

# **Rhythms**

## **Introduction**

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# Brain rhythms

Introduction:

What are they?

Where do they come from?

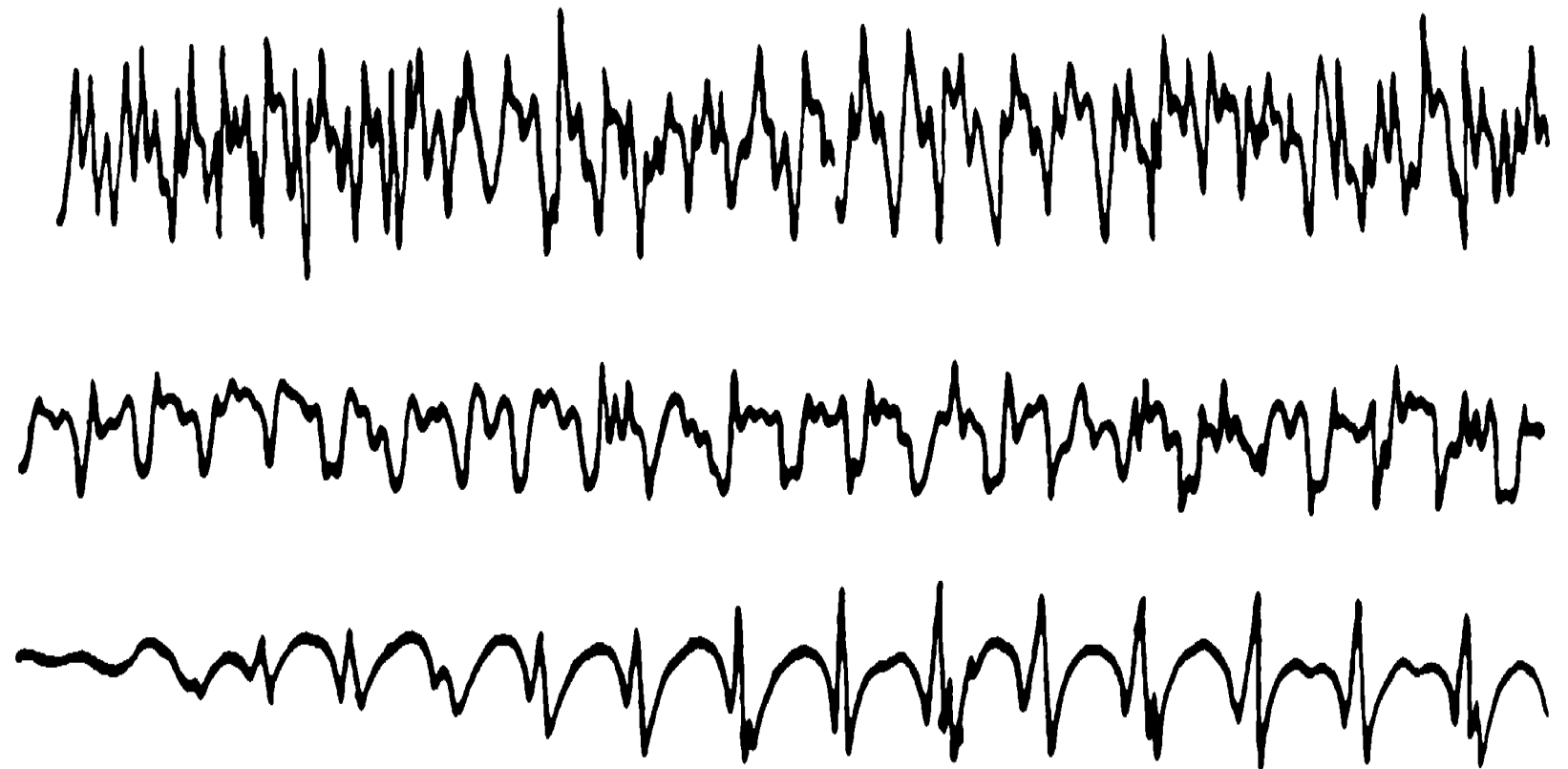
Method of analysis.

# Brain rhythms

Fact: The brain can generate rhythmic activity.

**Ex**: Scalp electroencephalogram (EEG)

Note: Rhythms also appear in LFP, MEG, fMRI, ...

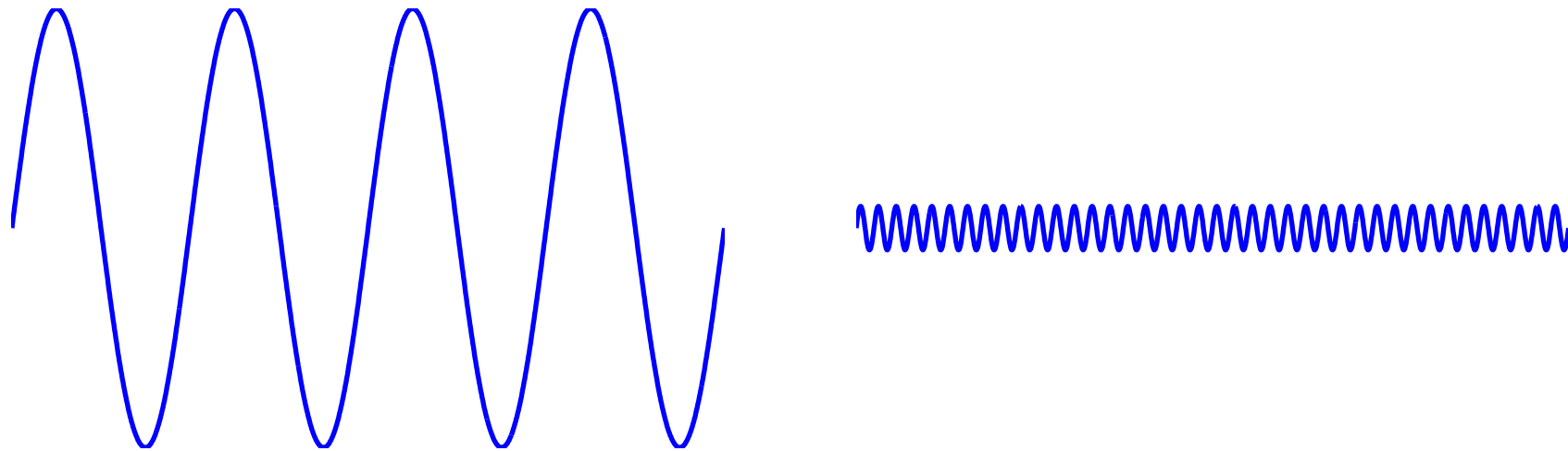


Observe: Different shapes. Different frequencies.

# Brain rhythms: basic facts

Some basic facts about EEG rhythms:

- Slower rhythms tend to be larger amplitude.

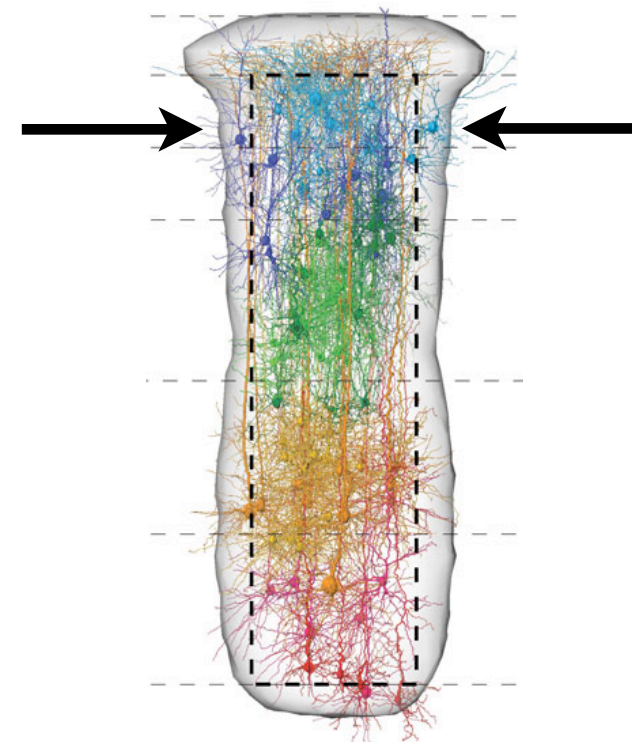


- The EEG is thought to be (mostly) generated by coordinated synaptic transmembrane currents of many neurons.

Very complicated.

