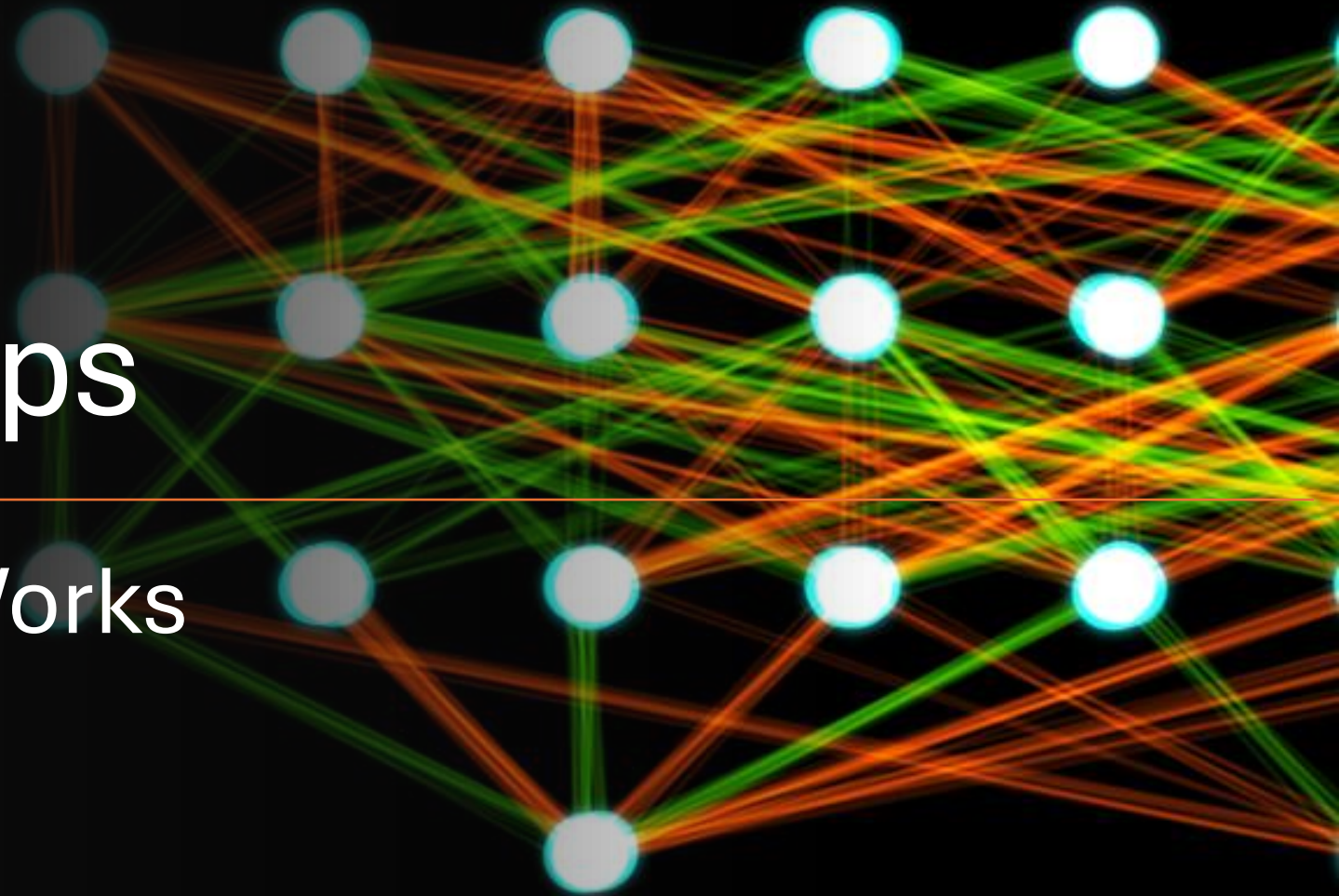


# Autocorrect With Extra Steps

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How Generative AI Works

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# Why Learn How AI Works?

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- Hype & anxiety make it hard to understand its capabilities and limitations
  - Understanding makes it easier to detect and work with
  - The tech and math is awesome
- 



# History of Language Models

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- Research started in 1960s at MIT
  - Early generators used strict rules
  - ELIZA was a simulated therapist – responded to keywords
  - SHRDLU could take limited actions in a virtual world
- 



# Statistical Generators

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- Rule-based systems had obvious patterns & limitations
  - Impossible to write a rule for every situation
  - In the early 90s, research shifted to statistical methods
  - Instead of writing rules, build systems to analyze and mimic existing text
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# Markov Chains

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- Markov Chains are the simplest text generators
- Train it on a source by going word-by-word
- Build a table of what follows each specific word
- Build a sentence by choosing a word, then choosing the next word from the table



# Markov Chain Example

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Test Case: *Green Eggs & Ham* by Dr Seuss

