# A criticality approach to monitoring cascading failure risk and failure propagation in transmission systems

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### Levels of modeling for cascading blackouts

#### ANALYTIC MODELS

(branching process, CASCADE)

### **SIMULATIONS**

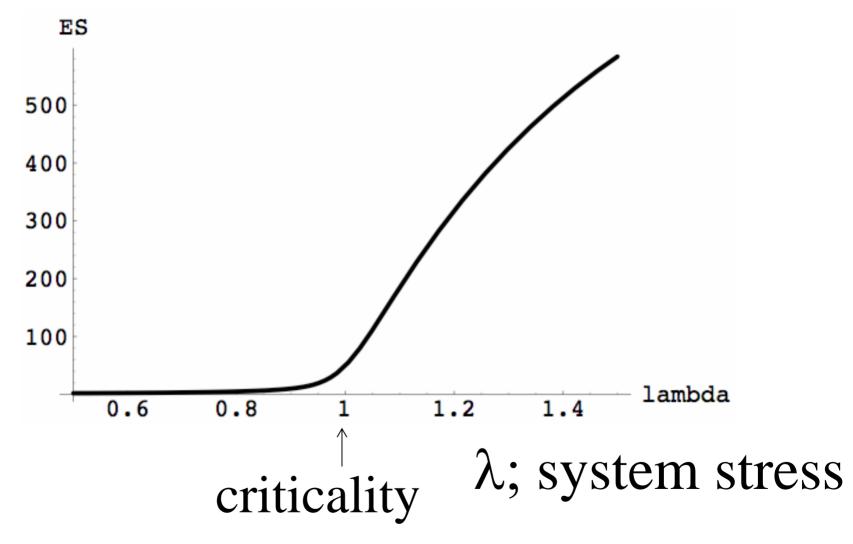
(OPA, hidden failure, Kirschen, CMU, TRELSS)

REAL BLACKOUTS

The probability of cascading failure blackout increases with system loading

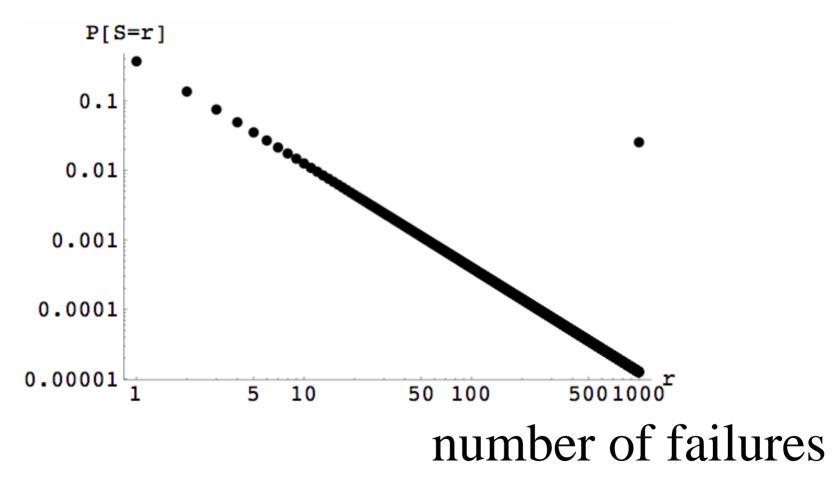
Exactly how does it increase?

### Mean number of failures

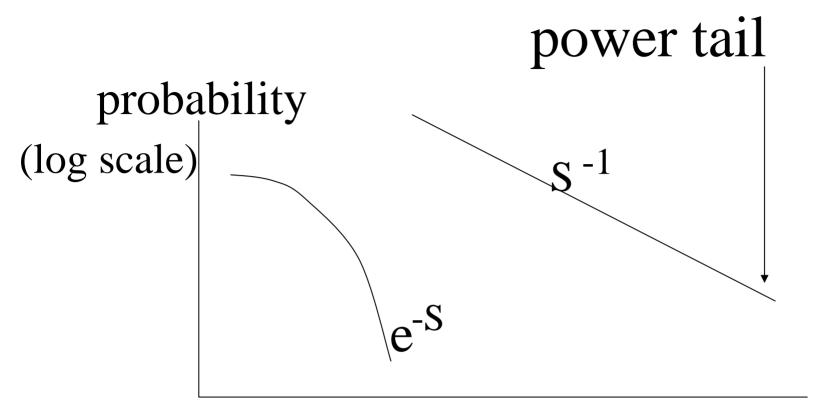


Branching model with 1000 components

# probability



Branching model with 1000 components at criticality



blackout size S (log scale)
power tails have huge impact
on large blackout risk.