Not completely understood.

[Buzsáki et al. Nat Rev Neurosci (2012)]

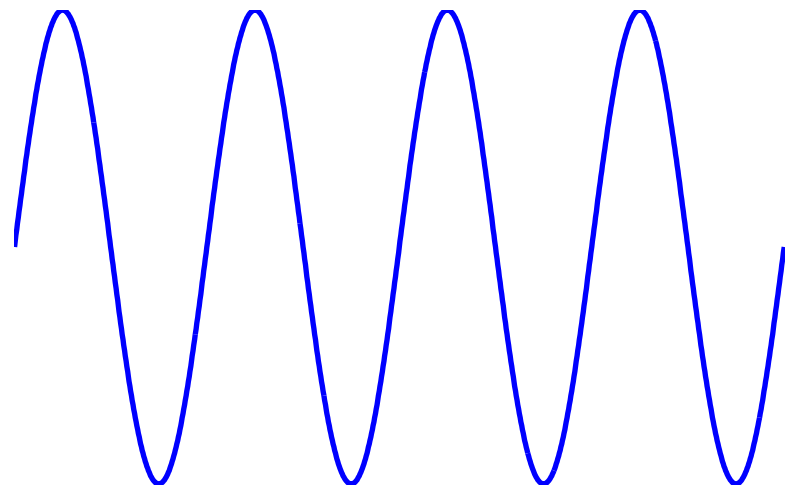


# Brain rhythms: basic facts

EEG measures cortical arousal

–Appears in the firing patterns of cortical neurons, measured in the EEG.

Low frequency, high amplitude: low cortical arousal



Neural activities are . . . synchronized.

Groups of cells act in concert =  
large EEG signal.

High frequency, low amplitude: high cortical arousal

 Neural activities are . . . desynchronized.

Groups of cells involved in separate activities = small EEG signal.

Note: A healthy brain is a desynchronized brain.

# Brain rhythms: characterization

**Q:** How do we characterize these rhythms?

To start, we can visualize and describe the rhythms.

Typical features:

**Amplitude:** Large or small ?

**Frequency:** Fast or slow ?

**Shape:** Sinusoidal, square, triangle, . . . ?

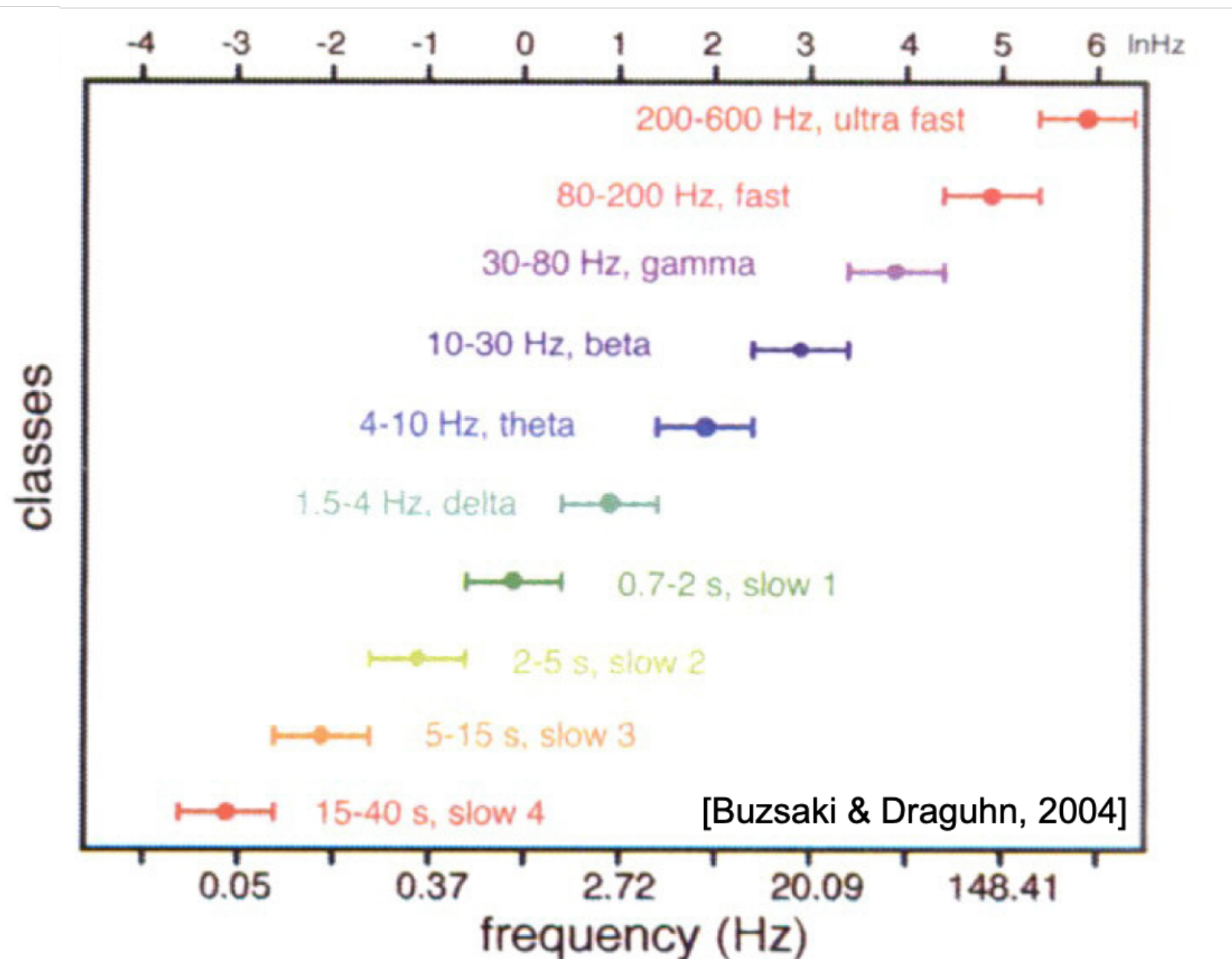
Our focus (usually) is frequency: How fast or slow is a rhythm?

**Q:** Why? Because different frequency rhythms are associated with different functions ...

# Brain rhythms and functions

Many frequency bands, each associated with different functions.

**Ex:**



[Buzsáki & Draguhn, Science, 2004]

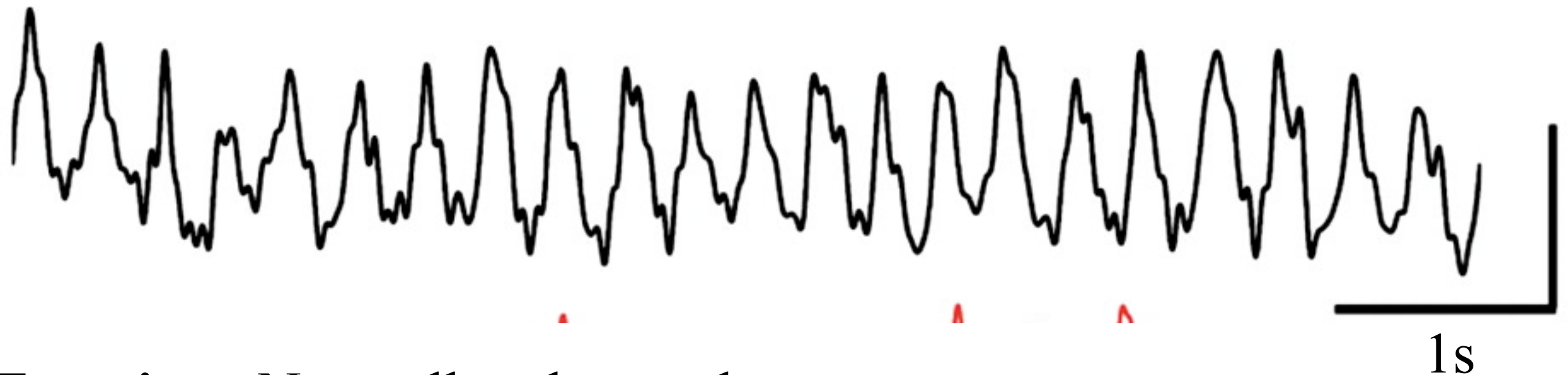
Let's look more carefully at some of these frequency bands . . .



# Brain rhythms: theta

**Theta:** 4-8 Hz

Note: Theta frequency range different; the borders of ranges are not exact.



