

A baseline flare prediction using only event statistics

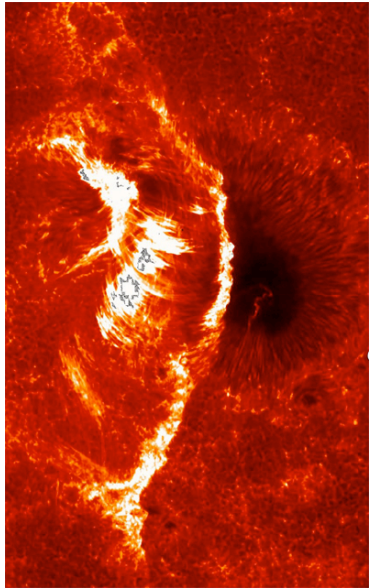
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Forecasting the Operational All-Clear
22-24 April 2009



The University of Sydney



Overview

Method

Persistence

Flare statistics

Event statistics method

Workshop applications

Whole-Sun, active-region prediction of GOES flares

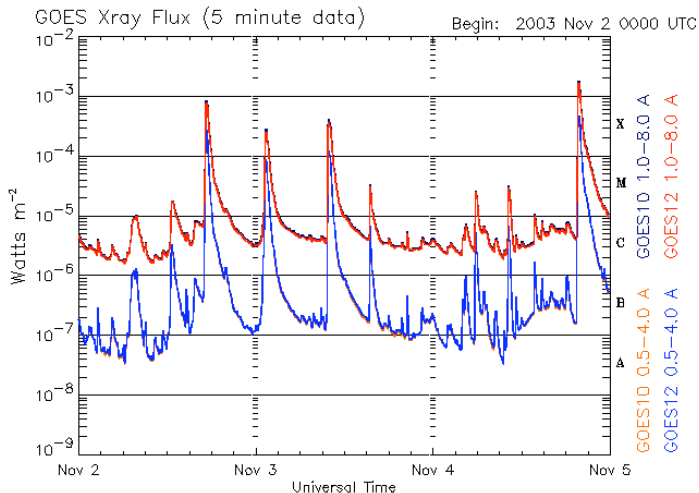
Validation

Accuracy of probabilistic forecasts

Summary and comments

Persistence

- Past flaring history is an indicator of future flaring



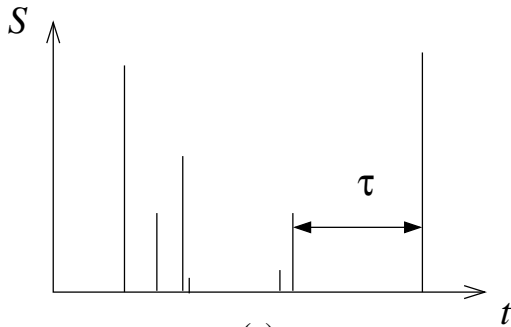
Flare statistics

- ▶ Flares obey a power-law frequency-size distribution (Drake 1971)

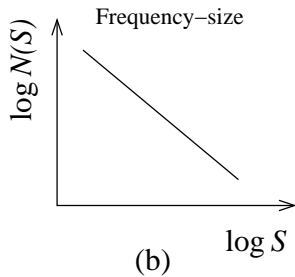
$$N(S) = \lambda_1(\gamma - 1)S_1^{\gamma-1}S^{-\gamma} \quad (1)$$

- ▶ size S : energy, or peak flux in X-ray,...
 - ▶ $N(S)$ is number of flares per unit time, per unit S
 - ▶ $\gamma = \frac{3}{2}$ to $\gamma = 2$ (depends on specific choice of S)
 - ▶ $\lambda_1 = \lambda_1(t)$ is total rate above size S_1
- ▶ Occurrence in time may be modelled as a Poisson process (e.g. Wheatland 2001)
 - ▶ if λ does not vary, distribution of waiting times τ :

