

Statistical tests for evaluating predictability experiments in Japan

Jeremy Douglas Zechar

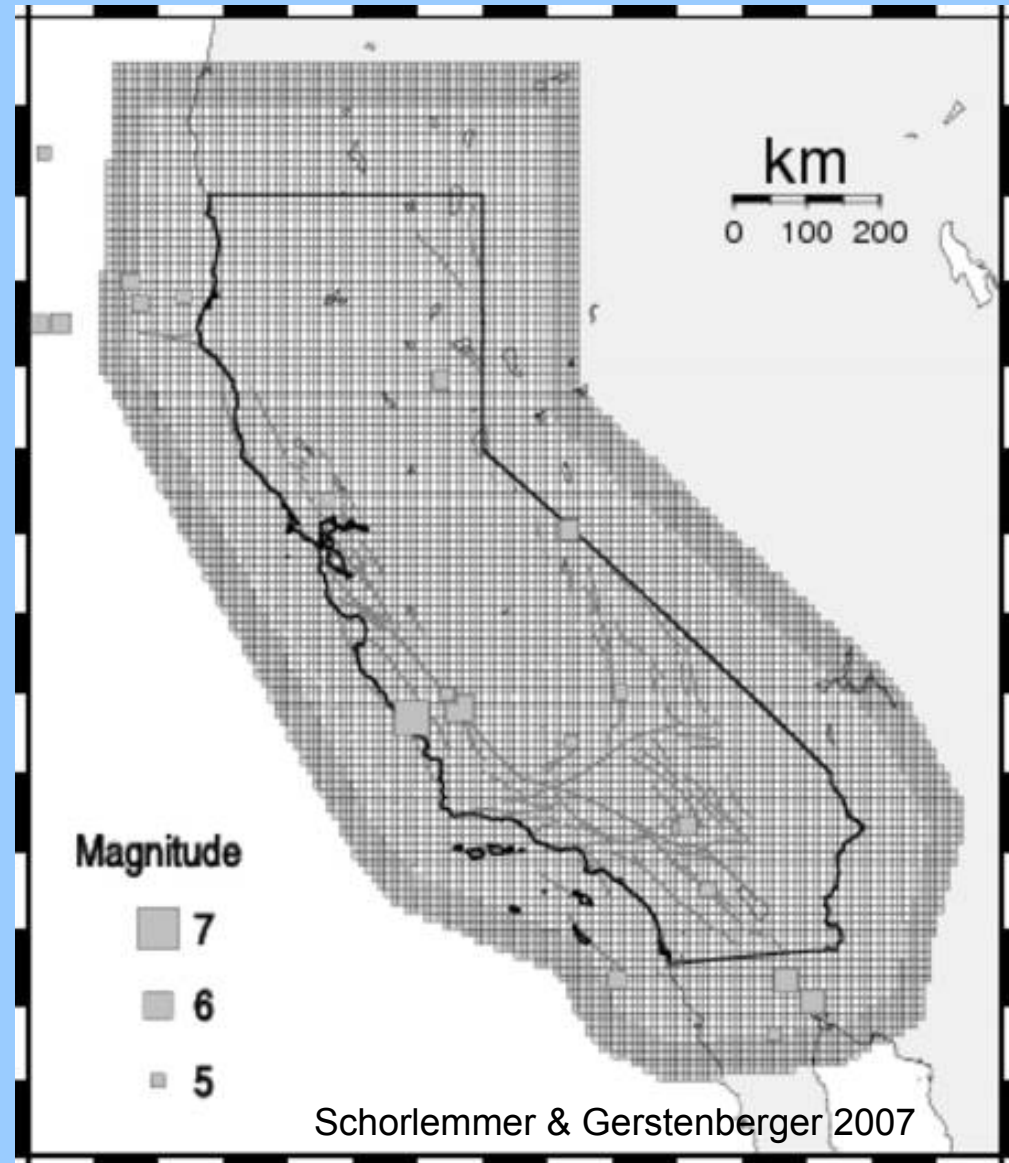
Lamont-Doherty Earth Observatory of Columbia University

Outline

- Likelihood tests, inherited from RELM
- Post-RELM tests
 - Recent extension of likelihood tests
 - Receiver Operating Characteristic (ROC)
 - Molchan error diagram
 - Area skill score
- Process for implementing test

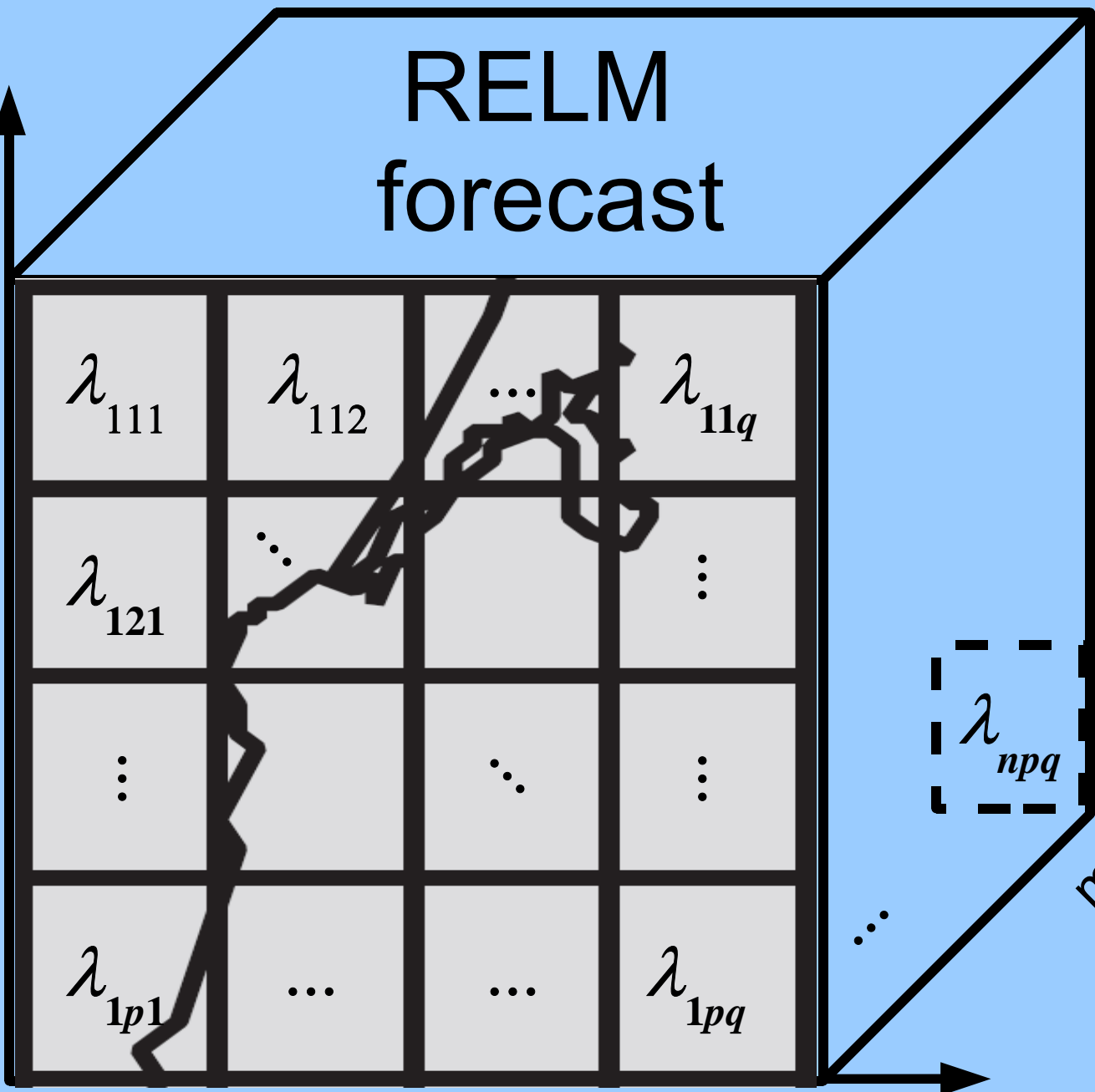
Regional Earthquake Likelihood Model (RELM) experiment in California