**Function:** Not well understood.

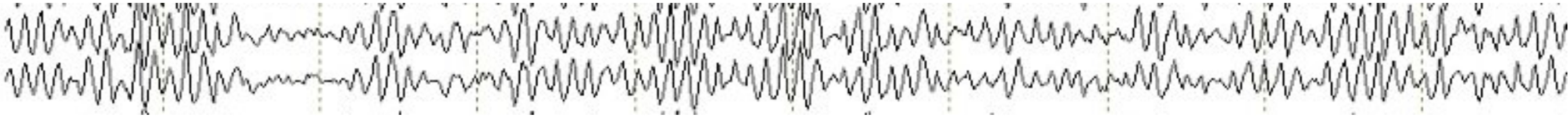
In rats: learning and memory  
location  
motor behavior  
sleep  
emotional arousal  
fear conditioning



# Brain rhythms: alpha

**Alpha:** 8-12 Hz

Note: This band not in Slide #7 !



- The first EEG wave studies [Berger 1931]
- In EEG, strongest above occipital lobes when eyes closed at rest.

**Function:** “idling rhythm” - alert but still brain state  
cortical operations in the absence of sensory inputs  
disengagement of task-irrelevant brain areas

However, alpha also associated with attention,  
sensory awareness.

# Brain rhythms: beta

**Beta:** 12-30 Hz

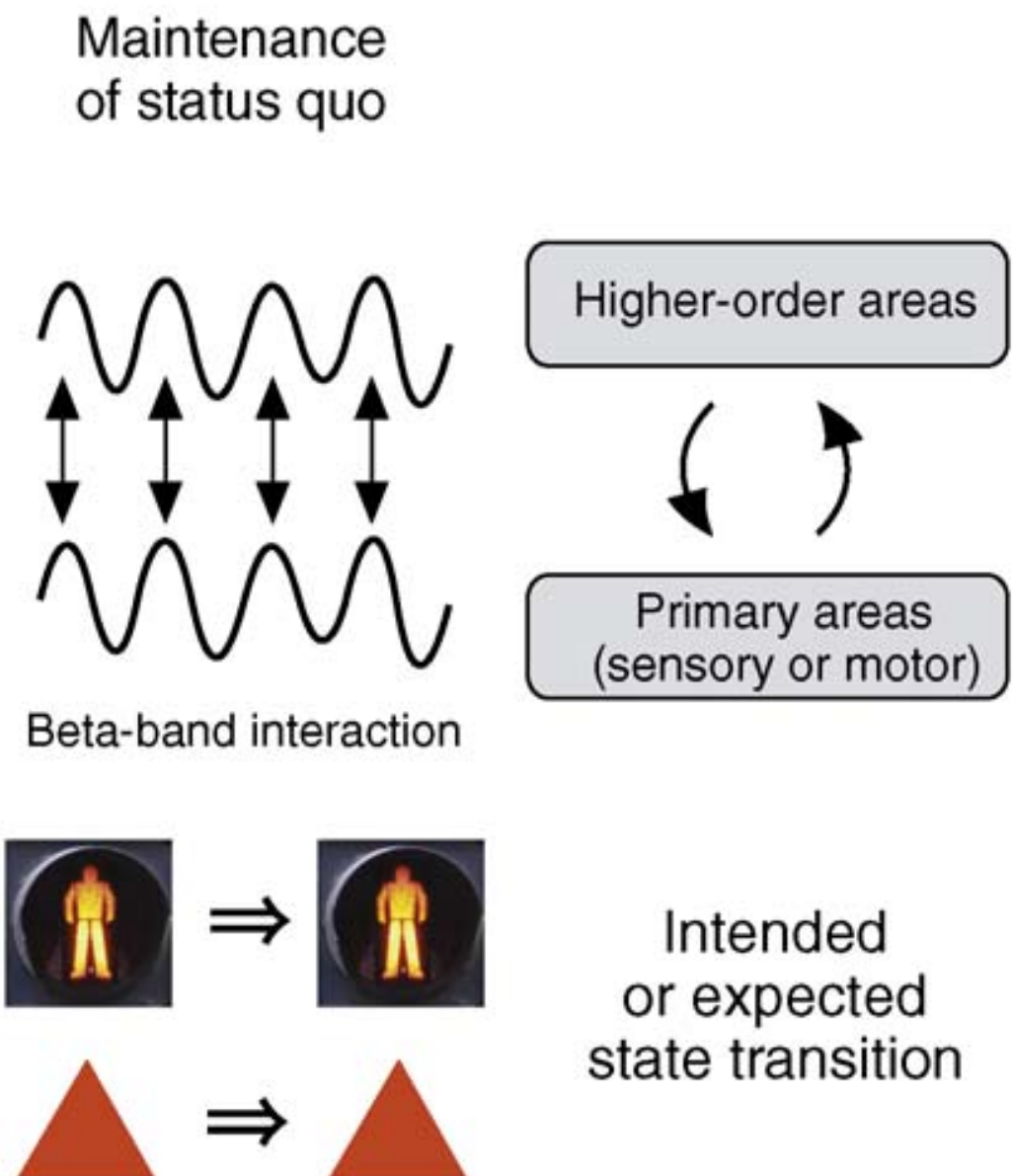


**Function:** Not well understood.

motor functions (e.g., steady-state contractions).

“Maintenance of the status quo”

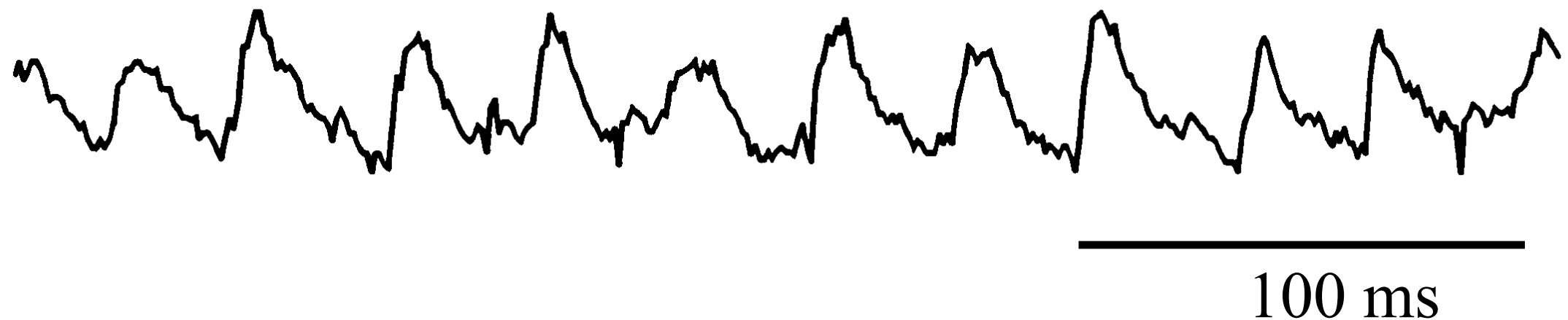
“Couple” distant brain regions.



[Engel, Fries 2010]

# Brain rhythms: gamma

**Gamma:** 30-50 Hz



**Function:** Associated with a broad range of processes:  
“binding”  
attention  
movement preparation  
memory formation  
conscious awareness

Note: Typically hard to see in EEG.      Q: Why?

# Brain rhythms: Other bands

There are many other frequency bands . . .

## Slower

- |  |                             |
|--|-----------------------------|
| – <b>Delta:</b> 1-4 Hz                     | Sleep, learning, motivation |
| – <b>Slow cortical potential:</b> $< 1$ Hz | Emergence of consciousness? |

## Faster

- |  |  |
|--|--|
| – <b>High gamma:</b> 50-120 Hz         | Coordination of neural activity                |
| – <b>Ripples, HFO, UFO:</b> $> 120$ Hz | Replay of memories,<br>onset of seizures . . . |

# Brain rhythms in disease

Rhythms are sometimes associated with pathologies.

**Ex:** Seizure



**Q:** What rhythms do we see?

HFO

Beta

Alpha

Theta

Delta

**Q:** How does the amplitude change?

Note: Used the band labels, but “function” very different.

# Brain rhythms are meaningless . . .

**H:** Brain rhythms are epiphenomena.

Large buildings: sway in the wind.

These oscillations are not performing a function, in fact they're unwanted.