risk = probability x cost

### a working definition of critical loading

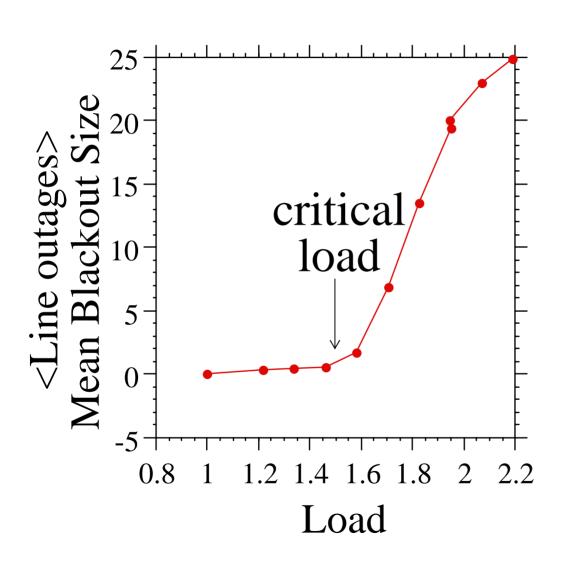
# at criticality:

- there is a power law in the distribution of blackout size
- there is a kink or jump in mean blackout size

# Summary of OPA blackout model

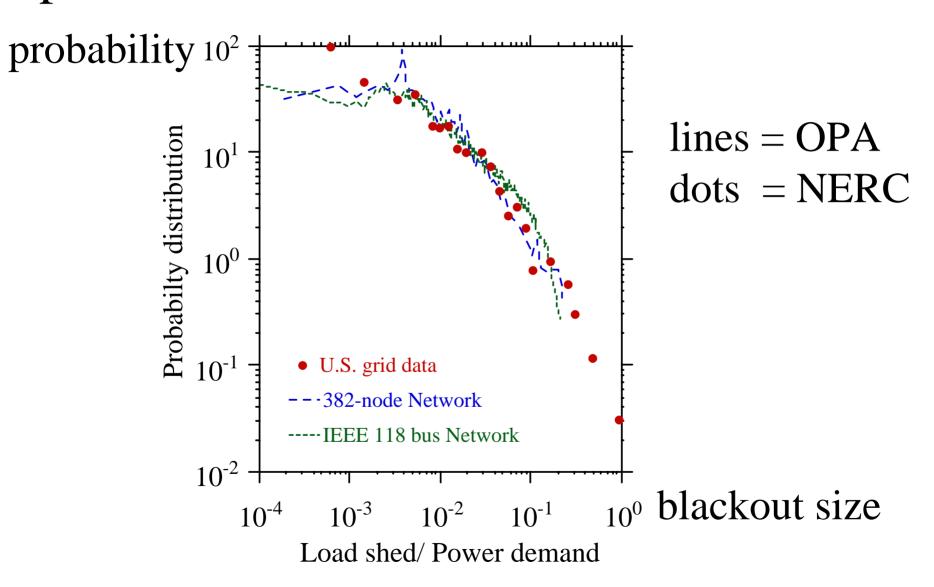
- transmission system network modeled with DC load flow and LP dispatch
- random initial disturbances and probabilistic cascading line outages and overloads

### Critical loading in OPA blackout model



Mean blackout size sharply increases at critical loading; increasing risk of cascading failure.

### power laws in OPA model & NERC data



How does the probability distribution of blackout size change as system is loaded?

