can draw a state diagram for what's going on here

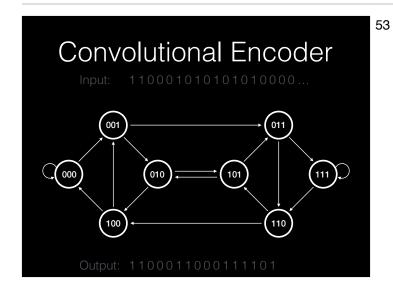
Convolutional Encoder

Input: 110001010101010000

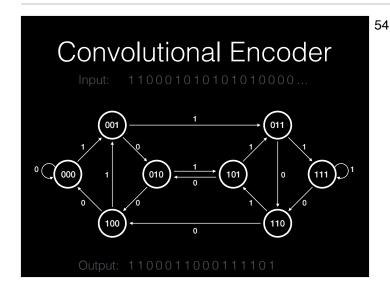
Output: 110001100011110



Each of these states corresponds to something our 3-bit window might see

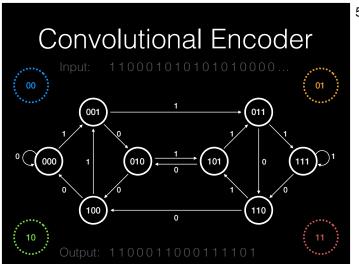


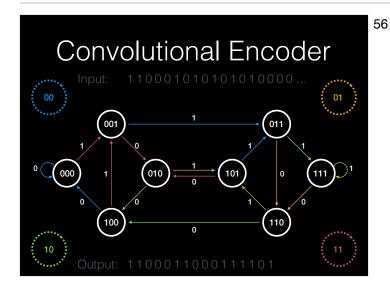
We can draw in the transitions with arrows



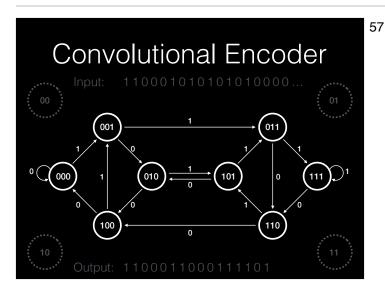
Each transition represents one of the two possible input bits.

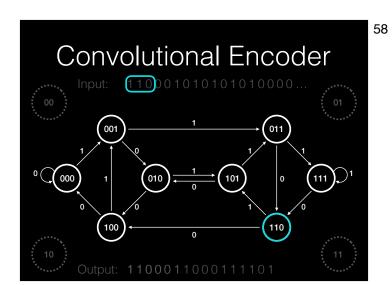




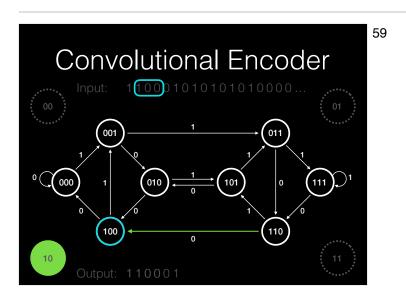


Each transition also produces two output bits





Now if we start in state 110, we can rerun our input and visualize the encoder state