**Q:** Is the same true in the brain?  
Oscillations are an echo of some underlying function?

**Q:** Why so many oscillations?



# Big questions

**Q:** Why do we observe rhythms in the brain?

- Functional role
- Epiphenomena

**Q:** What mechanisms support rhythms?

- Biological
- Dynamical

**Q:** Why do we observe different frequency bands, not a continuum?

**Q:** Why do rhythms interact?

- Mechanisms
- Functions
- Measures . . .

We need answers to these questions ...  
Data analysis and models ...



# Characterize brain rhythms

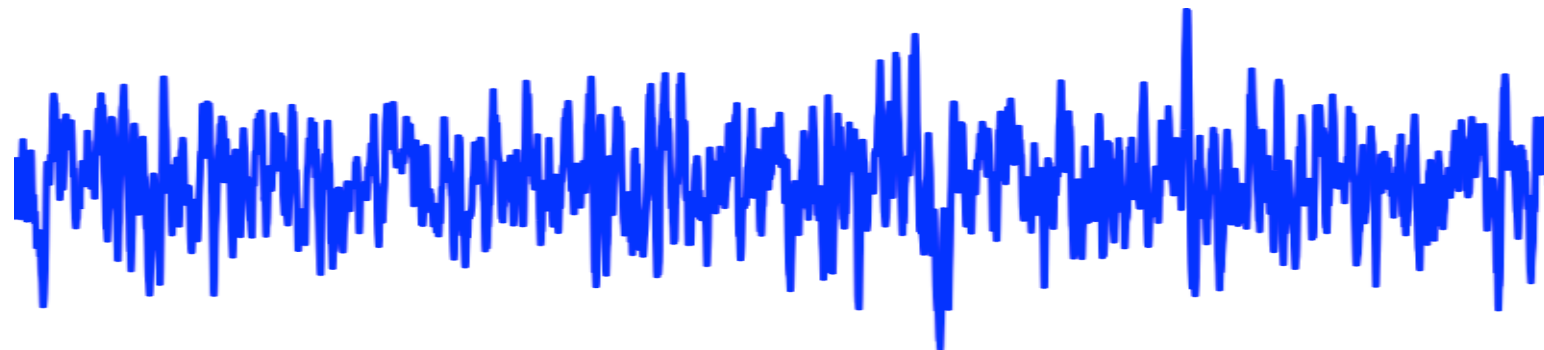
Many sophisticated tools to do so.

Today, consider two:

- Visual inspection
- Intuition for the power spectrum

Idea:

**Visual inspection:** Plot the data. What do you see?



**Power spectrum:** Break down the data into sinusoids . . .

**Remember: sinusoids ...**

$$V[t] = A \cos(2\pi f t)$$

Voltage as a  
function of time

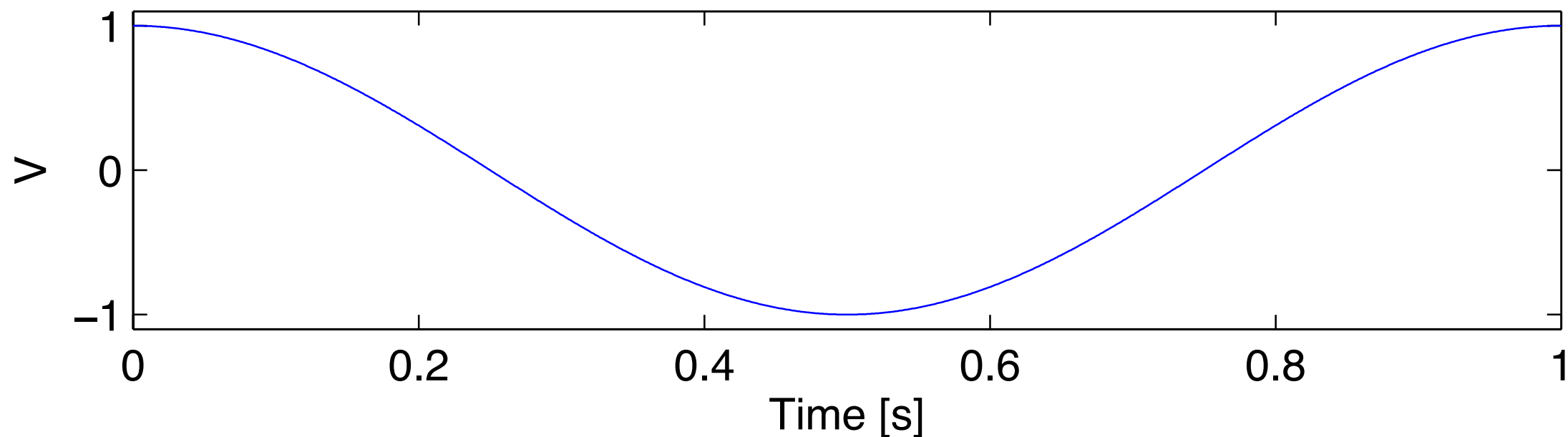
Amplitude

Frequency [Hz]

Time [s]

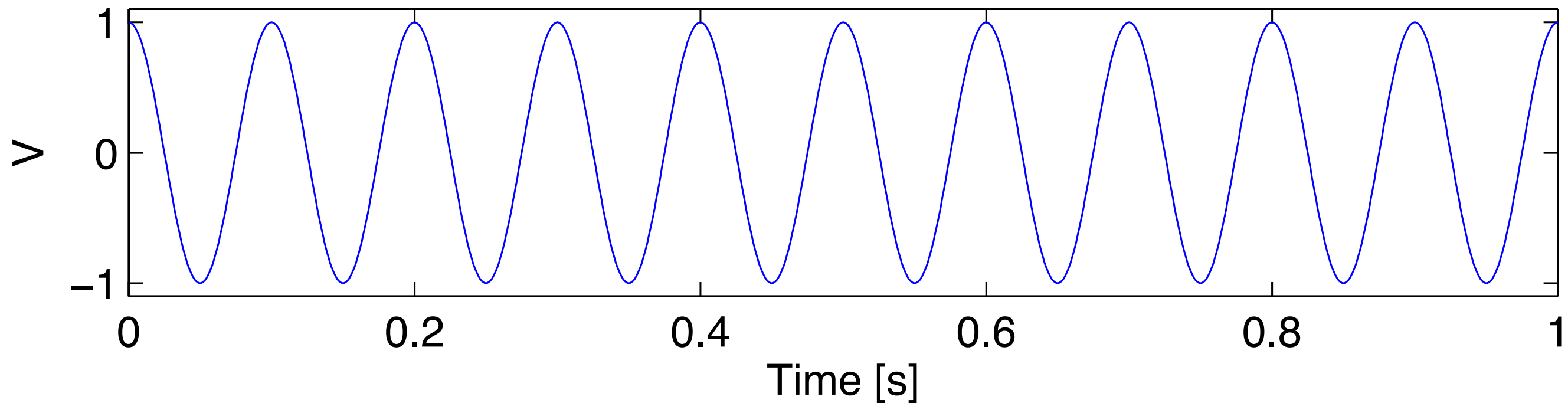
**Ex.** Consider:  $f = 1$  Hz

**Q:** What does it look like?



## Remember: sinusoids . . .

**Q:** What is the frequency of this sinusoid?



**A:** 10 cycles in 1 second, so 1 Hz.

Note: Visual inspection is often useful.