- Nineteen 5-year forecasts
 - Target eqks $M_{\text{ANSS}} \geq 4.95$
- Forecasts are specified in terms of expected number of earthquakes in lat/lon/mag bins.
- Currently being tested within CSEP
- Forecasts are evaluated for consistency with observations using likelihood tests.
- Similar format as current prototype testing center in Japan



RELM forecast

latitude



$$\Lambda =$$

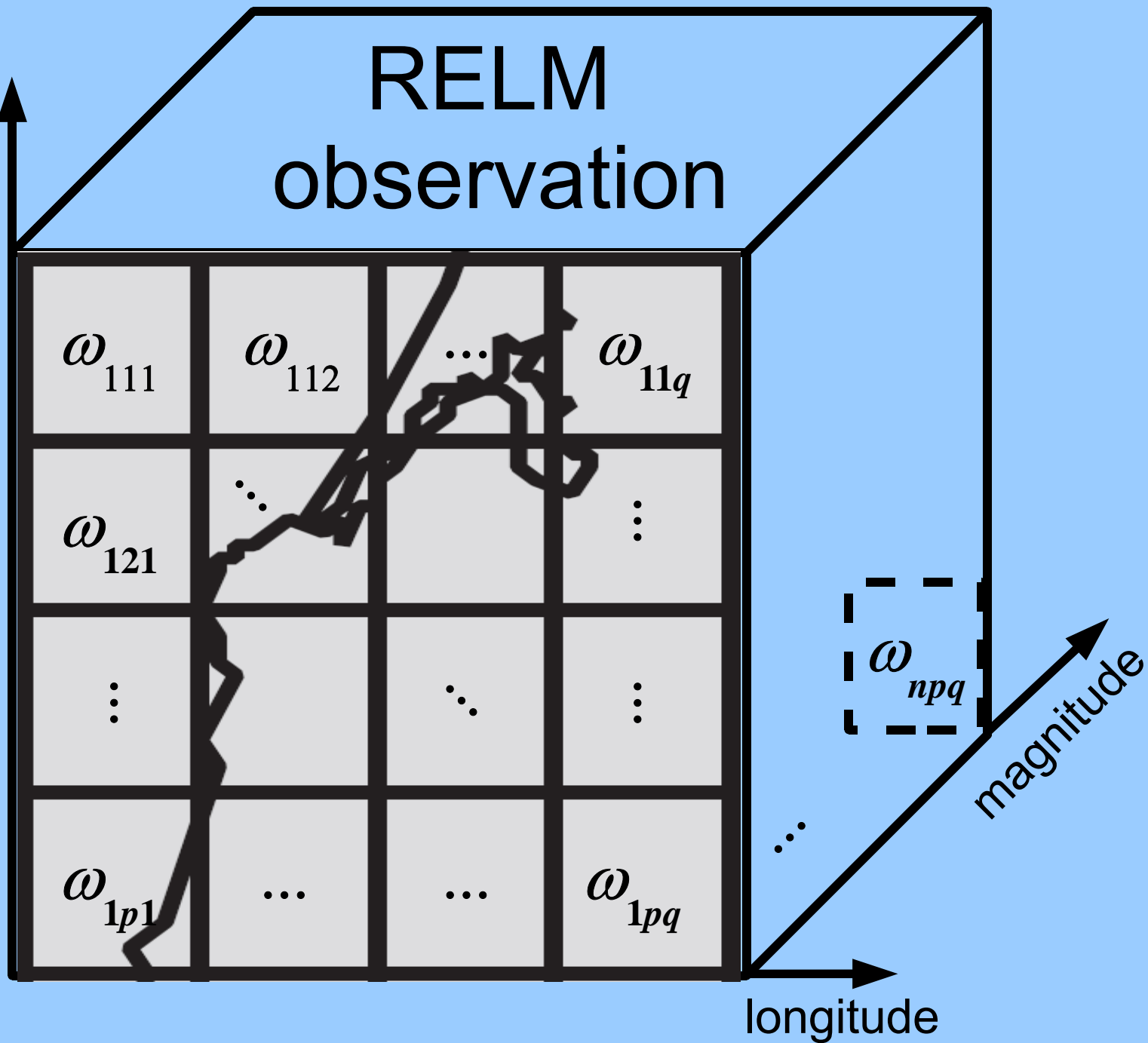
longitude

magnitude

RELM observation

latitude

$\Omega =$



RELM likelihood statistics

Number of earthquakes forecast

$$N_{\Lambda} = \sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q \lambda_{ijk}$$

Log-likelihood of observation given forecast

$$L(\Omega|\Lambda) = \sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q (\omega_{ijk} \log \lambda_{ijk} - \lambda_{ijk} - \log \omega_{ijk}!)$$