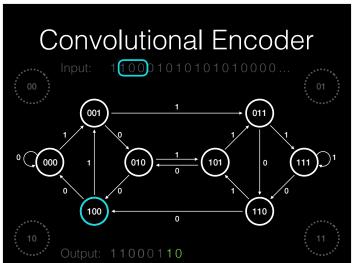
Convolutional Encoder

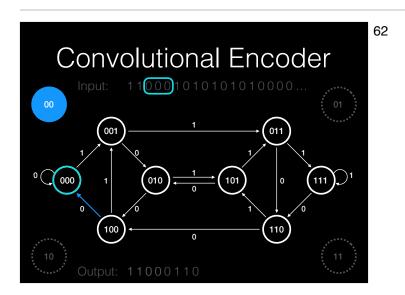
Input: 10001010101010000

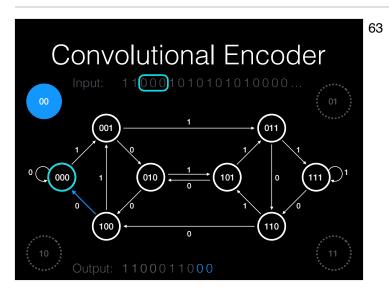
1000

Output: 11000110

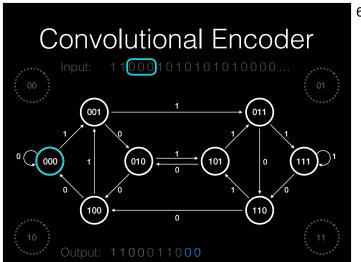


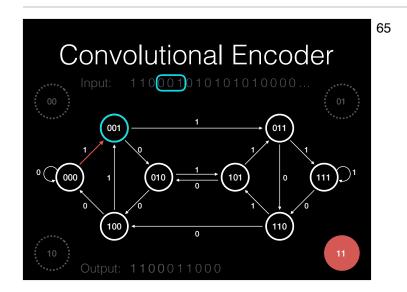


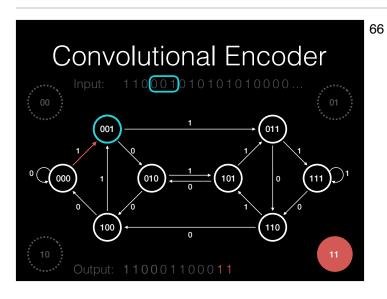




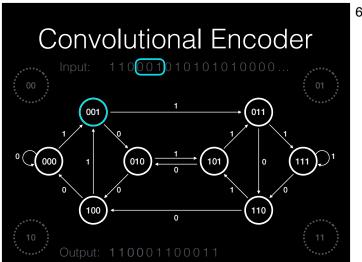


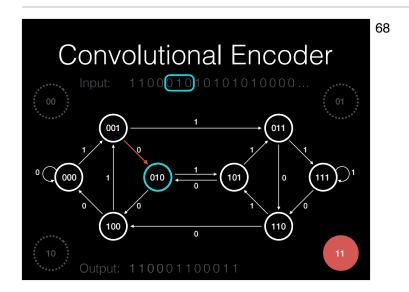


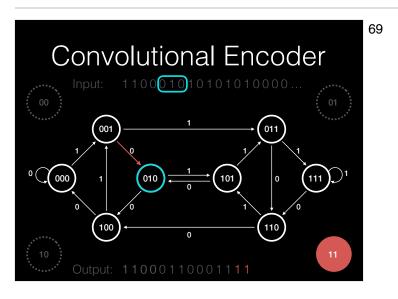




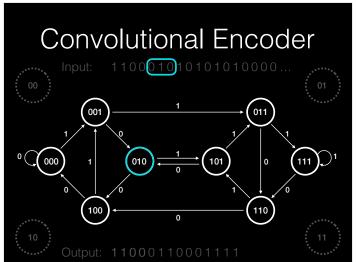


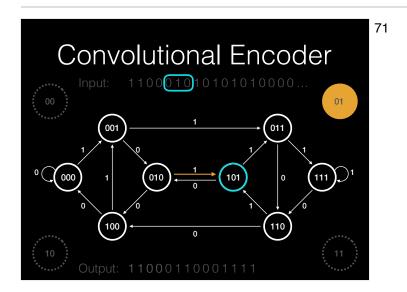


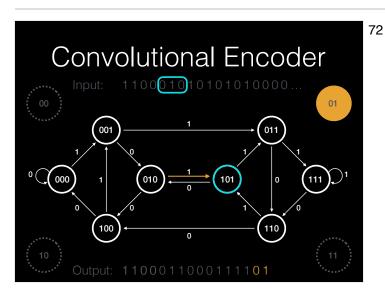


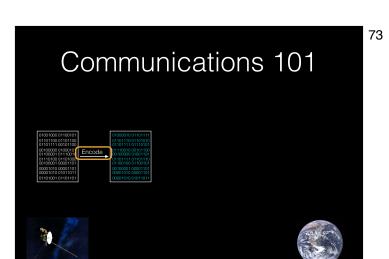




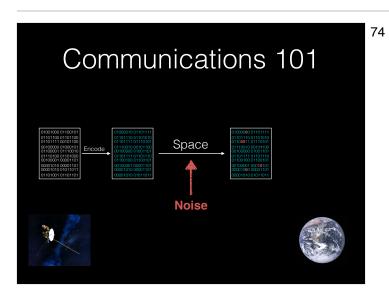




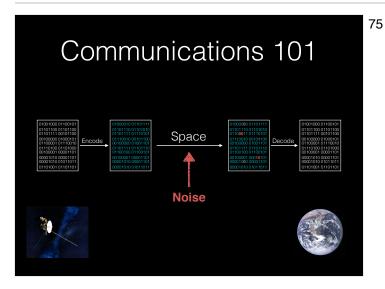




So that's the encoder

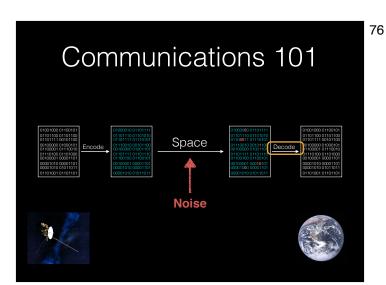


We still need to get our message through space



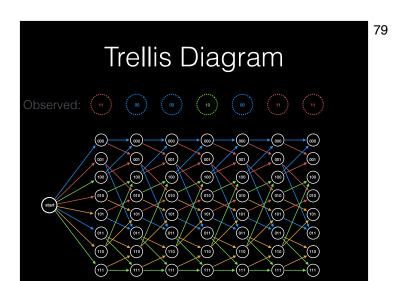
And then decode the noisy results to get our message back out.



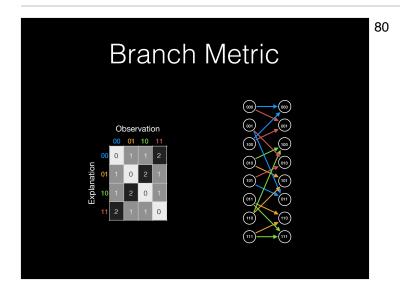


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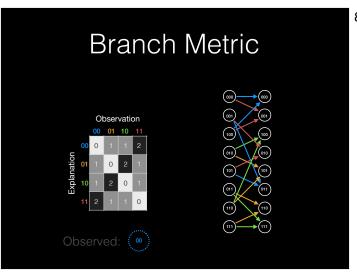