## Remember: sinusoids ...

In addition to cosine, there's also sine:

$$V[t] = B \sin(2\pi f t)$$

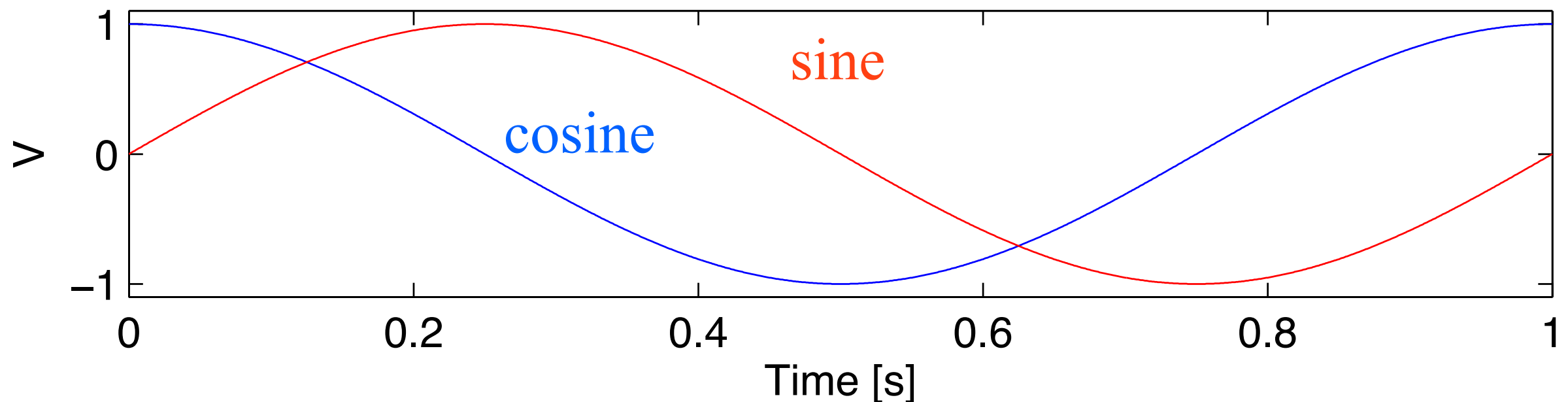
Voltage as a function of time

Amplitude

Frequency [Hz]

Time [s]

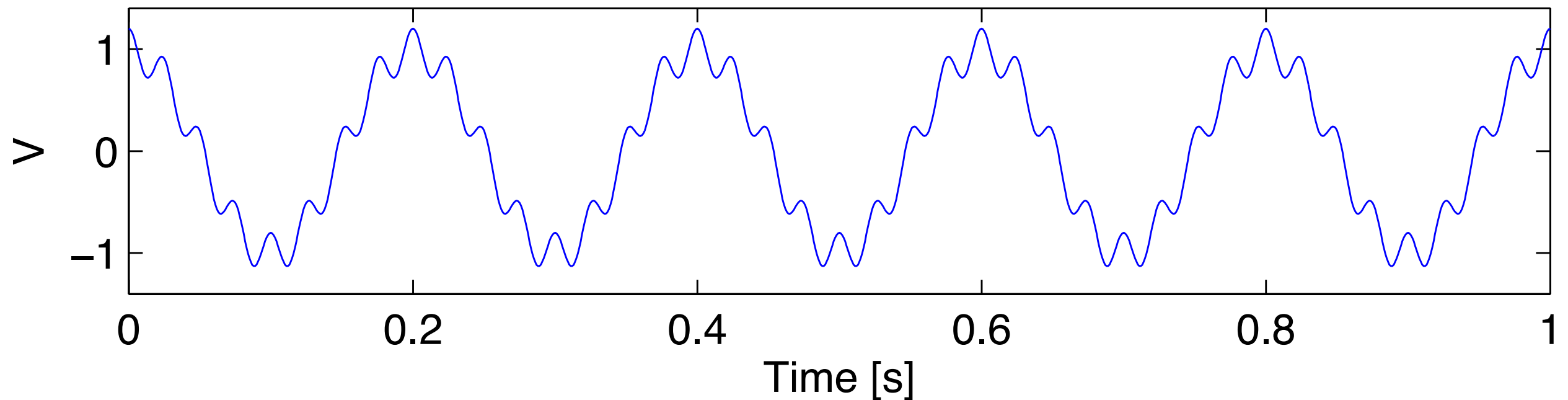
**Ex.** Consider:  $f = 1$  Hz



**Q:** What's the difference?

# Example: rhythmic signal

Consider the signal below:



**Q:** What are the rhythms?

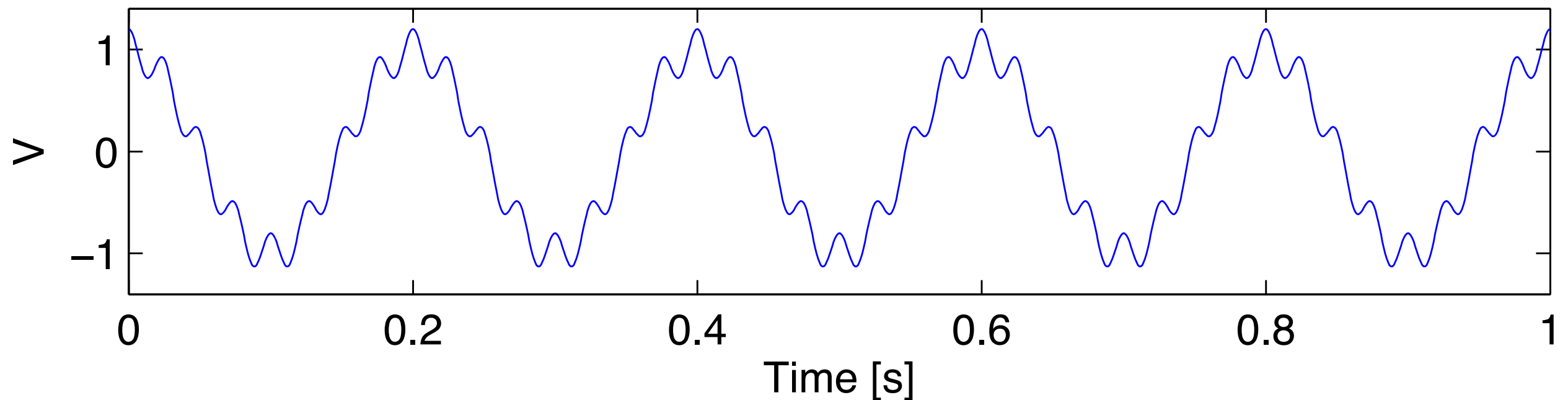
**A:** Apply visual inspection . . .

Slow	and fast
5 Hz	40 Hz

**Q:** What has larger amplitude?

## Example: rhythmic signal

So, we can represent this signal . . .



. . . as the sum of two sinusoids:

A slow, large amplitude sinusoid + a fast, small amplitude sinusoid

$$V[t] = A_1 \cos(2\pi f_1 t) + A_2 \cos(2\pi f_2 t)$$

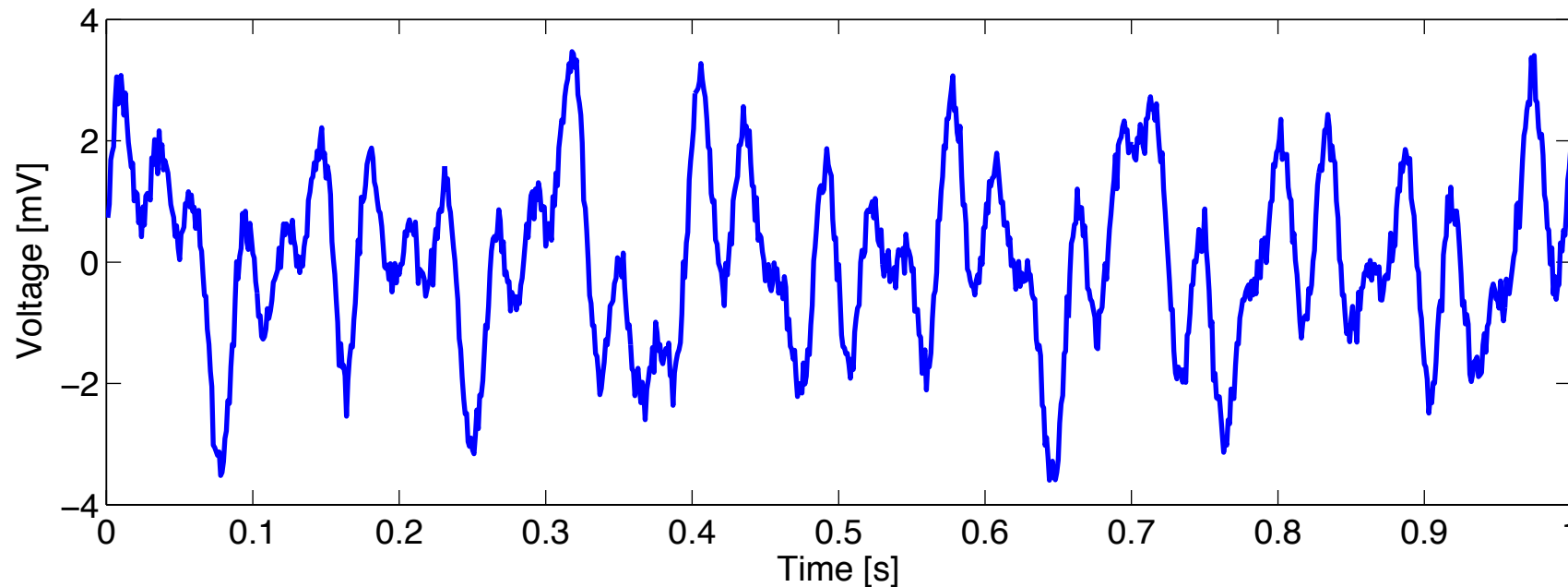
1.0                      5 Hz                      0.1                      40 Hz

We get a simpler representation of the signal. That's the idea of the power spectrum.