Don't think about avoiding all blackouts; think about managing the probabilities of small, medium and large blackouts

# Effect of Loading

log log plots

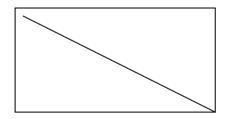
probability

- VERY LOW LOAD
  - failures independent
  - exponential tails

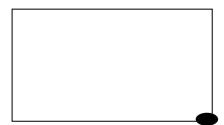


blackout size

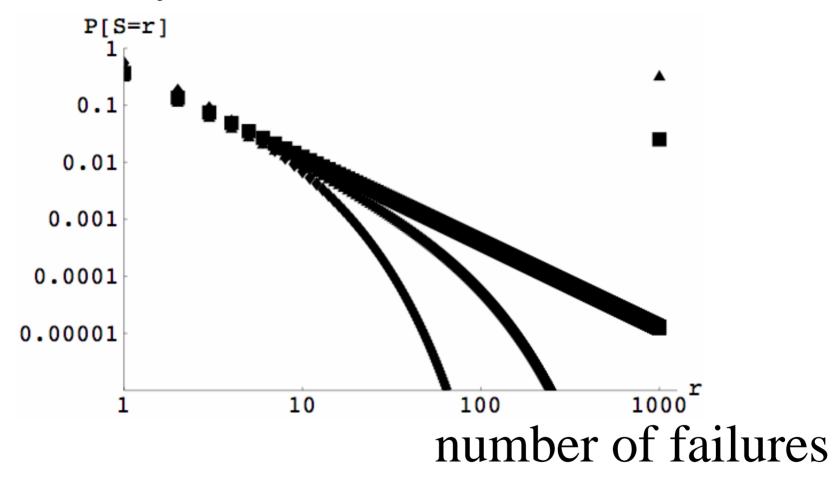
- CRITICAL LOAD
  - power tails



- VERY HIGH LOAD
  - total blackout likely

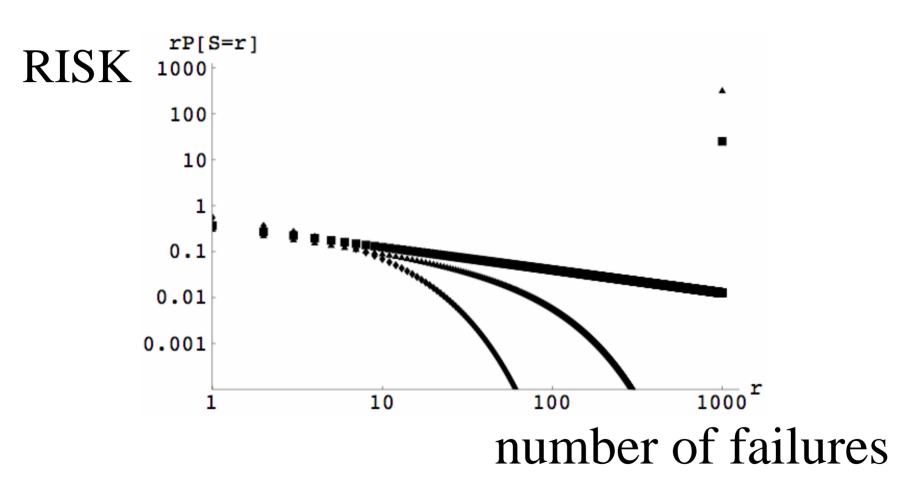


## probability



Branching model with 1000 components below, at, and above criticality

# By assuming cost proportional to size, can obtain risk distribution from pdf



Branching model with 1000 components below, at, and above criticality

# Significance of criticality

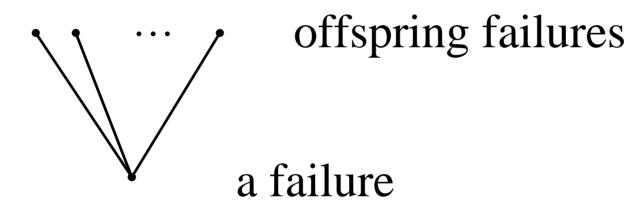
- At criticality there is a power tail, sharp increase in mean blackout size, and an increasing risk of cascading failure -- signatures of a phase transition in statistical physics.
- Criticality gives a power system limit or reference point with respect to cascading failure.
- Margin to criticality can be related to blackout risk
- How do we practically monitor or measure margin to criticality?