78 The Problem Have: Want: Sequence of observations Sequence of states • Viterbi Decoder: Find most-likely original transmitted message for a given set of noisy observed bits

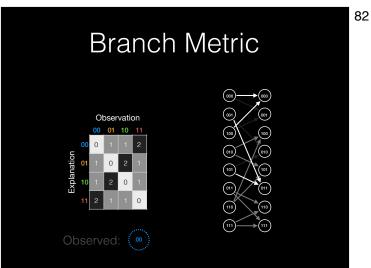


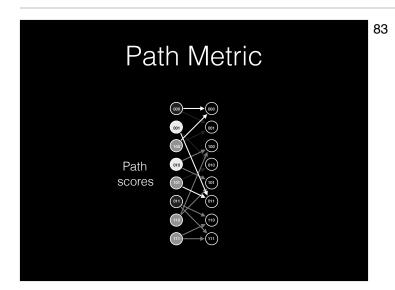
We're going to do the same thing as before, but this time the trellis diagram is a lot more complex.



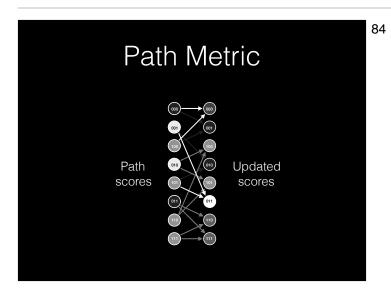
We're also going to make use of a few optimizations. Instead of computing conditional probabilities, we're going to just compute the Hamming distance between the observation and each explanation. Order stays the same, it's just simpler.



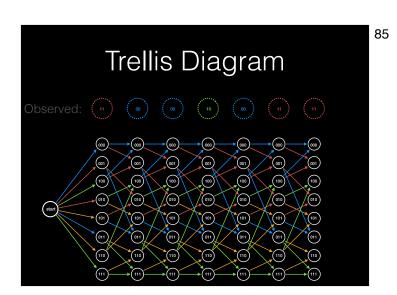




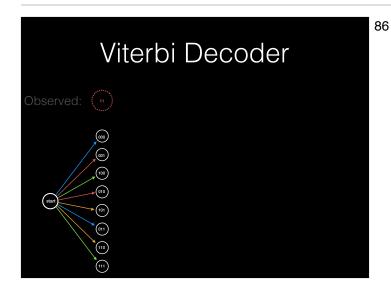
The other optimization is that instead of multiplying out our probabilities, we'll just add the Hamming errors for each branch. Again, all we care about is the most likely path, and this will give the same result. We take our path scores, and add the branch metrics.



