

Circuit-Level Metropolis Algorithm

① Start w/ a set of errors \vec{E} which is known to cause a logical error

② Select a (gate, fault) pair at random from the circuit, (g, f)

Initially:
MC sampling at $p=10^{-3}$:
→ Logical error rate
→ And initial sample \vec{E}

②a If g is not an element of \vec{E} , then set:

$$\vec{E}' = \vec{E} \cup \{(g, f)\}$$

$$Pr(\vec{E}' | \vec{E}) = \frac{Pr(g)}{1 - Pr(g)} Pr_g(f)$$

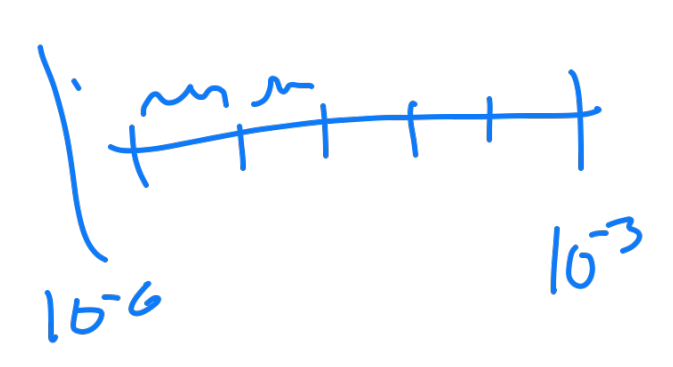
Definitions

① $Pr(g) :=$ probability that the gate g has some failure

②b If $\exists (g, h) \in \vec{E}$ for some h , then set:

$$\vec{E}' = \{\vec{E} \cup \{(g, f)\}\} \setminus \{(g, h)\}$$

② $Pr_g(f) :=$ probability that, given g has failed, the failure corresponds to f .



(a) $f = h \Rightarrow$ we've effectively removed an element from \vec{E}

$$\Rightarrow Pr(\vec{E}' | \vec{E}) = 1$$

(b) $f \neq h$

$$\Rightarrow Pr(\vec{E}' | \vec{E}) = Pr_g(f)$$

Circuit (noiseless)

Noise Model (circuit)

Noisy version ckt

Through this process, each \vec{E}' that gets generated appears as though it's a sample from the $\pi_{i|F}$ distribution (eventually)

What we're trying to find is:

$$r = \frac{\bar{P}_{i+1}}{\bar{P}_i} = \frac{\pi_{i+1}(F)}{\pi_i(F)} = C \frac{E_{i|F} \left[g \left(C \frac{\pi_i}{\pi_{i+1}} \right) \right]}{E_{i+1|F} \left[g \left(C \frac{\pi_{i+1}}{\pi_i} \right) \right]} = \frac{\frac{1}{N} \sum g(C)}{\frac{1}{N} \sum g(C)}$$

$\bar{P}_i = r \bar{P}_0$

How can we estimate the empirical average if this quantity is unknown?

We approximate the numerator & denominator using empirical averages:

$$E_{i|F} \left[g \left(C \frac{\pi_i}{\pi_{i+1}} \right) \right] \approx E_{i|F}^{est} \left[g \left(C \frac{\pi_i}{\pi_{i+1}} \right) \right] := \frac{1}{N} \sum_{j=1}^N g \left(C \frac{\pi_i(E_j)}{\pi_{i+1}(E_j)} \right)$$

where $E_1, \dots, E_N \in F$ drawn from $\pi_{i|F}$, i.e., the samples obtained from running the Metropolis algorithm. → Have to run MCMC some number of times until convergence, and then take N samples

These terms are known from the ckt probs.
This is given from the paper as $\frac{1}{1+x}$

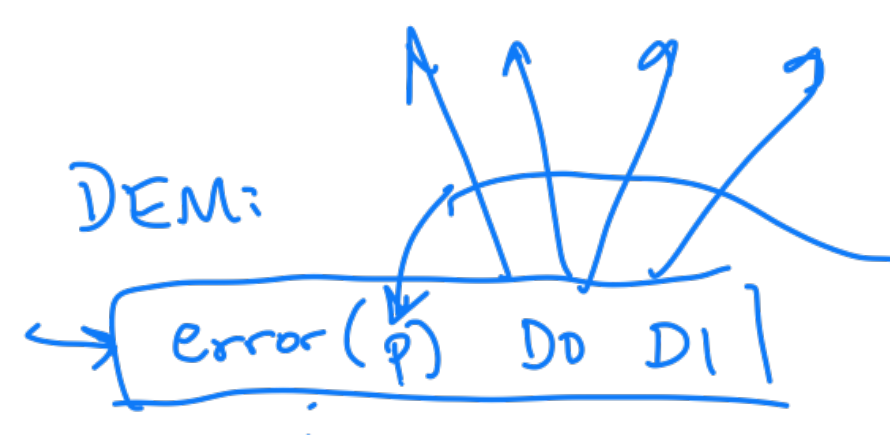
$$g(x) = \frac{1}{1+x}$$

$$= \frac{1}{1+C()}$$

From these estimates, we want to find C s.t:

$$E_{i|F}^{est} \left[g \left(C \frac{\pi_i}{\pi_{i+1}} \right) \right] = E_{i+1|F}^{est} \left[g \left(C \frac{\pi_{i+1}}{\pi_i} \right) \right]$$

$$\Rightarrow \frac{\bar{P}_{i+1}}{\bar{P}_i} \approx C \frac{E_{i|F}^{est} \left[g \left(C \frac{\pi_i}{\pi_{i+1}} \right) \right]}{E_{i+1|F}^{est} \left[g \left(C \frac{\pi_{i+1}}{\pi_i} \right) \right]} = C \cdot 1 = C$$



error(p) DN-1 DN
0, error() [detectors]
N error(p) [detectors]

stim. circuit, explain

[Error 1: → DEPOLARIZE(1/2)(p)
Error 2
Error N:]

gate⁰, x
gate⁰, z