# Idea: Power spectrum

Consider:

$V =$



- Decompose signal into oscillations at different frequencies.

$$\begin{aligned} V = & \text{[sinusoid]} \quad \begin{array}{l} \updownarrow A_1 \\ f_1 \end{array} \\ & + \text{[sinusoid]} \quad \begin{array}{l} \updownarrow A_2 \\ f_2 \end{array} \\ & + \text{[sinusoid]} \quad \begin{array}{l} \updownarrow A_3 \\ f_3 \end{array} \\ & + \text{[sinusoid]} \quad \begin{array}{l} \updownarrow A_4 \\ f_4 \end{array} \\ & + \dots \end{aligned}$$

Represent  $V$  as a sum of sinusoids (e.g., part 7 Hz, part 10 Hz, . . . )



# Idea: Power spectrum

So, in equations:

$$V[t] = \underset{\substack{\uparrow \\ \text{amplitude}}}{A_1} \cos(\underset{\substack{\uparrow \\ \text{frequency}}}{2\pi f_1 t}) + B_1 \sin(2\pi f_1 t) + \underset{\substack{\uparrow \\ \text{amplitude}}}{A_2} \cos(\underset{\substack{\uparrow \\ \text{frequency}}}{2\pi f_2 t}) + B_2 \sin(2\pi f_2 t) + \dots$$

Or, more generally:

$$V[t] = \sum_j \underset{\substack{\uparrow \\ \text{amplitude}}}{A_j} \cos(\underset{\substack{\uparrow \\ \text{oscillation at} \\ \text{frequency } f_j}}{2\pi f_j t}) + B_j \sin(2\pi f_j t)$$

sum over all possible frequencies

Note:  $A_j$  and  $B_j$  can be zero.

(some rhythms make no contribution to  $V[t]$ )

Note:  $A_j$  and  $B_j$  are large when  $f_j$  is a good match to the data.

## Idea: Power spectrum

Note: Think of sine & cosine as accounting for **phase**.

$$C_j \cos(2\pi f_j t + \phi_j)$$

Aside:  $\cos(a + b) = \cos(a)\cos(b) - \sin(a)\sin(b)$

$$= \boxed{C_j} \cos(2\pi f_j t) \boxed{\cos(\phi_j)} - \boxed{C_j} \sin(2\pi f_j t) \boxed{\sin(\phi_j)}$$

$A_j$   $B_j$

$$= \boxed{A_j} \cos(2\pi f_j t) + \boxed{B_j} \sin(2\pi f_j t)$$

Decompose  $V[t]$  into sine/cosine or amplitude/phase.

# Idea: Power spectrum

**Q:** How do we find  $A_j$  and  $B_j$  ?

**A:** Consider  $A_j$  and use **orthogonality of sinusoids**.

$$\int_0^T \cos(2\pi f_j t) \cos(2\pi f_k t) dt = \begin{cases} 0 & \text{if } f_j \neq f_k \\ T/2 & \text{if } f_j = f_k \end{cases}$$

Choose  $T$  so  $f_j$  and  $f_k$  complete an integer number of cycles.

frequencies

integrate over time

# Idea: Power spectrum

*Python*

# Idea: Power spectrum

**Q:** How do we find  $A_j$  and  $B_j$  ?

**A:** Consider  $A_j$  and use **orthogonality of sinusoids**.

$$\int_0^T \cos(2\pi f_j t) \sin(2\pi f_k t) dt = 0 \quad \text{for all } f_j, f_k$$



sine

## Idea: Power spectrum

Return to our original equation

$$V[t] = \sum_j A_j \cos(2\pi f_j t) + B_j \sin(2\pi f_j t)$$

Pick frequency  $f_k$ , multiply both sides by  $\cos(2\pi f_k t)$ , and integrate over time ...

$$\begin{aligned} \int_0^T V[t] \cos(2\pi f_k t) dt &= \int_0^T \sum_j A_j \cos(2\pi f_j t) \cos(2\pi f_k t) dt \\ &\quad + \int_0^T \sum_j B_j \sin(2\pi f_j t) \cos(2\pi f_k t) dt \end{aligned}$$

Consider each integral ...

# Idea: Power spectrum

by orthogonality

$$\int_0^T \sum_j A_j \cos(2\pi f_j t) \cos(2\pi f_k t) dt = 0 \text{ if } f_j \neq f_k, \text{ or } T/2 \text{ if } f_j = f_k$$

$$\int_0^T \sum_j B_j \sin(2\pi f_j t) \cos(2\pi f_k t) dt = 0$$

$A_k T/2$  if  $j = k$ , 0 otherwise

$$\begin{aligned} \text{So } \int_0^T V[t] \cos(2\pi f_k t) dt &= \int_0^T \sum_j A_j \cos(2\pi f_j t) \cos(2\pi f_k t) dt \\ &+ \int_0^T \sum_j B_j \sin(2\pi f_j t) \cos(2\pi f_k t) dt \\ &= A_k T/2 \end{aligned}$$



## Idea: Power spectrum

$$\int_0^T V[t] \cos(2\pi f_k t) dt = A_k T/2$$

Solve for  $A_k$

$$A_k = \frac{2}{T} \int_0^T V[t] \cos(2\pi f_k t) dt$$

We can solve for amplitude  $A_k$

Depends on observed data  $V[t]$  multiplied by cosine we choose ( $f_k$ )

# Idea: Power spectrum

Return to our original equation

$$V[t] = \sum_j A_j \cos(2\pi f_j t) + B_j \sin(2\pi f_j t)$$

**Q:** How do we find  $A_j$  and  $B_j$  ?

**A:** Consider  $A_j$  and use **orthogonality of sinusoids**.

Similarly

$$A_k = \frac{2}{T} \int_0^T V[t] \cos(2\pi f_k t) dt$$

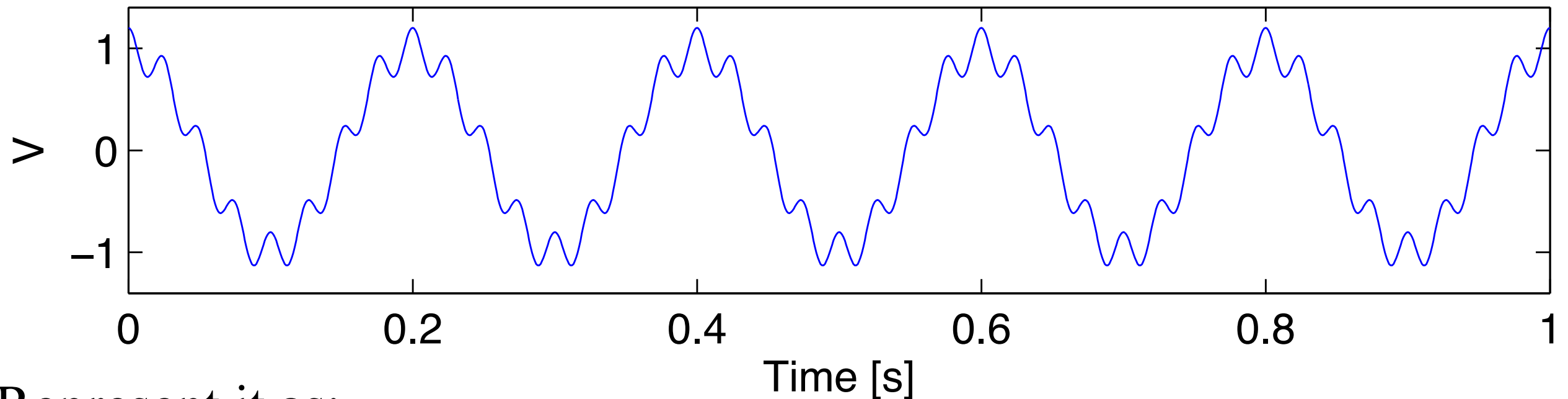
$$B_k = \frac{2}{T} \int_0^T V[t] \sin(2\pi f_k t) dt$$