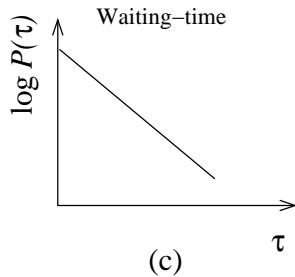
$$P(\tau) = \lambda \exp(-\lambda\tau) \quad (2)$$



(a)



(b)



(c)

Event statistics method (Wheatland 2004, ApJ 609, 1134)

- ▶ S_1 =size of “small” event, S_2 =size of “big” event
- ▶ λ_1 =observed rate above S_1 ; PL size distribution \Rightarrow

$$\lambda_2 = \lambda_1 (S_1/S_2)^{\gamma-1} \quad (3)$$

- ▶ even if no big events have been observed
- ▶ Probability of at least one big event in time T_P is

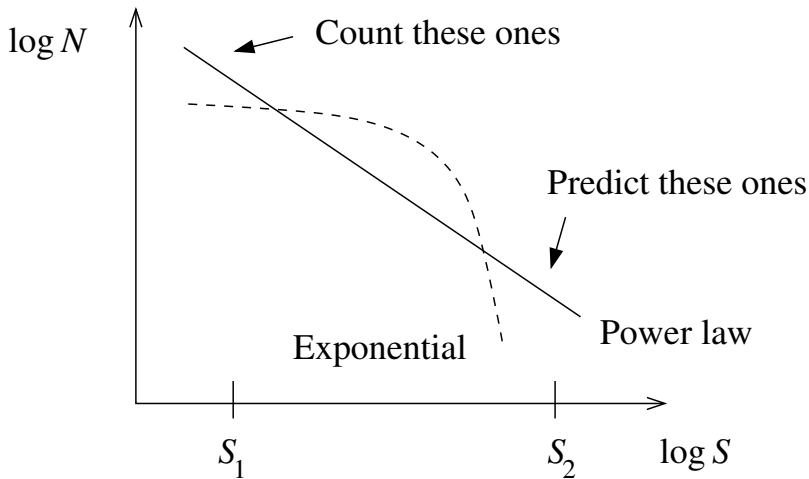
$$\begin{aligned} \epsilon &= 1 - \exp(-\lambda_2 T_P) \\ &= 1 - \exp \left[-\lambda_1 (S_1/S_2)^{\gamma-1} T_P \right] \end{aligned} \quad (4)$$

- ▶ assuming Poisson waiting times
- ▶ If M events are involved in inferring λ_1 then

$$\sigma_\epsilon/\epsilon \approx 1/\sqrt{M} \quad (5)$$

- ▶ accurate if many small events observed

- Method well-suited to a power-law size distribution



Bayesian version

- ▶ Data D : events s_1, s_2, \dots, s_M at times $t_1 < t_2 < \dots < t_M$
- ▶ Infer $P_\gamma(\gamma|D)$ and $P_1(\lambda_1|D)$
- ▶ Calculate

$$\begin{aligned} P_2(\lambda_2|D) &= \int_1^\infty d\gamma \int_0^\infty d\lambda_1 P_1(\lambda_1|D) P_\gamma(\gamma|D) \\ &\times \delta[\lambda_2 - \lambda_1 (S_1/S_2)^{\gamma-1}] \end{aligned} \quad (6)$$

and then

$$P_\epsilon(\epsilon|D) = P_2[\lambda_2(\epsilon)|D] \left| \frac{d\lambda_2}{d\epsilon} \right| \quad (7)$$

where

$$\lambda_2(\epsilon) = -\ln(1 - \epsilon)/T_P \quad (8)$$

- ▶ Problem: infer power-law index and rate of small events

Inferring the power-law index γ

- ▶ Events power-law distributed, so (Bai 1993)

$$P(D|\gamma) \propto \prod_{i=1}^M (\gamma - 1)(s_i/S_1)^{-\gamma} \quad (9)$$

- ▶ uniform prior over a range $\gamma_1 < \gamma < \gamma_2$ used
- ▶ for $M \gg 1$, peaked around