Limited 50-word vocabulary makes a great test case

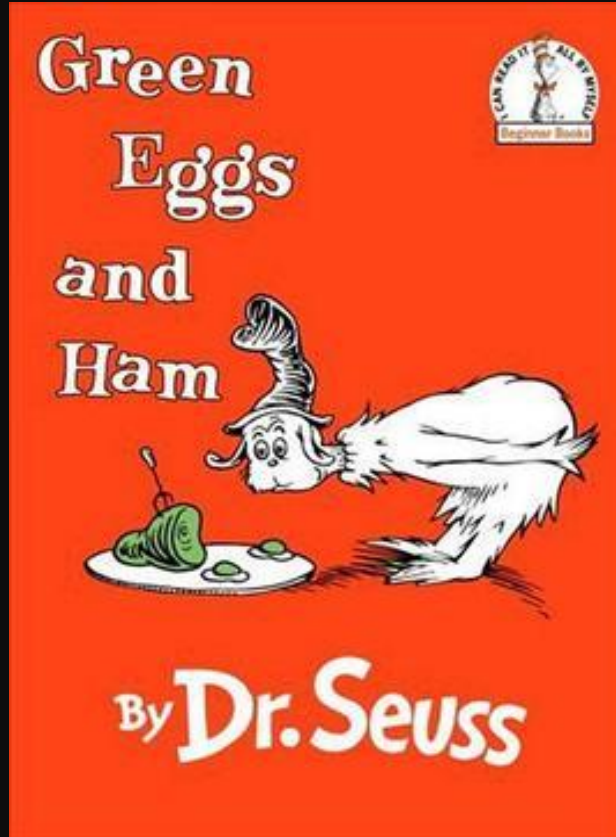


Table ["Green"]

- 100% "Eggs"

Table ["Could"]

- 48.2% "Not"
- 44.4% "You"
- 7.4% "Eat"

Table ["Ham"]

- 100% End Sentence

# Improvements

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## PROBLEMS:

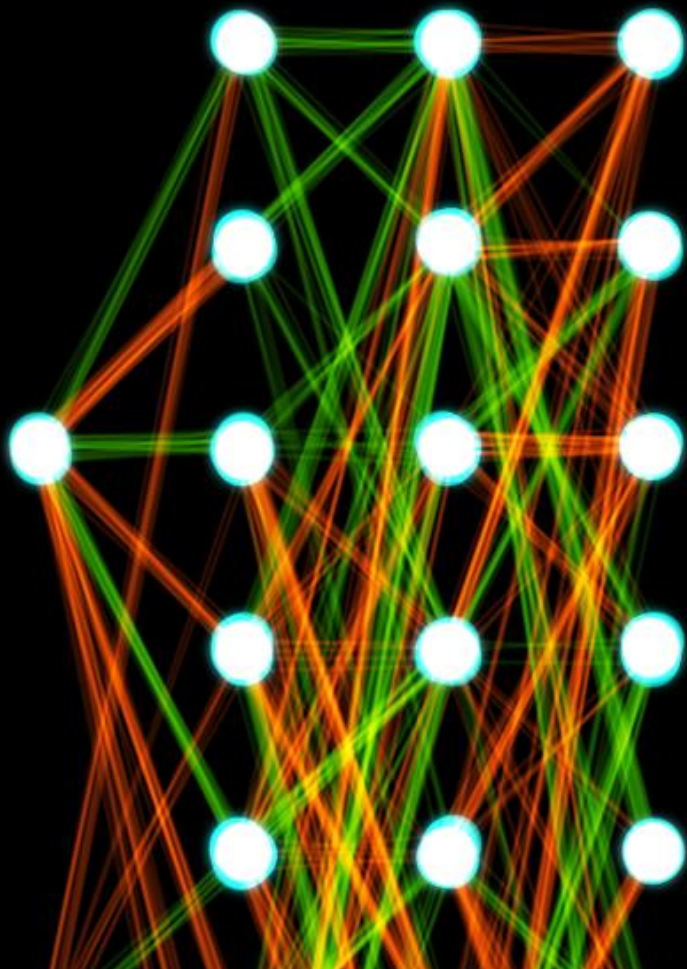
- Can only use word patterns it's seen before
- Common phrases are randomized
- Words have no context

## SOLUTIONS:

- Add more sources
  - Look at multiple words, not just one
    - Ex: "Eggs and" always gives "Ham"
  - Code some manual rules – parts of speech, common phrases
-

# Neural Networks

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- In the early 2000s, researches tried structuring systems like neurons
  - Add more inputs, multiple layers, and random results to the lists
  - Give feedback on what is good output vs bad output
  - Let it tweak its own rules based on feedback
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# Large Language Models (Chat-GPT)

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- Adds memory, attention, and context
  - Context - Looks at prompt to limit data to the topic
  - Attention - Focuses on relevant keywords
  - Memory – Adjusts word probabilities as it goes for consistency
- 



# The *Large* in LLM

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- LLMs require a huge amount of data to pretrain their neural networks
  - Chat-GPT 4 cost >\$100 million to train
  - Training data sets can reach 500TB
  - LLMs use the same energy as 15,000 – 30,000 U.S. household each day
- 



# What Does All This Mean?

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- Each word is calculated by massive amounts of math calculated from massive amounts of examples
  - Results become based on averages, giving safe results
  - Does not have actual creativity or reasoning  
(Only remixes from existing works or examples)
  - Will prioritize things that look right instead of being right  
(Citations are especially prone to this)
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# Valid Uses

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- Search engine for common information
- Inspiration
- An outline of an essay
- A common consensus or outline of a topic

# Bad Uses

- Searching for specific information or jargon
  - Creative work
  - An uncommon argument
  - Anytime you need 100% accurate information
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# Detecting AI Works

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- Arguments are "safe". Gives both sides instead of persuading
  - Padding phrases: "delve into", "a tapestry of", "in the realm of"
  - For art, fine details are missing or excessive. Look at hands & feet
  - For photos, the scene may be too perfect
- 





# Conclusions

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- Modern AI generators use tons of data and math to create content
  - Fine for inspiration or searching for common, non-critical info
  - Focuses on looking right over being right
- 



# Questions?

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