Log-likelihood ratio of two forecasts

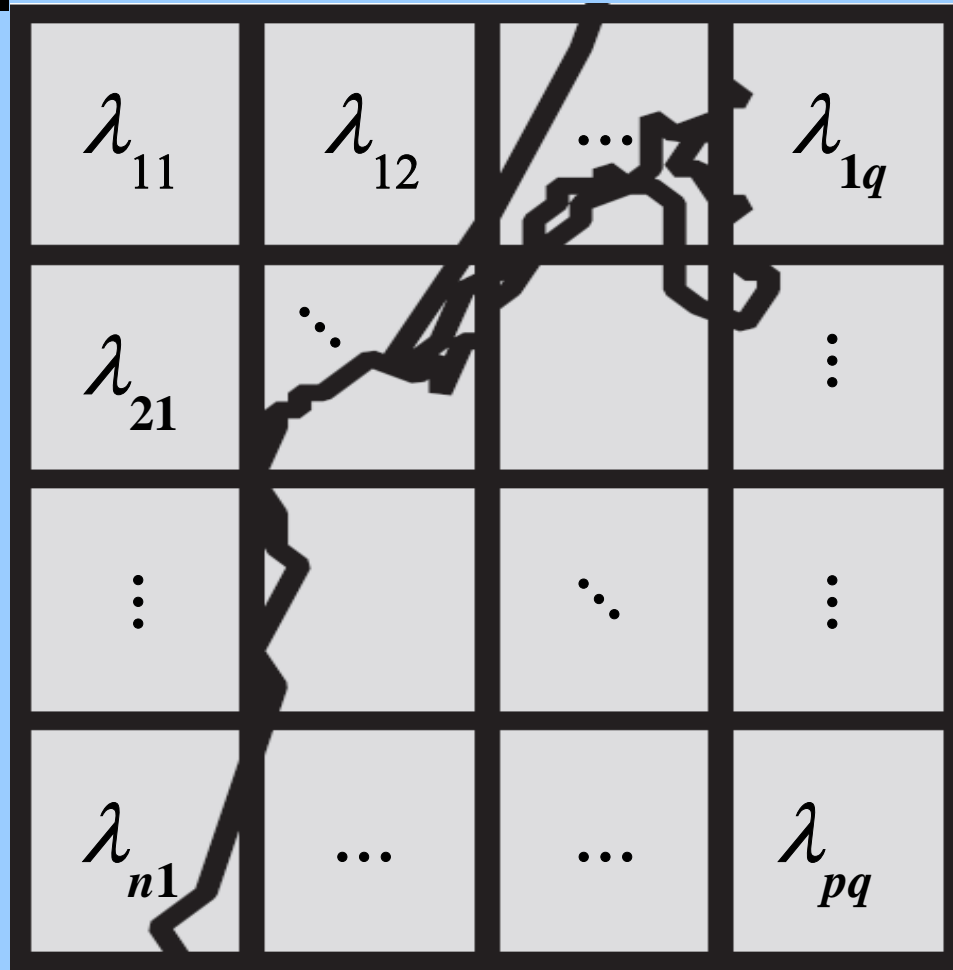
$$R_{xy} = L(\Omega|\Lambda_x) - L(\Omega|\Lambda_y)$$

RELM evaluation metrics

- $\underline{\delta}$ Compare number forecast with number observed—did the forecast predict an unreasonably high or unreasonably low seismicity rate?
- $\underline{\gamma}$ Compare forecast distribution with observed distribution—did the forecast obtain an unreasonably low log-likelihood?
- $\underline{\alpha}$ Compare log-likelihood ratio of two forecasts—when taken as null hypothesis, can a forecast be “rejected” by another? And *vice versa*?
- Further details (e.g., catalog uncertainty) in Schorlemmer *et al.* 2007 SRL