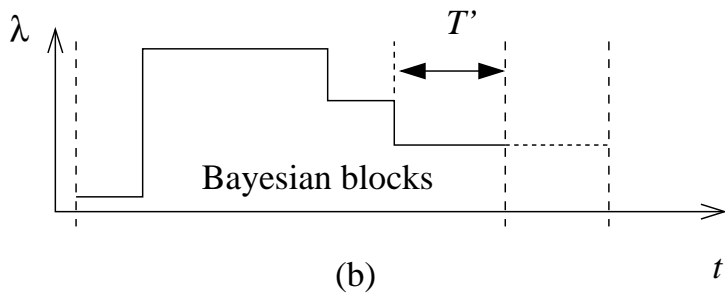
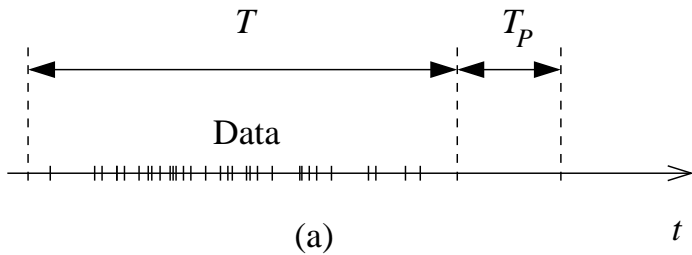
$$\gamma_{\text{ML}} = \frac{M}{\ln \pi} + 1 \quad \text{where} \quad \pi = \prod_{i=1}^M \frac{s_i}{S_1} \quad (10)$$

- ▶ use $P_\gamma(\gamma|D) = \delta(\gamma - \gamma_{\text{ML}})$

Inferring the rate of small events λ_1

- ▶ Complicated by time variation
- ▶ Bayesian blocks used (Scargle 1998)
 - ▶ iterative comparison of one- versus two-rate Poisson models
 - ▶ decomposition into piecewise-constant Poisson process
 - ▶ data D' in last piece (block): M' events in time T'
 - ▶ most recent interval with constant rate

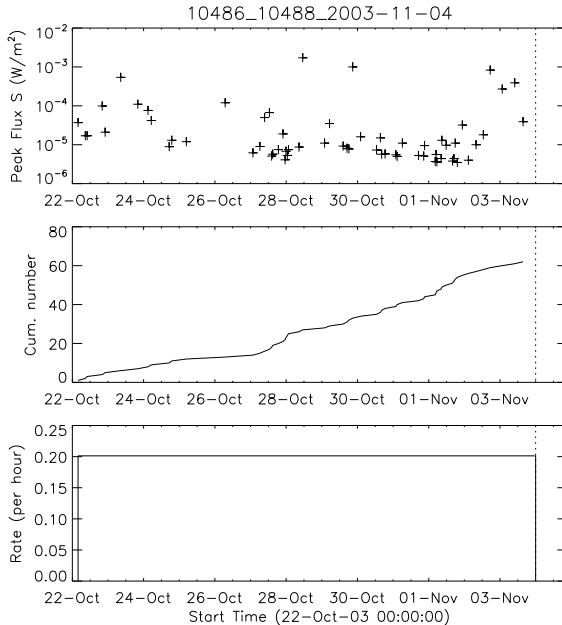
$$P_1(D'|\lambda_1) \propto \lambda_1^{M'} e^{-\lambda_1 T'} \quad (11)$$