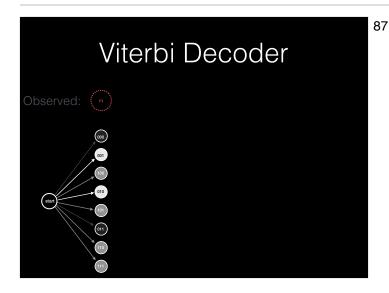
Then we compare the results, and select the better score as our next path metric.



Now we can go back to the trellis diagram, and start calculating.

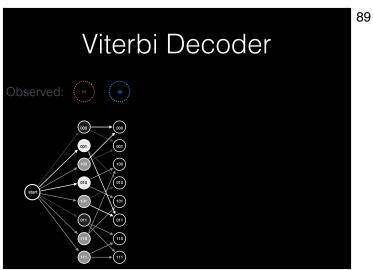


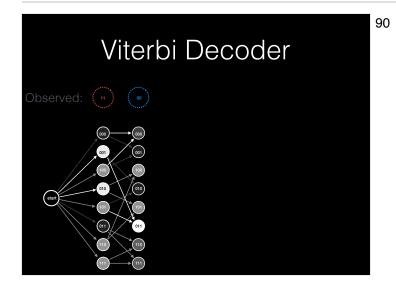
At the first time step, we observe a '11'.



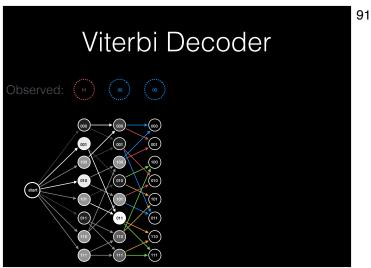
We compute initial branch metrics, and those give us our starting path metrics.