RELM spatial forecast

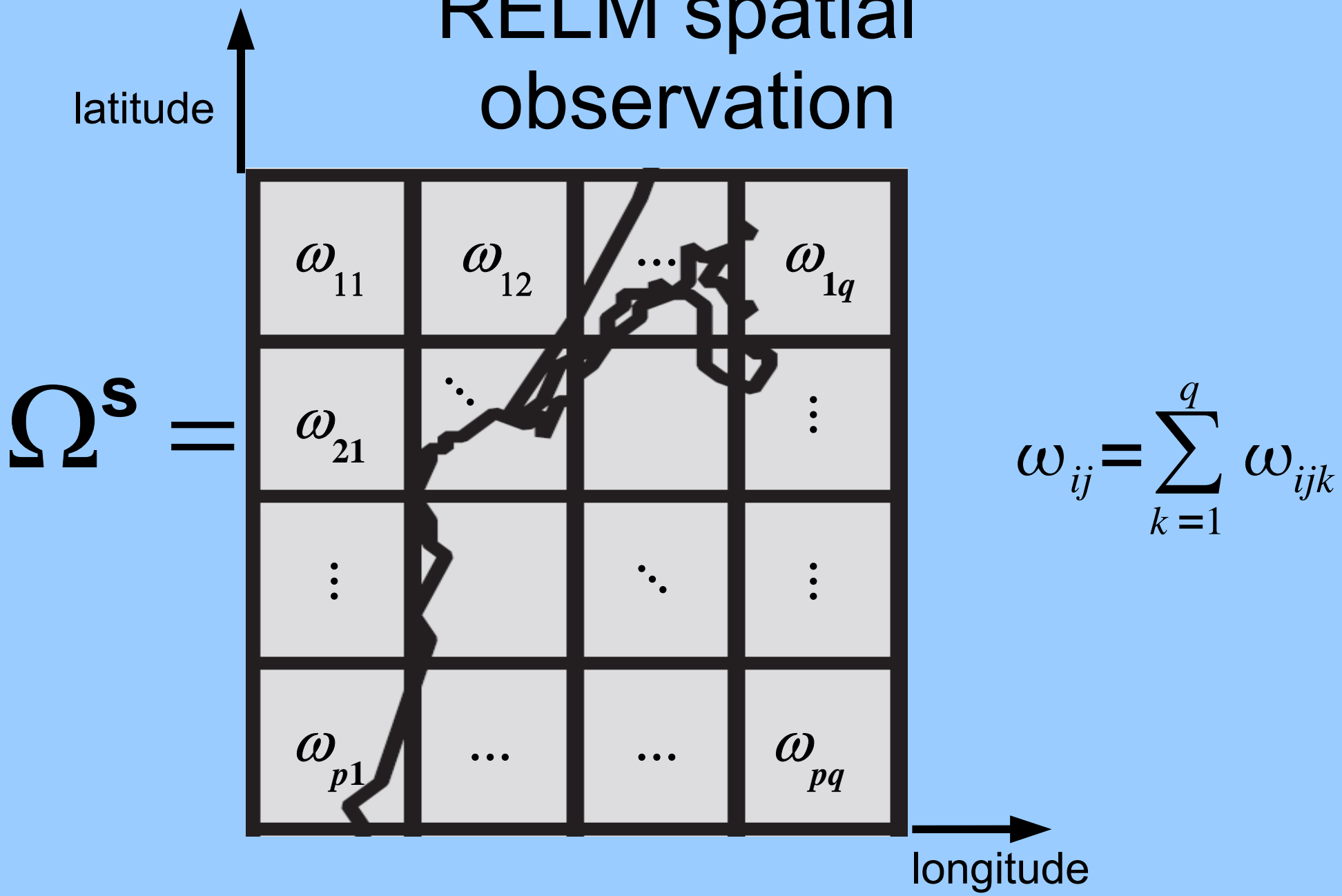
latitude



$$\Lambda^{\mathbf{s}} =$$

$$\lambda_{ij} = \sum_{k=1}^q \lambda_{ijk}$$

RELM spatial observation



Post-RELM likelihood statistics

Log-likelihood of observation given *spatial* forecast

$$S(\Omega^s | \Lambda^s) = \sum_{i=1}^p \sum_{j=1}^q (\omega_{ij}^s \log \lambda_{ij}^s - \lambda_{ij}^s - \log \omega_{ij}^s !)$$

Log-likelihood of observation given *magnitude* forecast

$$M(\Omega^m | \Lambda^m) = \sum_{i=1}^n (\omega_i^m \log \lambda_i^m - \lambda_i^m - \log \omega_i^m !)$$

Post-RELM evaluation metrics

- ζ Compare forecast spatial distribution with observed spatial distribution—did the forecast obtain an unreasonably low log-likelihood?
- κ Compare forecast magnitude distribution with observed magnitude distribution—did the forecast obtain an unreasonably low log-likelihood?

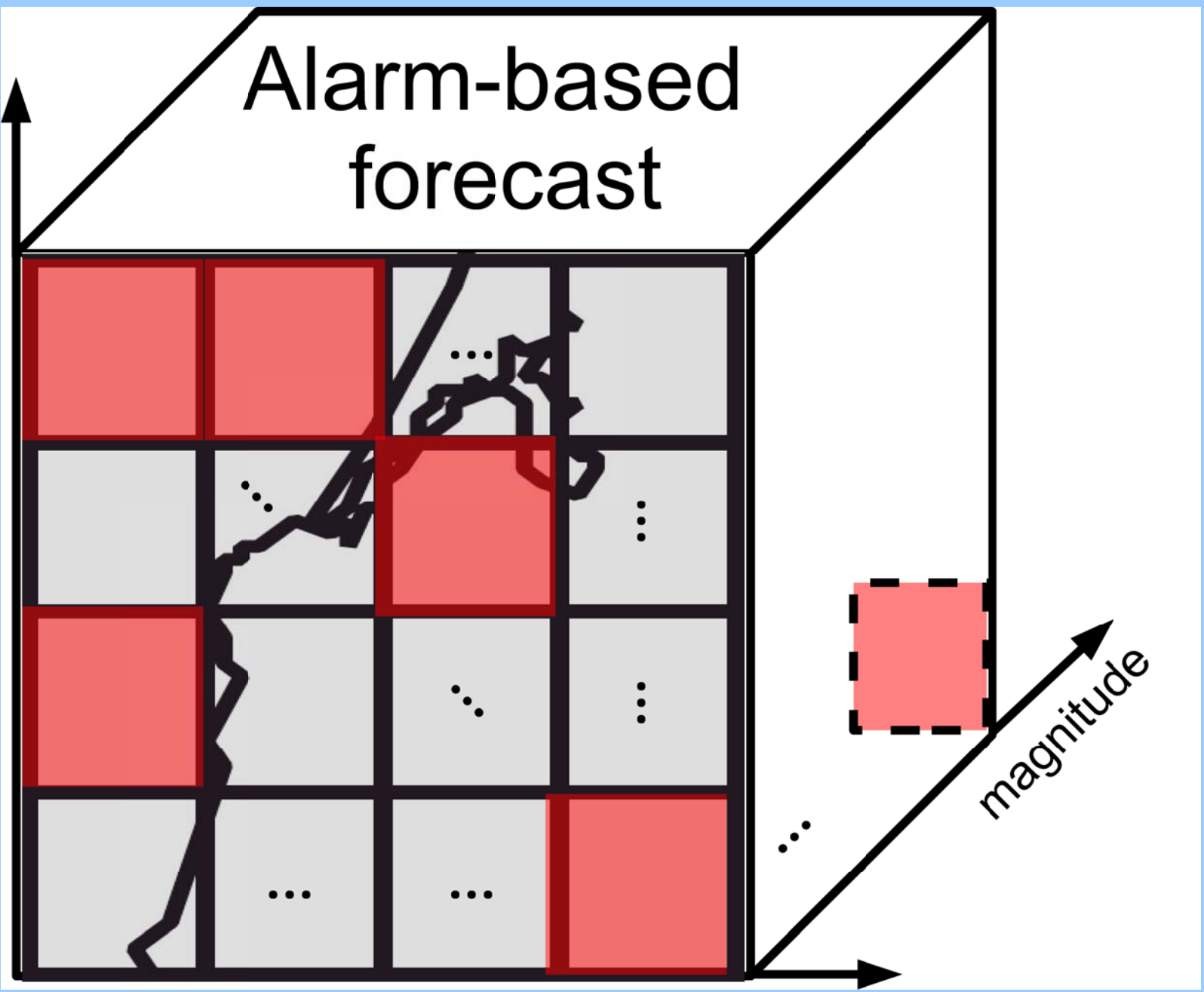
Likelihood tests summary

- Tests require gridded rate forecasts
- Each forecast is characterized by
 - Single $\delta, \gamma, \zeta, \kappa$ value
 - Vector of α values (when comparing N forecasts, $N-1$ elements)
- Results presented as plots or tables
 - Temporal variation of $\delta, \gamma, \alpha, \zeta, \kappa$ values
- Implemented in Python/MATLAB