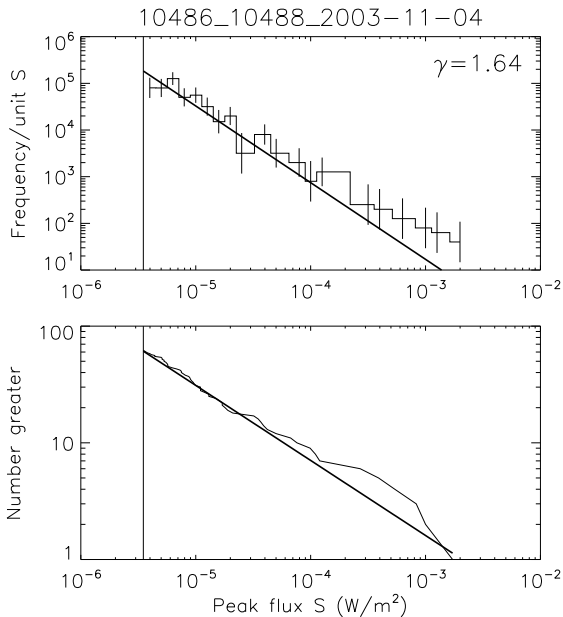
Whole-Sun, active-region prediction of GOES flares

- ▶ Method applied to to whole-Sun prediction of GOES events (Wheatland 2005, Space Weather 3, S07003)
 - ▶ this workshop: predictions for times of MDI observations
 - ▶ magnetogram information not used
- ▶ Applied here also to specified active regions
- ▶ Illustration for 10486 and 10488
 - ▶ largest soft X-ray flare: 4 Nov 2003 20:06UT in 10486
 - ▶ 10486, 10488: flare-productive active regions on disk
 - ▶ D : previous events from these active regions (62 events)
 - ▶ $S_1 = 3.6 \times 10^{-6} \text{ W m}^{-2}$ (smallest event)
 - ▶ inference of γ : ML approximation not made
 - ▶ predictions for C1.0 (24 hr), and M1.0, M5.0 (12 hr) at 00:00UT

► Bayesian blocks procedure



► Frequency-size distribution of events



► Posteriors for M1.0, M5.0

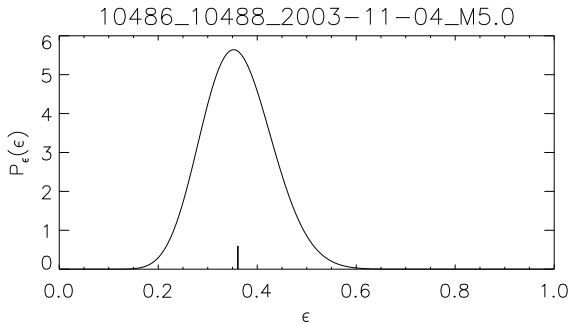
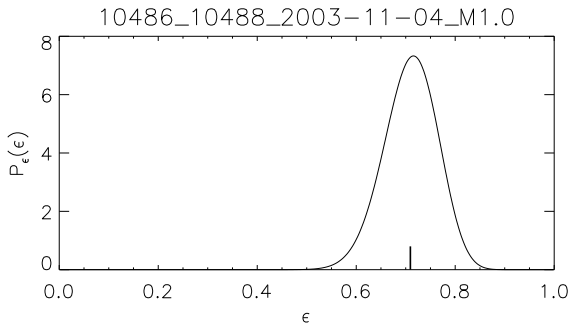


Table: Predictions for 10486, 10488 at 4 Nov 2003 00:00UT

Event	T_P	ε
C1	24 hrs	1.00 ± 0.00
M1	12 hrs	0.71 ± 0.05
M5	12 hrs	0.36 ± 0.07

- ▶ In fact five events \geq C1 within 24 hrs, one \geq M1, zero \geq M5 within 12 hrs
- ▶ Workshop: applied to ARs listed in each MDI file
 - ▶ magnetogram information not used

Accuracy of probabilistic forecasts

- ▶ f = forecast, x = observation (0 or 1)
- ▶ Mean square error

$$\text{MSE}(f, x) = \langle (f - x)^2 \rangle \quad (12)$$

- ▶ “Climatological skill score”:

$$\text{SS}(f, x) = 1 - \text{MSE}(f, x) / \text{MSE}(\langle x \rangle, x) \quad (13)$$

