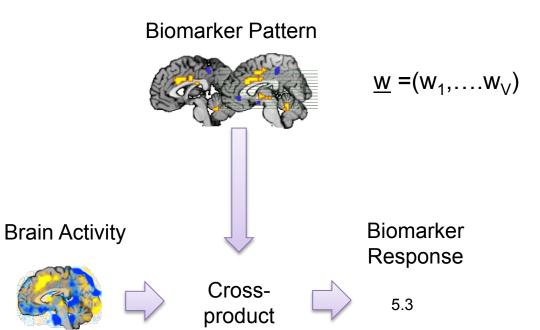
# Module 30: MVPA Example

#### Pain Biomarkers

- Pain is an important health problem affecting the quality of life in a large segment of the population.
  - It is usually measured using self-reports.
- The goal is to define robust and meaningful fMRIbased biomarkers for physical pain that can augment reported pain as outcome measures.

#### Illustration



$$\underline{\mathbf{x}} = (\mathbf{x}_1, \dots, \mathbf{x}_V)$$
  $\mathbf{y} = \underline{\mathbf{w}}^T \underline{\mathbf{x}}$ 

### Statistical Learning

- We develop the biomarker using data from a study of thermal pain (n=20).
  - Statistical learning techniques are used to identify spatial patterns of fMRI activity that accurately and specifically predict physical pain at the individual level.
- The development of this method entails:
  - defining appropriate features;
  - choosing a classifier;
  - training and testing the classifier; and
  - validating results on new data sets.

### Example

- Participants (n=20) received a series of thermal stimulations for 12 trials at each of four intensities: innocuous warmth and three levels of increasingly painful heat.
  - Each trial consisted of separate periods of anticipation, thermal pain and pain recall.
  - The outcome measure was a trail-specific pain rating reported on a continuous visual analogue scale.

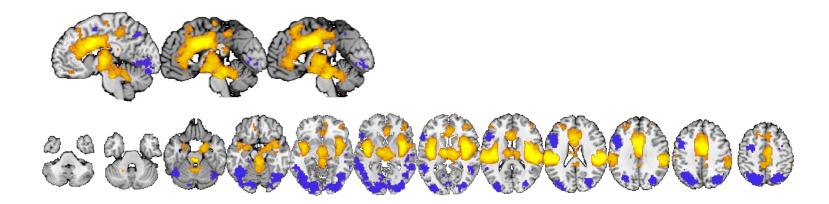
Cue	Pain	ISI 1	Rating	ISI2
8 s	10 s	14 s	4 s	10 s

### Performing MVPA

- The process of performing MVPA follows a series of steps:
  - Defining features and classes
  - Feature selection
  - Choosing a classifier
  - Training and testing the classifer
  - Examining results

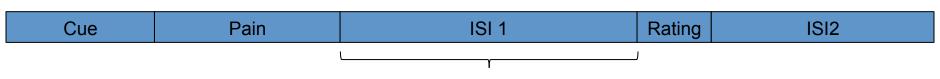
#### Feature Selection

- Voxels within pain-related brain regions were selected based on prior literature.
  - Performed a meta-analysis of 224 previous studies to select ~16,000 voxels associated with pain.



# Data Averaging I

- Time courses immediately following the end of heat application were included in the analysis;
  - Averaged brain activity at each temperature for each participant over a 14s post-stimulus window.
  - At each temperature we obtained 12 maps and 12 responses for each subject.



### Data Averaging II

- Data were averaged over repeated trials at each intensity level.
  - Averaged the 12 repetitions at each temperature for each participant to yield 4 unique maps per subject.
  - Averaged pain response for each participant in each condition.

Maps Responses

Subject i

# Machine Learning

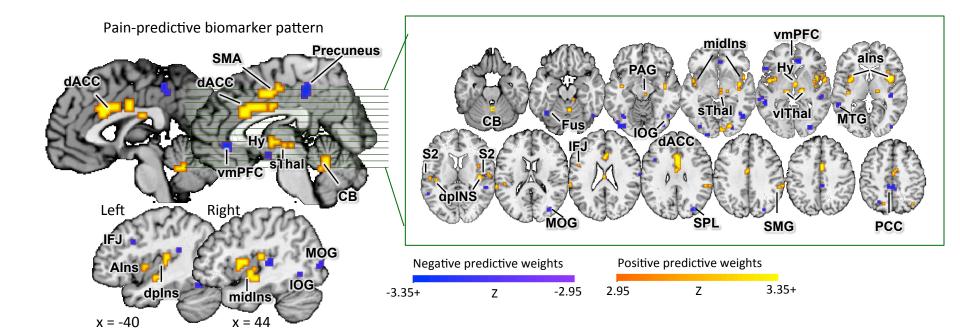
#### Classifier:

- LASSO-PCR
- Other classifier techniques gave comparable results.