Alarm-based forecast

latitude

$$\Lambda =$$



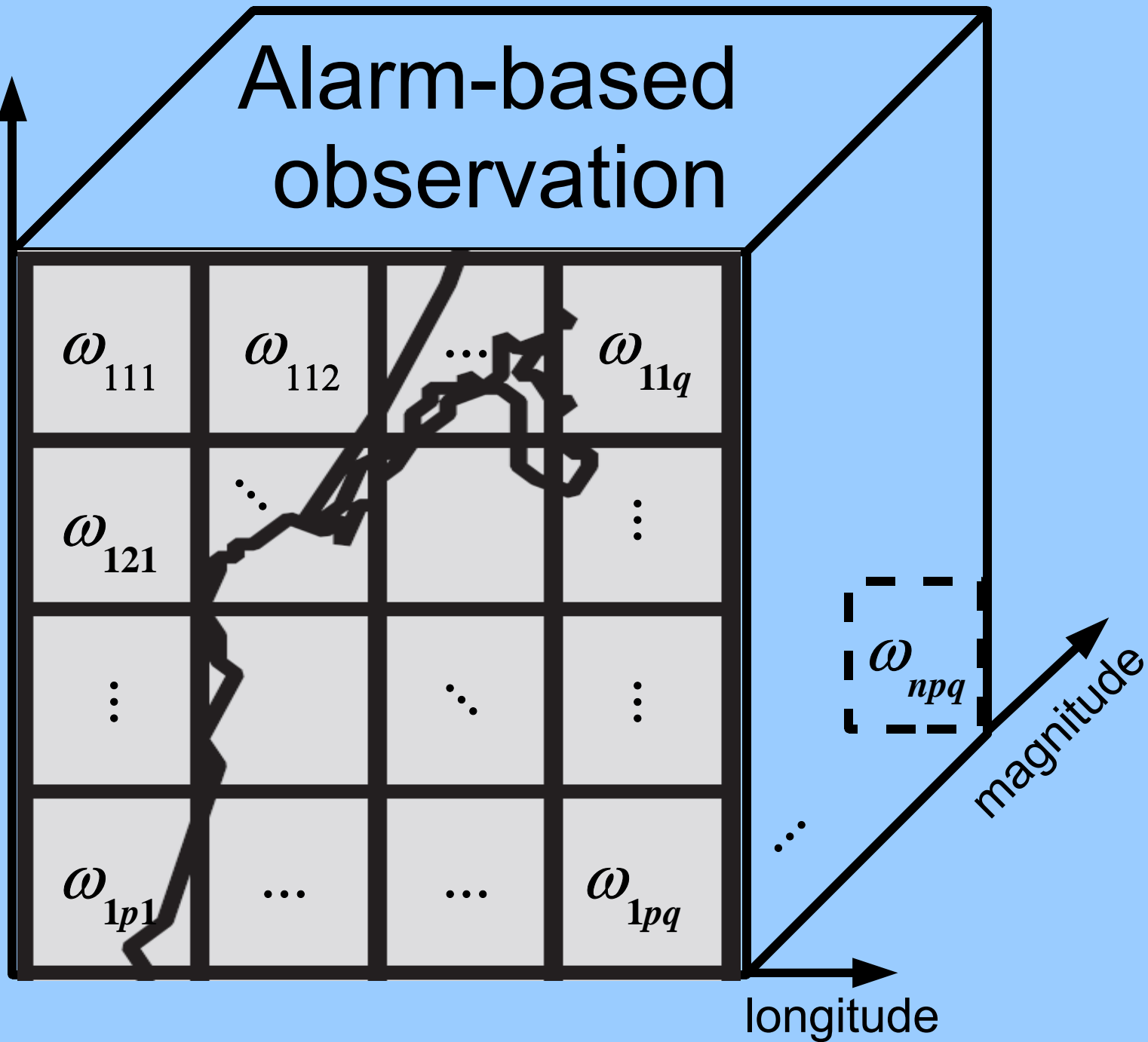
longitude

magnitude

Alarm-based observation

latitude

$\Omega =$



longitude

magnitude

Alarm-based statistics

Hit rate

$$H_{\Lambda} = \frac{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q (\lambda_{ijk} \cdot \omega_{ijk})}{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q \omega_{ijk}}$$

False alarm rate

$$F_{\Lambda} = \frac{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q (\lambda_{ijk} \cdot \mathbf{1}_{\omega_{ijk}=0})}{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q \lambda_{ijk}}$$

Fraction of space occupied by alarm*

$$\tau_{\Lambda} = \frac{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q \lambda_{ijk}}{n \cdot p \cdot q}$$