# approaches to margin to criticality

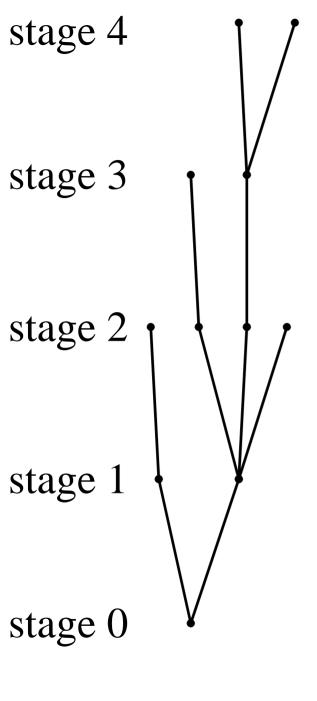
- 1) Increase loading in blackout simulation until average blackout size increases.
- 2) Monitor or measure how much failures propagate ( $\lambda$ ) from real or simulated data.

### Summary of branching process model

Branching from one failure:



number of offspring ~ Poisson( $\lambda$ ) mean number of failures =  $\lambda$ 



# **Branching Process**

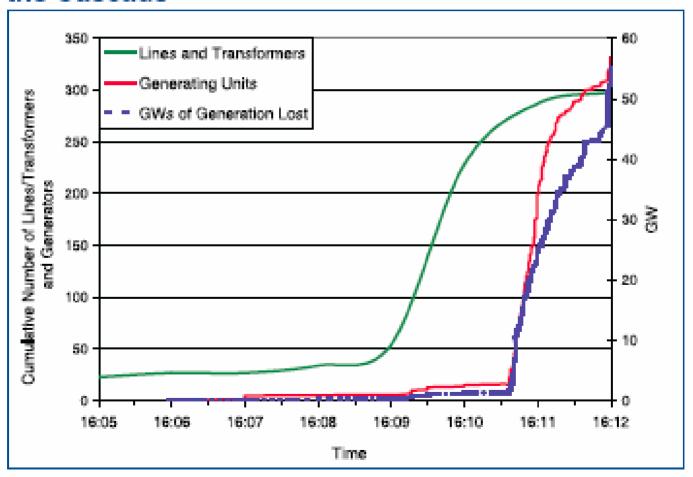
- each failure independently has random number of offspring in next stage according to  $Poisson(\lambda)$
- $-\lambda$  = mean number failures per previous stage failure
- $\lambda^k$  = mean number of failures in stage k

# λ controls failure propagation

- Subcritical case  $\lambda$ <1: failures die out
- Critical case  $\lambda=1$ : probability distribution of total number of failures has power tail
- Supercritical case  $\lambda > 1$ : failures can proceed to system size

### Cumulative Line Trips from August 2003 Blackout Final Report

Figure 6.1. Rate of Line and Generator Trips During the Cascade



### Estimates of $\lambda$

- (number of failures in stage k) 1/k
- (number of offspring)/(number of parents)
- slope of log(cumulative failures)

We are starting to explore the use of these statistical estimates in this and other papers

### Conclusions

- Criticality gives a reference point or system limit with respect to increasing risk of cascading failure.
- Margin to criticality and failure propagation  $\lambda$  can be related to distribution of risk of blackouts of all sizes. Goal: risk analysis for blackouts
- Analytic and simulation models and monitoring methods for cascading failure are emerging.
   One Goal: understand and quantify criticality and blackout risk from real and simulated data.

# Our general approach

- Instead of only looking at postmortem details of particular blackouts, also look at bulk system statistical properties (global, top-down).
- Instead of avoiding all blackouts, shape the pdf of blackout size to manage risk of all blackout sizes.
- Design and operate system to limit propagation of failures after they are initiated; complementary to limiting likely and foreseeable combinations of initial failures.
- Add an explicit operating limit for increased risk of cascading failure.
- For detail see papers at http://eceserv0.ece.wisc.edu/~dobson/home.html

### Research directions

- Gain research access to blackout data
- Estimate costs of large blackouts
- Confirm and study criticality
- Further develop power system and abstract models of cascading failure
- Develop practical methods of monitoring margin to criticality from real or simulated data