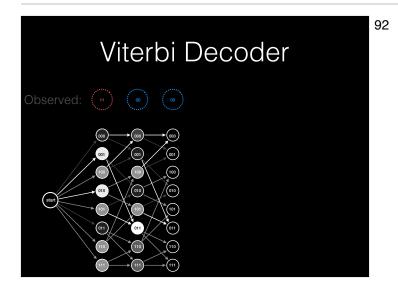


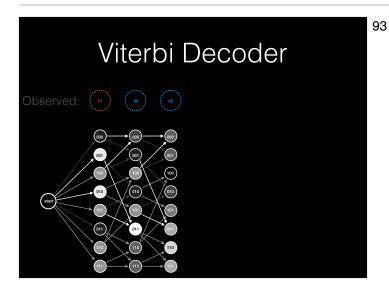




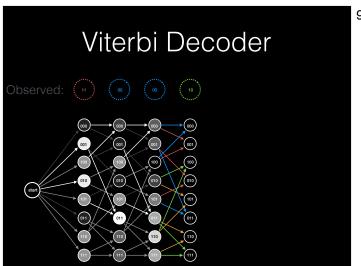


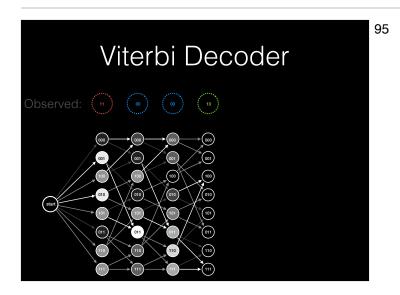


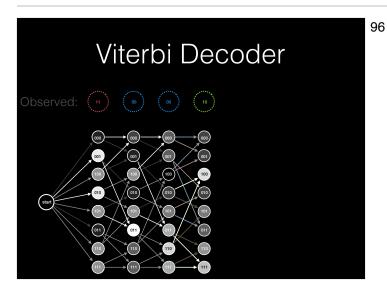


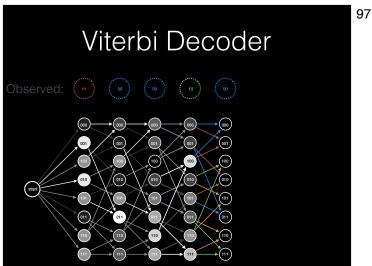


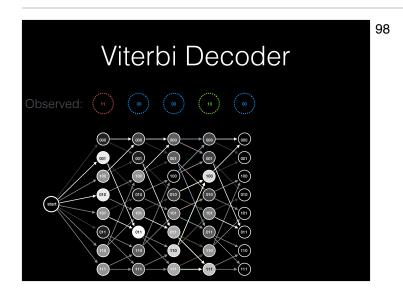


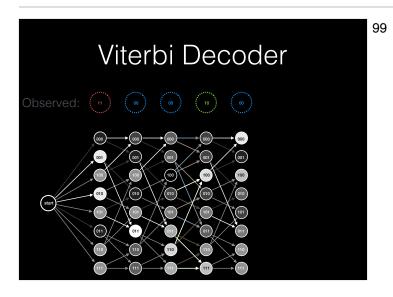


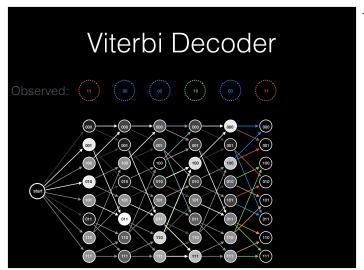


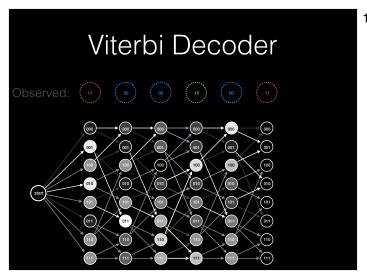


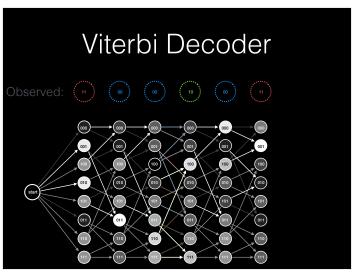


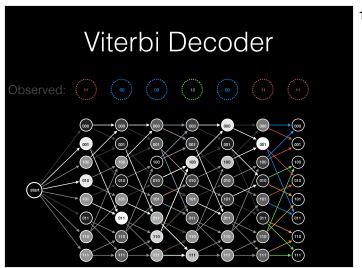


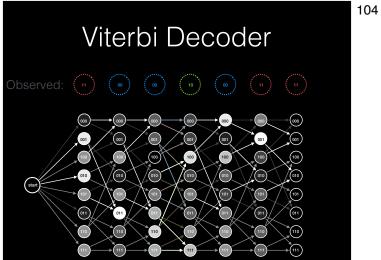


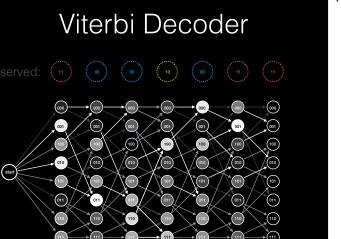


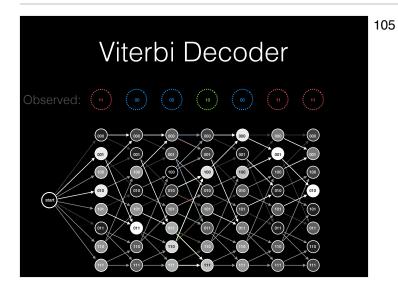




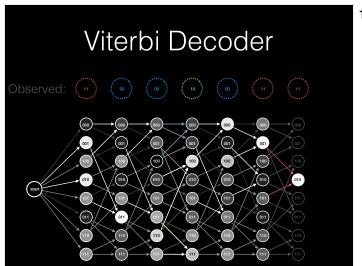


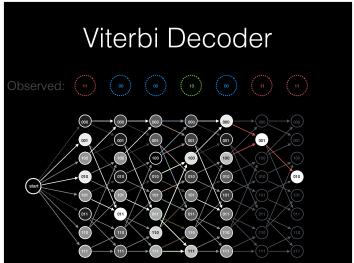


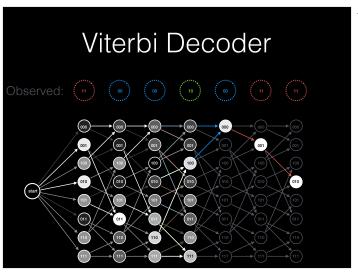


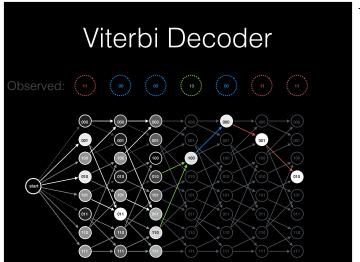


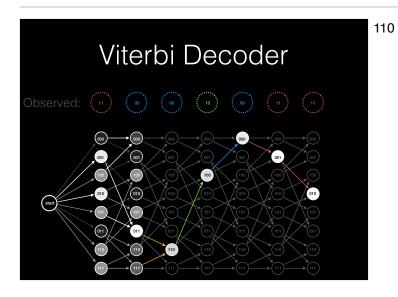
Now our trellis is complete, and we can trace back through it to find the optimal path.