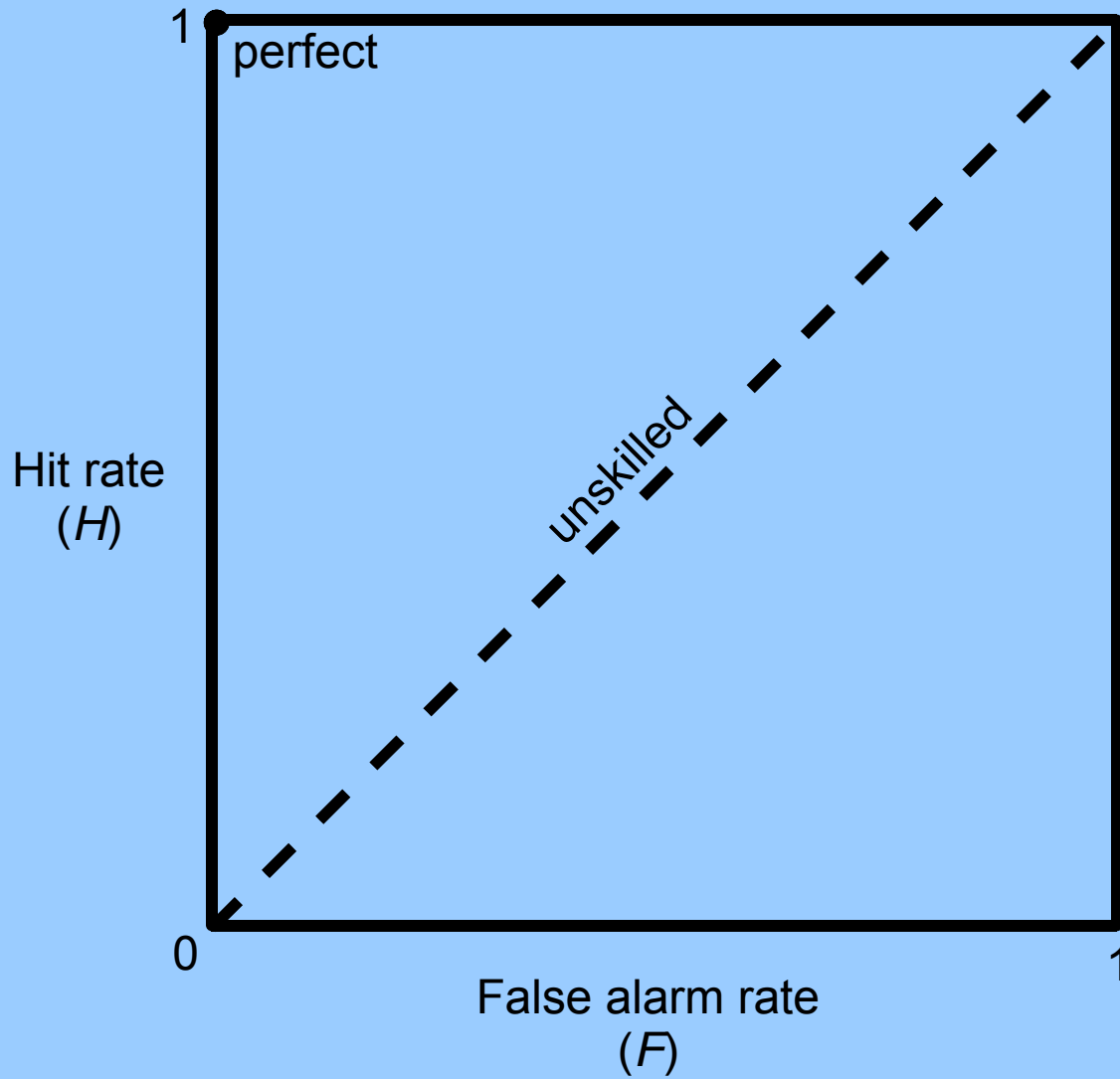
Miss rate

$$v_{\Lambda} = 1 - H_{\Lambda}$$

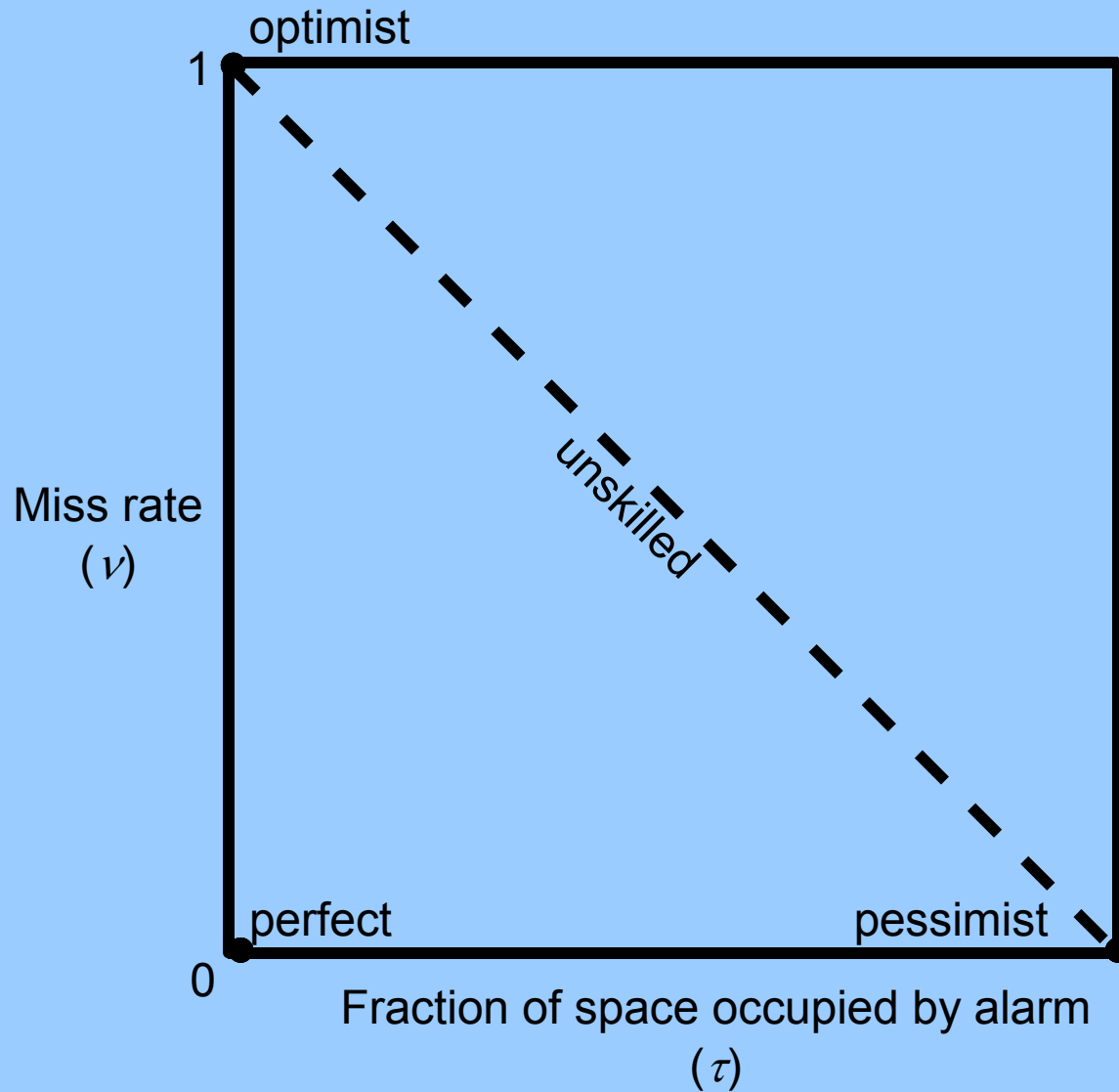
Alarm-based evaluation tools/metrics

- Receiver Operating Characteristic (ROC) – hit rate and false alarm rate
- Molchan error diagram – fraction of space occupied by alarm and miss rate
- Area skill score – derived from Molchan error diagram

ROC diagram



Molchan error diagram



Difference in unskilled reference

- For spatial forecast
 - ROC reference forecast is uniform
 - Molchan reference forecast is “user-defined,” should be best estimate of spatial distribution of seismicity