- ▶ improvement over forecasting the average
- ▶ Reliability plots
 - ▶ observed frequencies versus forecast probabilities

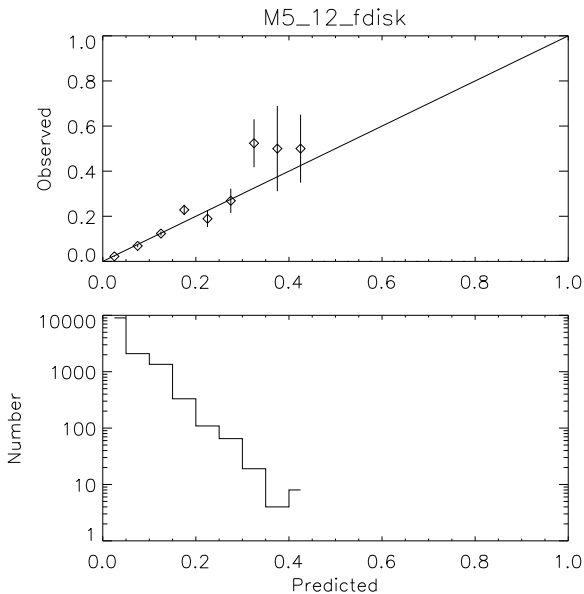
Table: Whole-Sun predictions for workshop

Event	T_P	$\langle f \rangle$	$\langle x \rangle$	$SS(f, x)$
C1	24 hrs	0.94	0.90	0.13
M1	12 hrs	0.28	0.23	0.16
M5	12 hrs	0.049	0.049	0.064

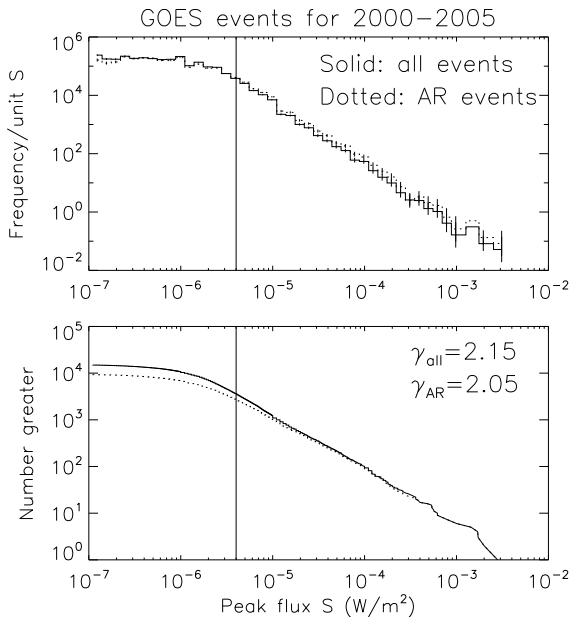
Table: Active-region predictions for workshop (≥ 10 events)

Event	T_P	$\langle f \rangle$	$\langle x \rangle$	$SS(f, x)$
C1	24 hrs	0.89	0.57	-0.27
M1	12 hrs	0.23	0.15	0.088
M5	12 hrs	0.0835	0.046	-0.022

► Reliability plot for whole-Sun, M5–12 hr prediction



► Problem of missed small events



Summary and comments

- ▶ Bayesian method of solar flare prediction
 - ▶ uses only event statistics
 - ▶ exploits power-law size distribution
 - ▶ quantifies “persistence”
 - ▶ applications to whole-Sun, AR prediction of GOES events
 - ▶ provides “baseline” predictions: does not use the MDI data
 - ▶ very simple to apply
- ▶ Method could be improved by:
 - ▶ more complete observations of small events
 - ▶ use of data less affected by event selection problems
 - ▶ inclusion of additional information (e.g. Wheatland 2006)