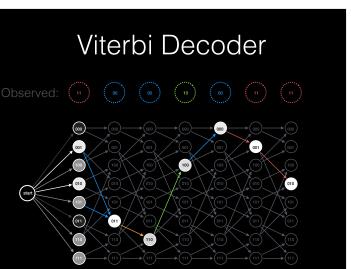


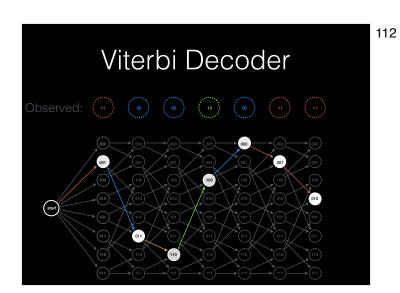




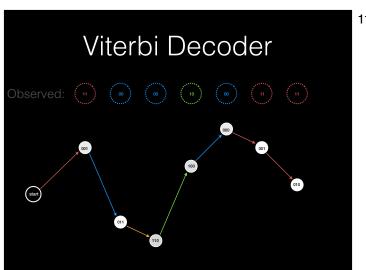


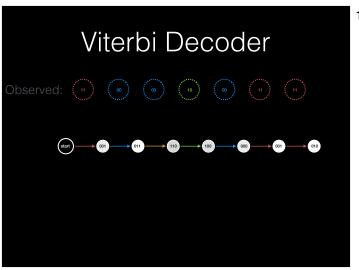


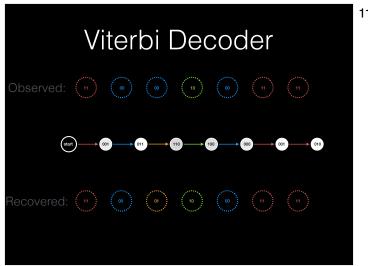


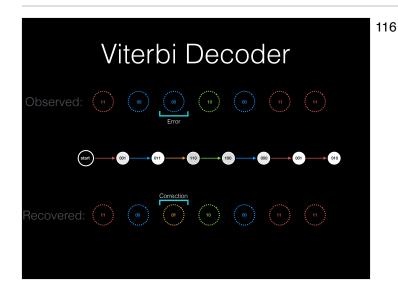


This is the most likely path through the trellis

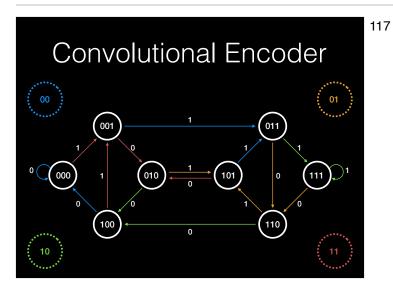




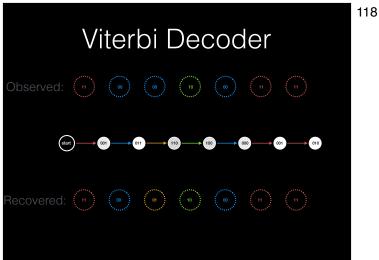


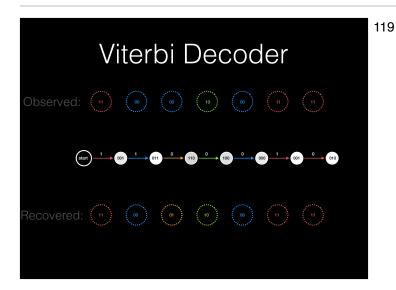


You can see the error correcting part is working. The 3rd observation "00" was most likely caused by a "01" transition instead. Now, each one of these arrows corresponds to a bit from the original message.

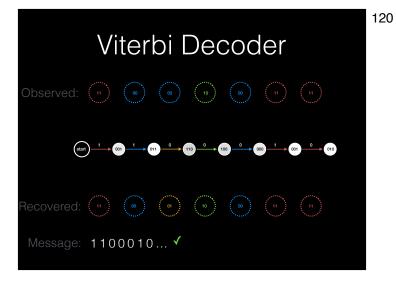


We can go back to the encoder state diagram to see which bit goes with each arrow.