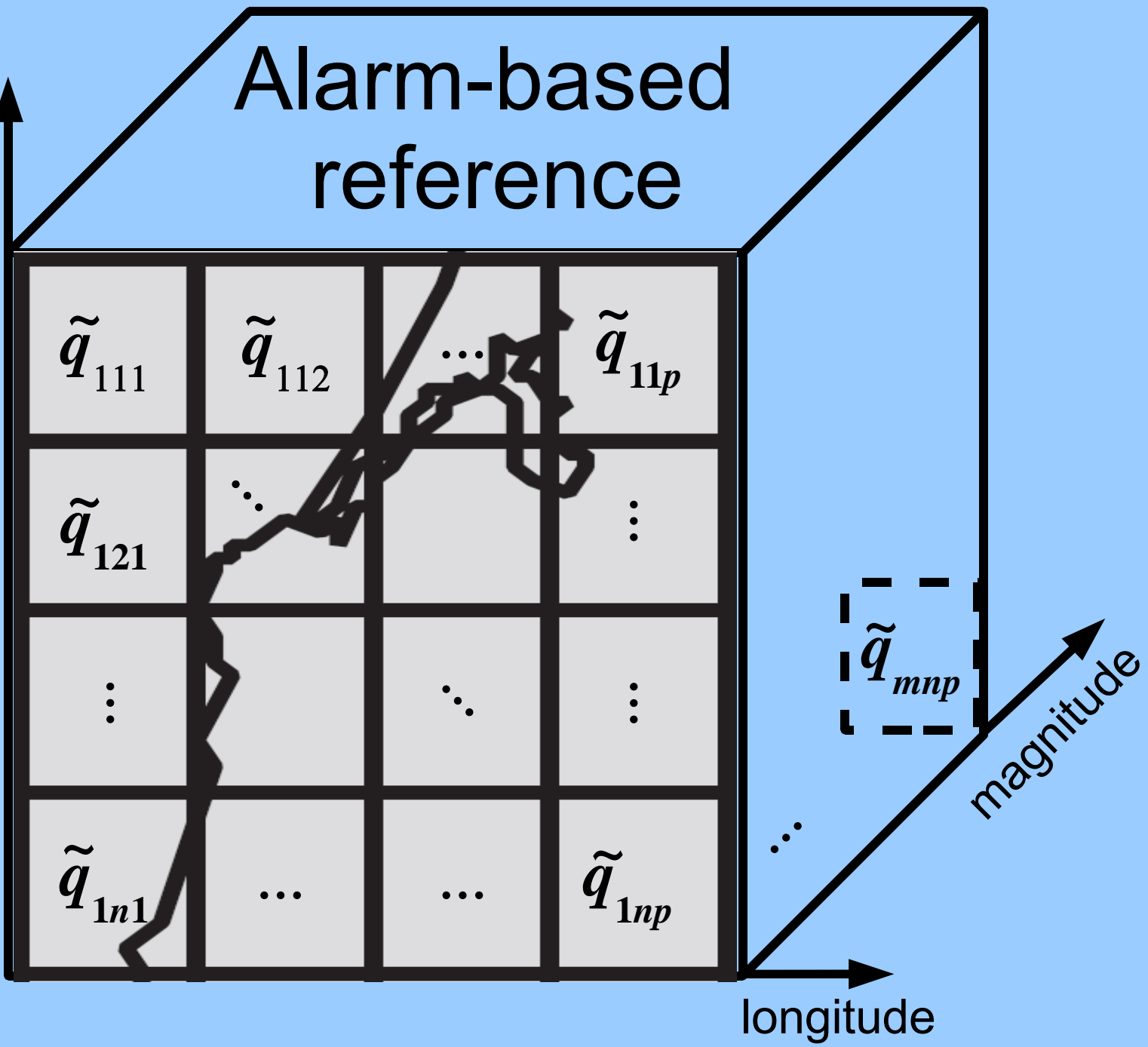
Fraction of space occupied by alarm

$$\tau_{\Lambda} = \frac{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q (\lambda_{ijk} \cdot \tilde{q}_{ijk})}{\sum_{i=1}^n \sum_{j=1}^p \sum_{k=1}^q \tilde{q}_{ijk}}$$

Alarm-based reference

latitude

$$\tilde{Q} =$$



longitude

magnitude

Post-RELM tests summary

- Tests require alarm function
 - Forecast orders regions of space/time/magnitude
 - Do not require rates
 - In principle, do not require gridding
- Tests allow/require choice of reference forecast
- Each forecast is characterized by
 - ROC: vector of (H, F) values (N elements, N is number of observed eqks)
 - Molchan: vector of (τ, ν) values (N elements)
 - Single area skill score value ($a_{\Delta}(\tau=1)$)
- Results presented as plots

Requirements for new test

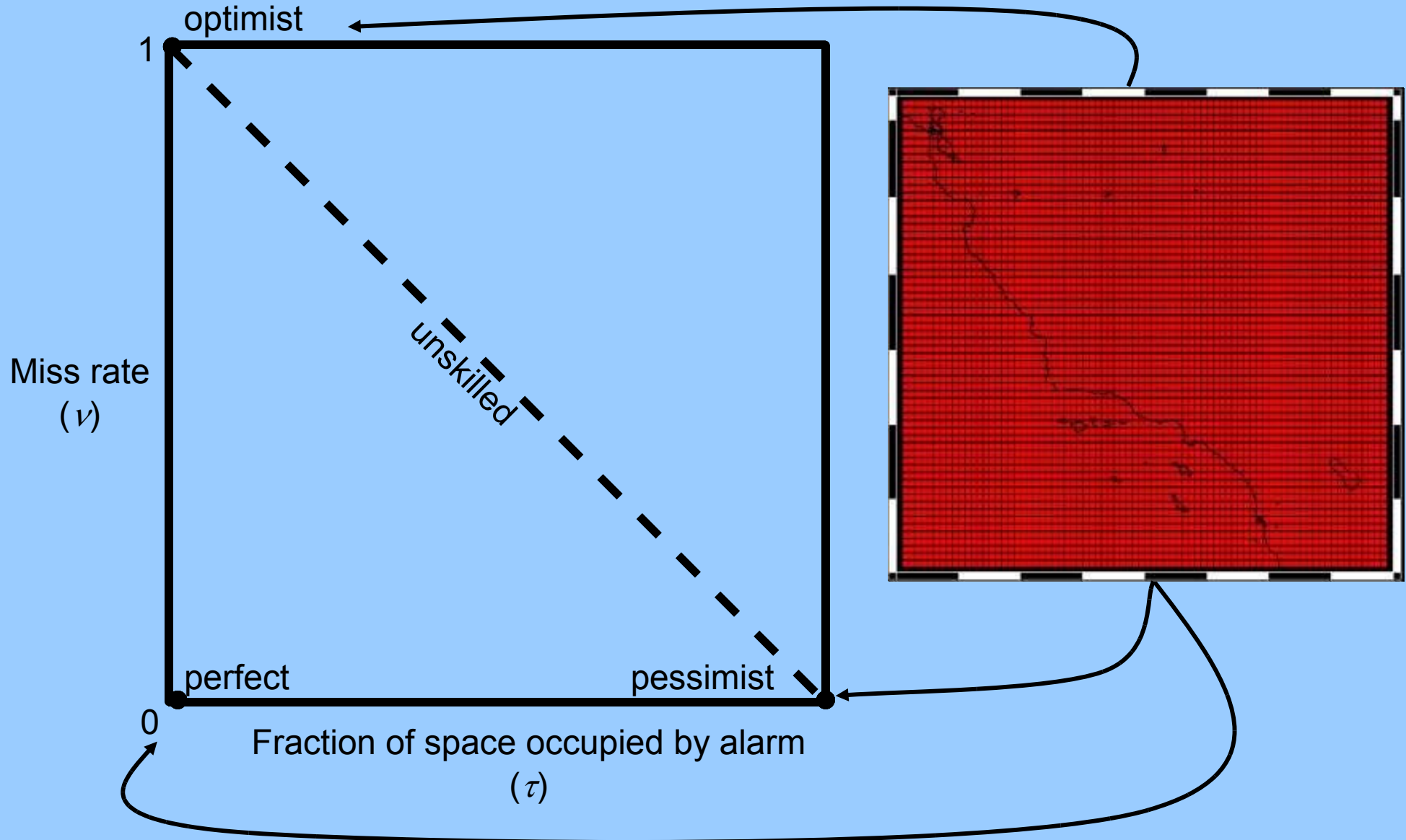
- Scientific justification
- Existence or introduction of suitable forecasts
- Technical
 - Codes should accept ForecastML and ZMAP formats as input, output ResultsML format
 - Codes should be documented and software dependencies stated explicitly