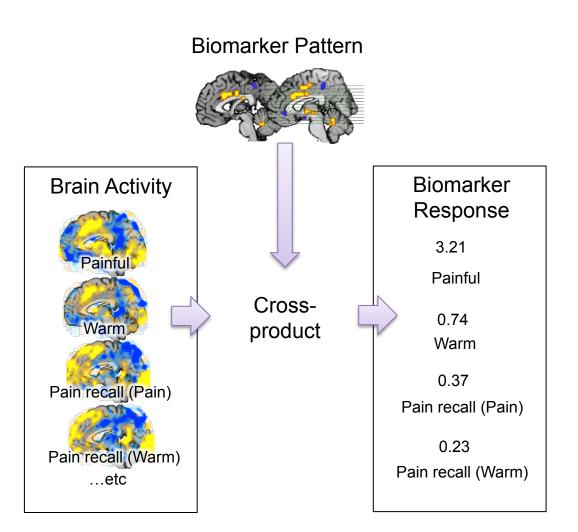
#### Training and testing:

Used leave-one-subject out cross-validation.

# Weight Maps

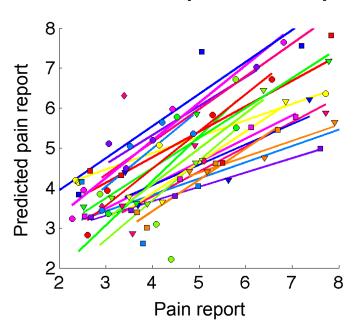


# **Analysis**

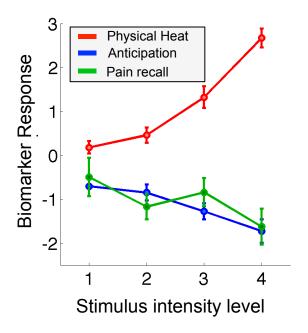


### Results

#### **Cross-validated prediction of pain**



#### Pain vs. other affective events

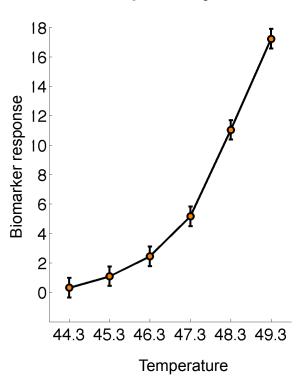


### Study II

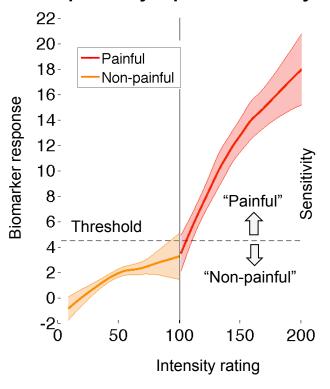
- Participants (n=33) received a series of 75 thermal stimulations across six different temperatures.
- After each stimulus, participants rated whether it was painful or not.
  - If non-painful, intensity was rated on a 100-point scale ranging from "no sensation at all" to "very warm but not yet painful."
  - If painful, intensity was rated on a 100-point scale ranging from "no pain" to "worst imaginable pain."

# Results - Study II

#### Biomarker response by condition



#### Response by reported intensity



#### An fMRI-Based Neurologic Signature of Physical Pain

Tor D. Wager, Ph.D., Lauren Y. Atlas, Ph.D., Martin A. Lindquist, Ph.D., Mathieu Roy, Ph.D., Choong-Wan Woo, M.A., and Ethan Kross, Ph.D.

N Engl J Med 2013; 368:1388-1397 | April 11, 2013 | DOI: 10.1056/NEJMoa1204471

### **End of Module**