Big idea: We can decompose  $V[t]$  into a sum of sin/cos functions, and we know how to find the amplitudes  $A_j, B_j$

## Example: rhythmic signal

**Q:** So what?

**A:** Represent  $V[t]$  in a simpler way ... remember:



Represent it as:

$$V[t] = A_1 \cos(2\pi f_1 t) + A_2 \cos(2\pi f_2 t)$$

1.0                      5 Hz                      0.1                      40 Hz

To represent  $V[t]$  we need 4 numbers:

Amplitudes =  $\{1, 0.1\}$

Frequencies =  $\{5, 40\}$  Hz

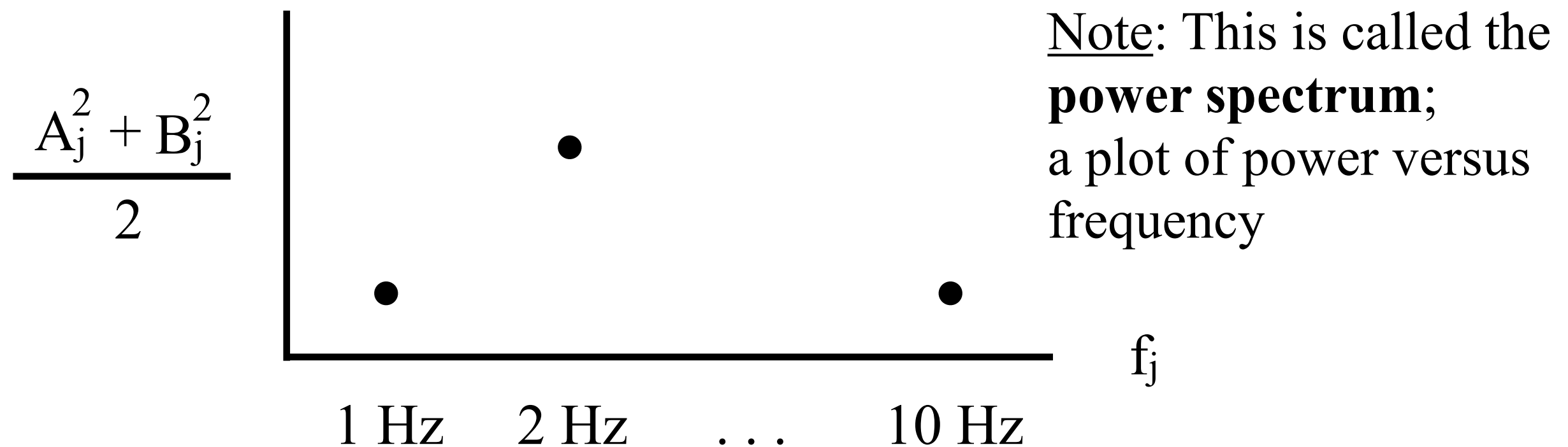
These 4 numbers completely summarize the data.

## Plot: Power spectrum

We can represent these amplitudes and frequencies graphically:

Plot:  $\frac{A_j^2 + B_j^2}{2}$  versus  $f_j$  for each  $n$

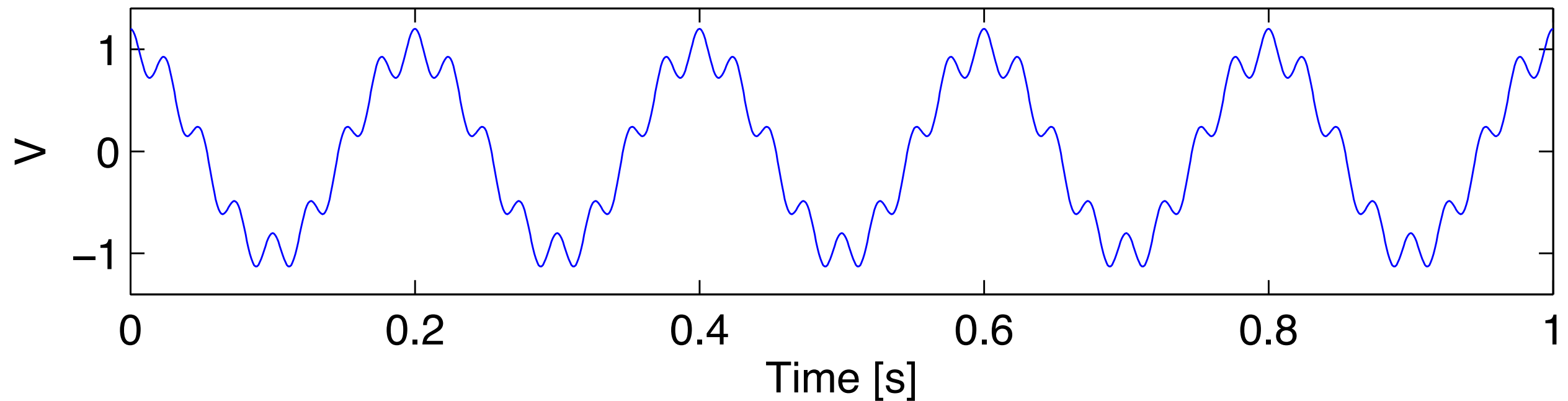
Note: The summed amplitudes squared.  
Called the “power”



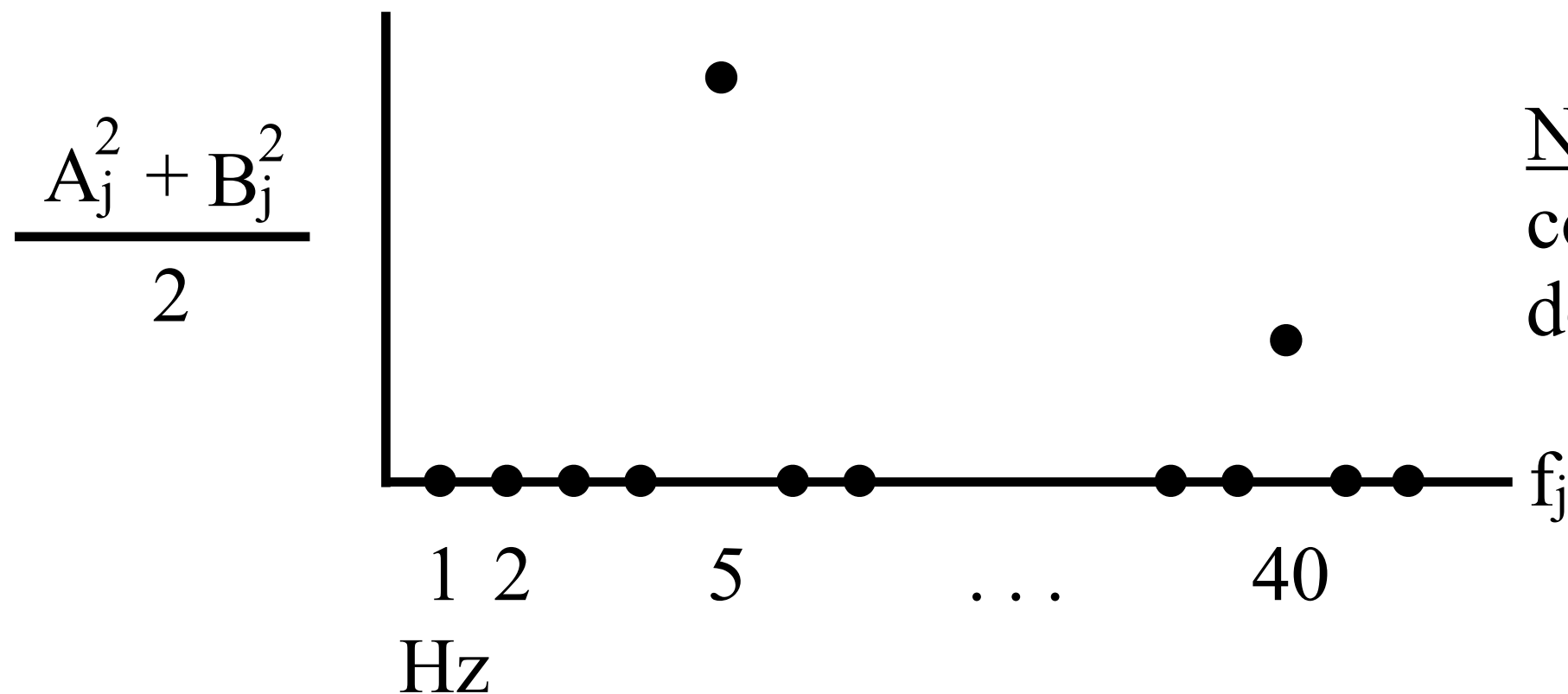
The peaks represent the dominant rhythms in the signal.