Process for new test implementation

- Work with CSEP Testing Center development team
 - Provide testing codes and support documentation
 - Provide reference data for unit test
 - For a given forecast, what is the expected result?
 - Use of random numbers constitutes a special case
- Aim for updates to operational system on quarterly basis
- Further details on CSEP computational infrastructure in Zechar *et al.* (ms. in review)

Thank you.

Molchan error diagram



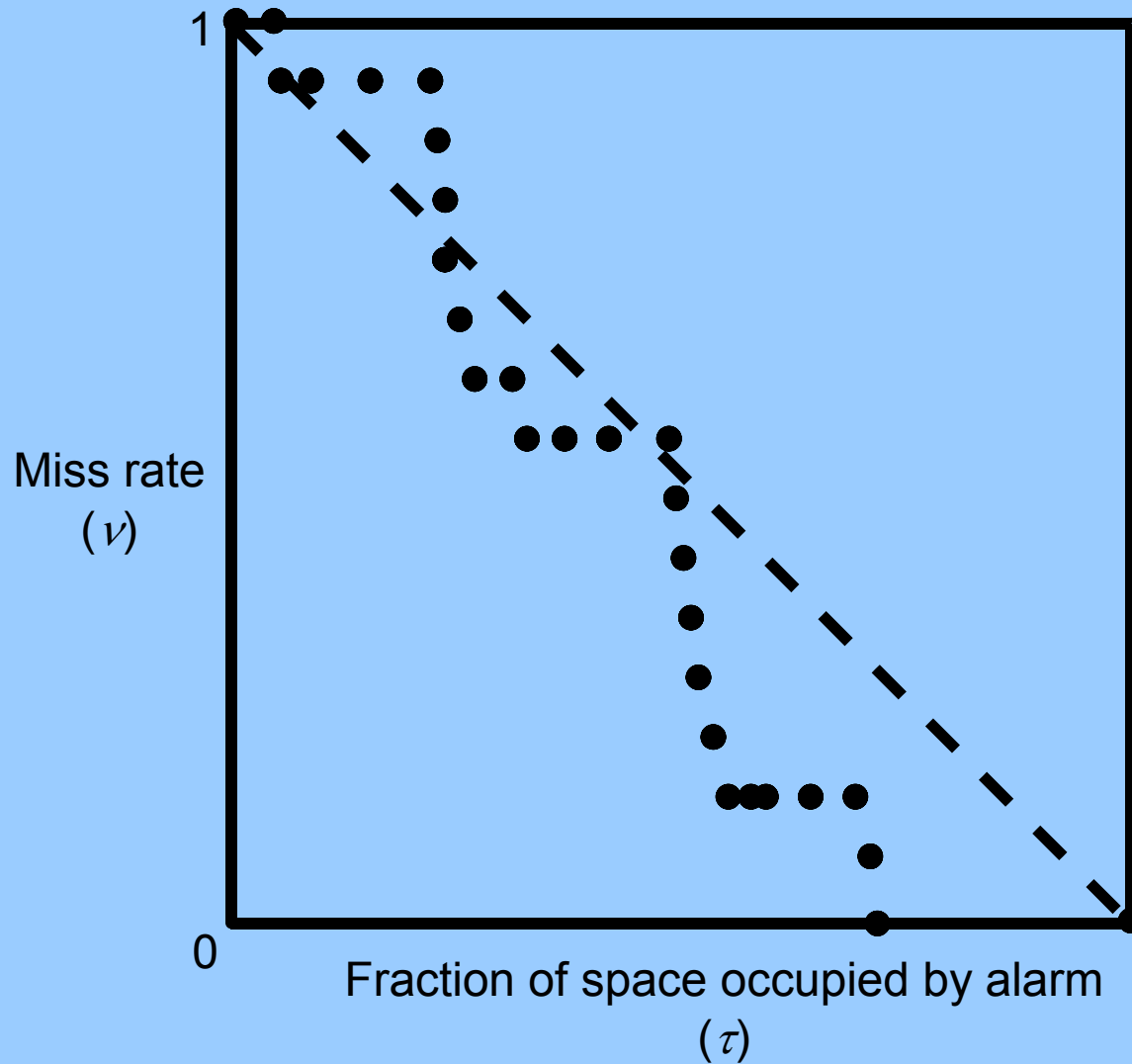
Generalize alarm set to alarm function



latitude

longitude

Molchan diagram



Molchan trajectory: collection of (τ, ν) points generated from alarm function

Area skill score

- Area above Molchan trajectory, normalized by τ

$$a_{\Lambda}(\tau) = \frac{1}{\tau} \int_0^{\tau} (1 - v_{\Lambda}(t)) dt$$

- Unskilled forecasts yield area skill score $\sim 1/2$; forecast skill is characterized by deviation.
- Further details in Zechar & Jordan 2008 GJI, Zechar & Jordan 2009 PAGEOPH