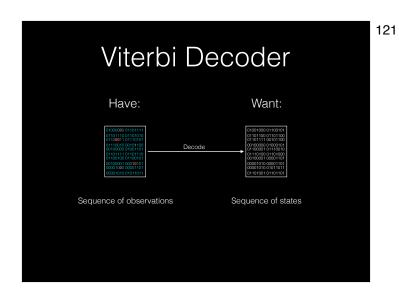




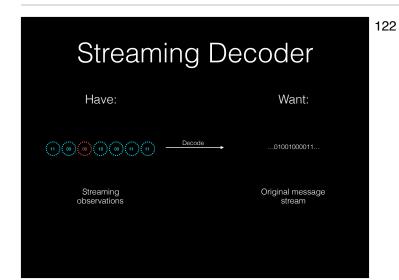
And then we can fill in the arrows to see what our message was.



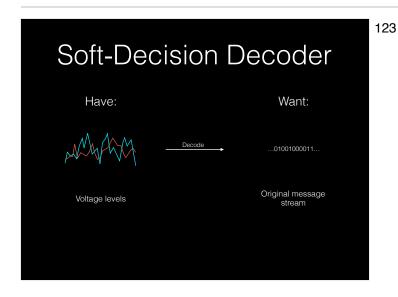
And if you don't remember, this message matches what we originally sent.



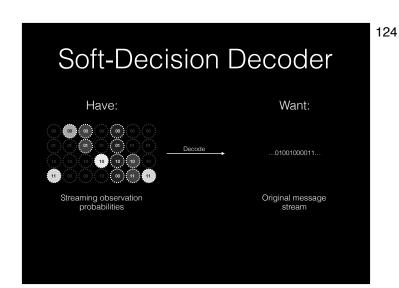
So that's the basic version of the Viterbi Decoder... There are a few variations.



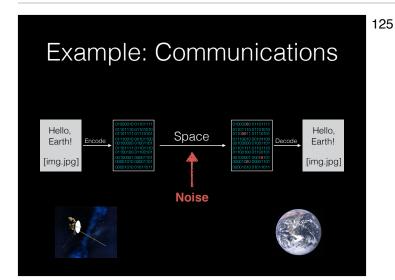
Streaming decoders allow you to do a 1-way streaming link, like for continuous sensor data



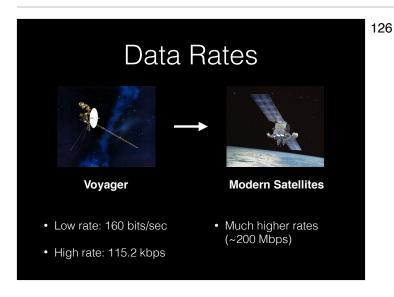
Soft decision decoders are more accurate. Here we have voltage levels from our antenna, and we want our original message stream. We could draw a line through the middle – call everything above it a 1, everything below a 0. (or...)



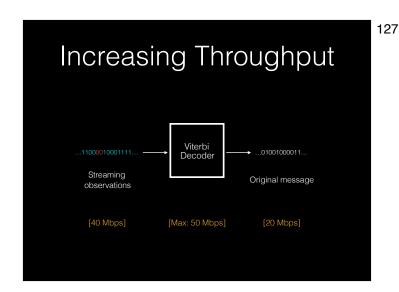
We can represent our observations as probability distributions. This allows you to give observations more weight when they have less uncertainty



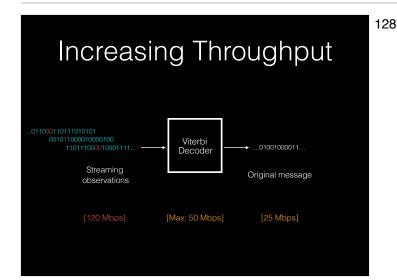
Remember, the goal is to be able to send messages robustly through a noisy channel



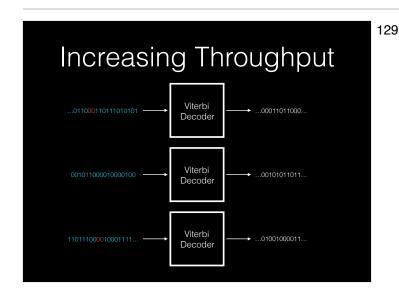
I worked on a project before coming to Duke where we needed to increase the throughput of a Viterbi decoder so we could handle higher data rates.



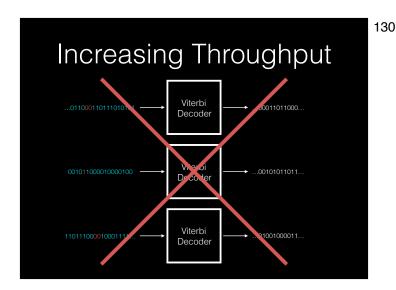
The problem was that our decoder could only handle data rates up to 50 Mbps



The final version needed to sustain several hundred Mbps.



Normally what you'd do in a situation like this is you'd just split the data as it comes in, send to different decoders, and combine the outputs



Increasing Throughput

...01100011011010101 \longrightarrow ...00011011000...

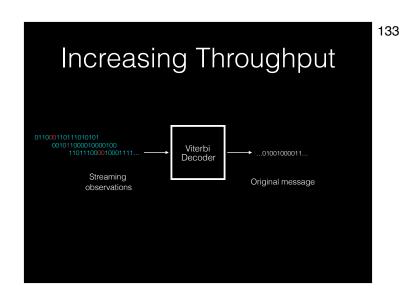
00101100001000100 \longrightarrow ...00101011011...

110111000010001111... \longrightarrow ...01001000011...

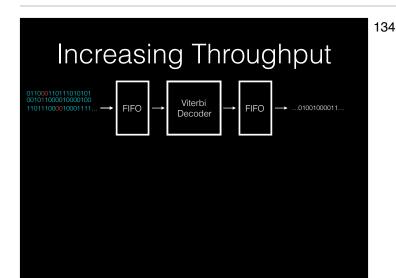
Remember, each decoder needs to keep track of state histories. Splitting up the data won't give you the right results.

Increasing Throughput

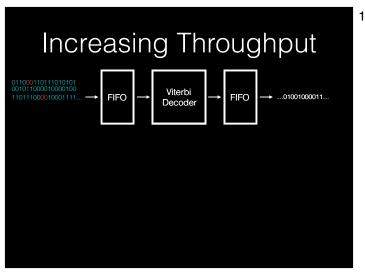
- Common parallelization techniques insufficient
 - Hysteresis
 - Streaming data, not block data
- Performance requirements → soft decisions
 - Published decoders used hard decisions
- · Not enough time to rewrite existing decoder

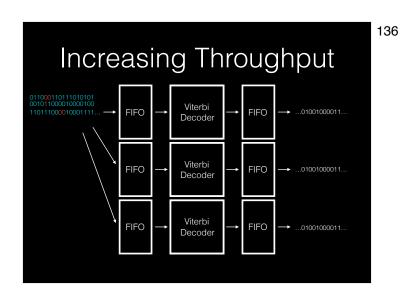


To solve it, we ended up doing something similar to our original plan.

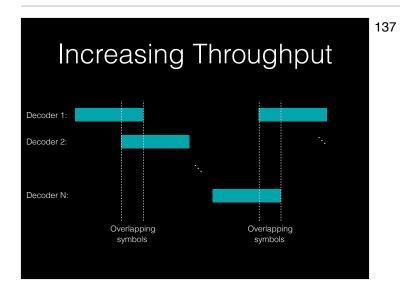


But we added these FIFOs on each end

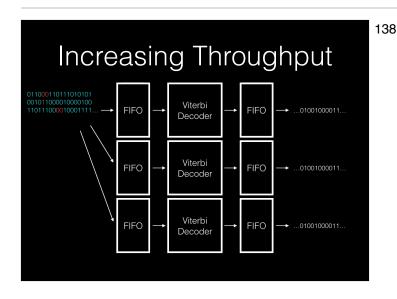




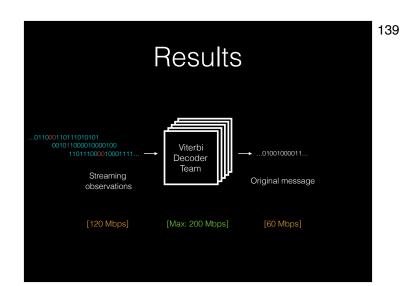
And we used multiple decoders, but we didn't split the data up like we did before...



Instead, I came up with a novel method which sent overlapping, consecutive samples to multiple decoders at same time, and then rotated batches of samples through the decoders in sequence



Doing this effectively initializes each with enough state history information to recover a piece of the original message. By combining the pieces, we could reproduce the entire message



Using this method, we able to meet arbitrarily high throughput requirements, and meet our project deadline, without needing to rewrite our existing decoder

References

• MIT Lecture Notes – Digital Communication Systems

• Lecture 8 – Convolutional Coding

• Lecture 9 – Viterbi Decoding of Convolutional Codes

• Wikipedia

• Viterbi Algorithm

• Viterbi Decoder

• Voyager 1