## Example: rhythmic signal

**Ex:**



Plot the power spectrum:

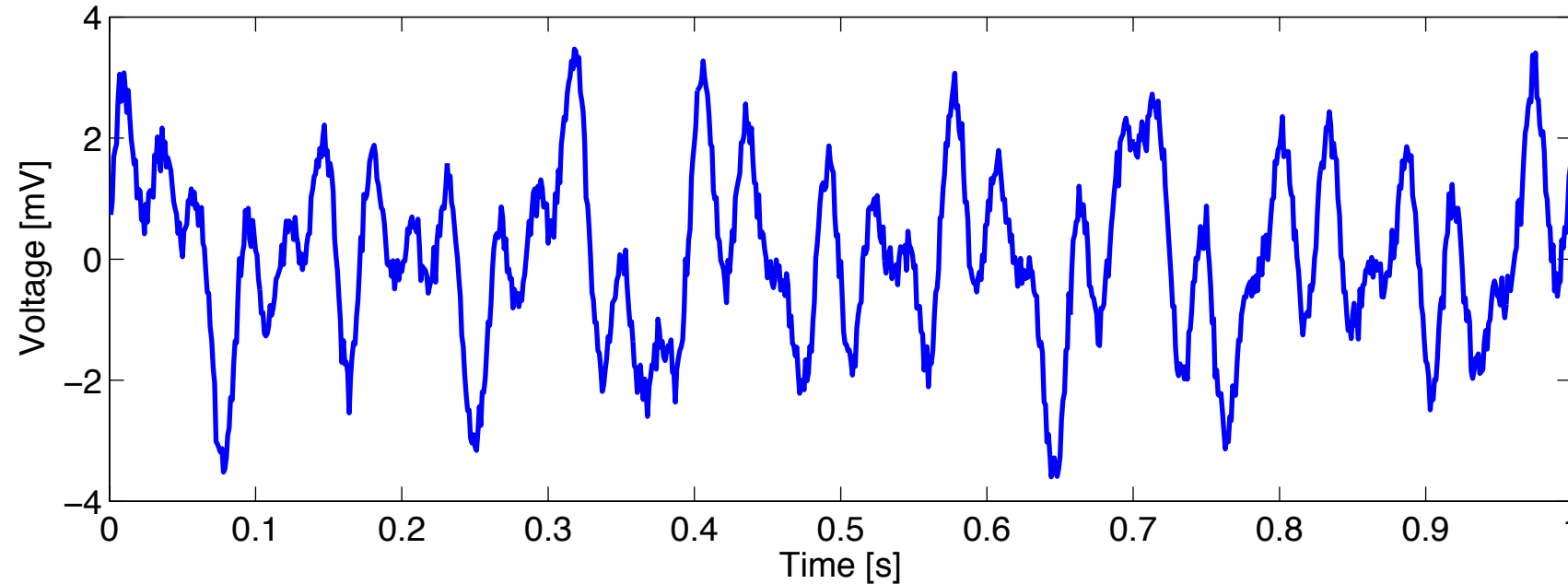


Note: The peaks correspond to the dominant rhythms

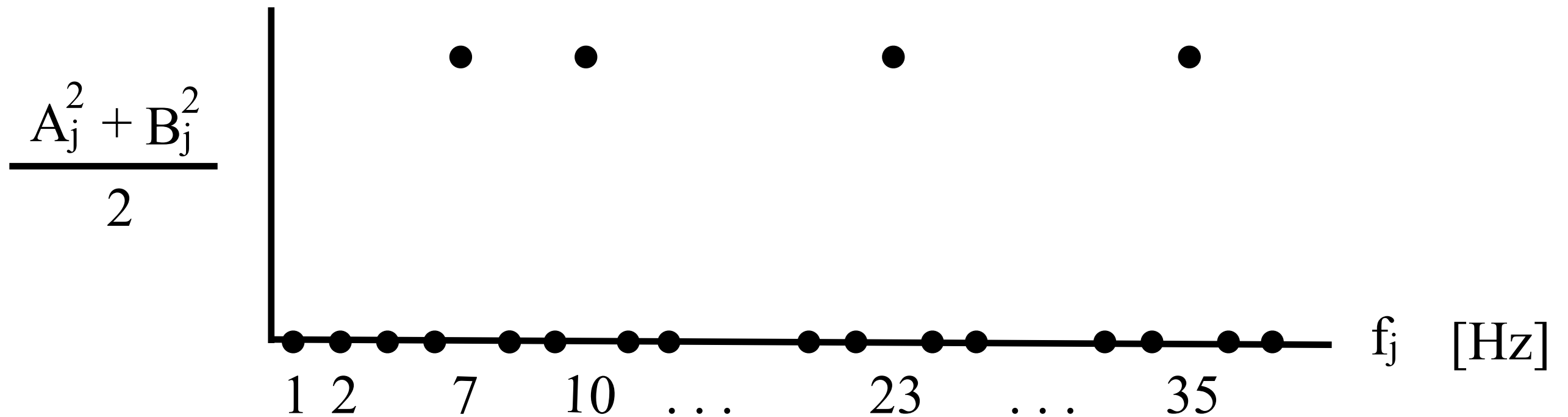
# Example: rhythmic signal

Ex.

$V =$



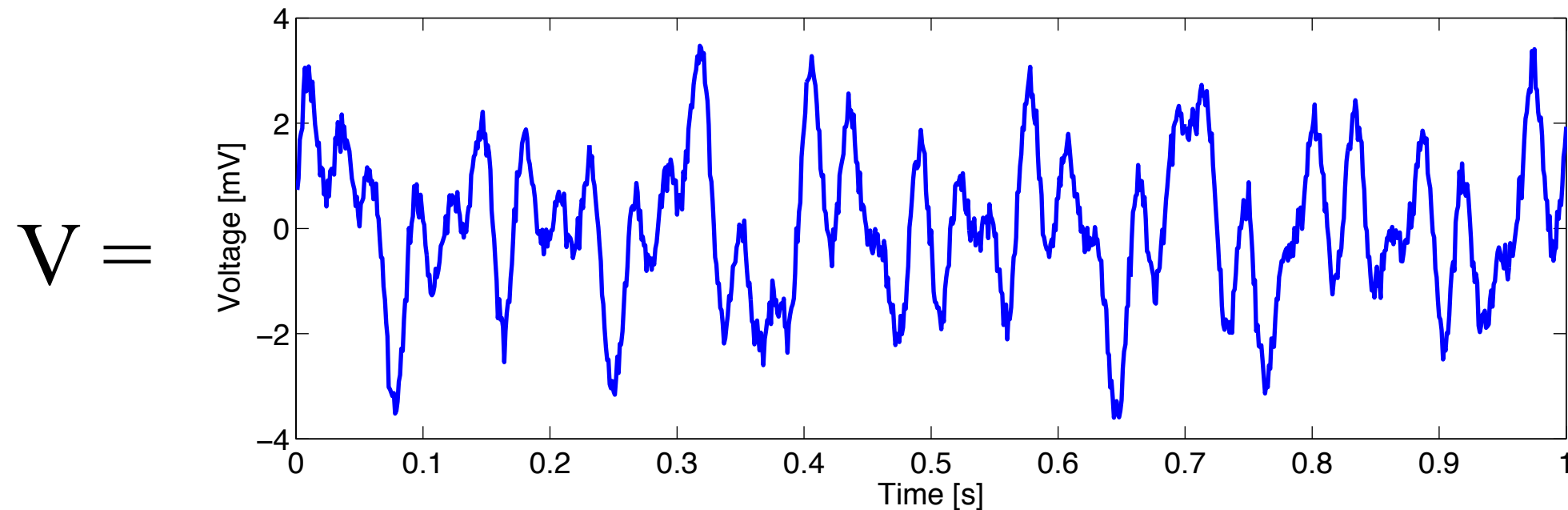
Note: It's complicated Plot the power spectrum:



**Q:** What's happening here?

## Example: rhythmic signal

So, by computing the power spectrum, we find the complicated signal:



We find it's the sum of 4 sinusoids at frequencies:

7 Hz, 10 Hz, 23 Hz, and 35 Hz

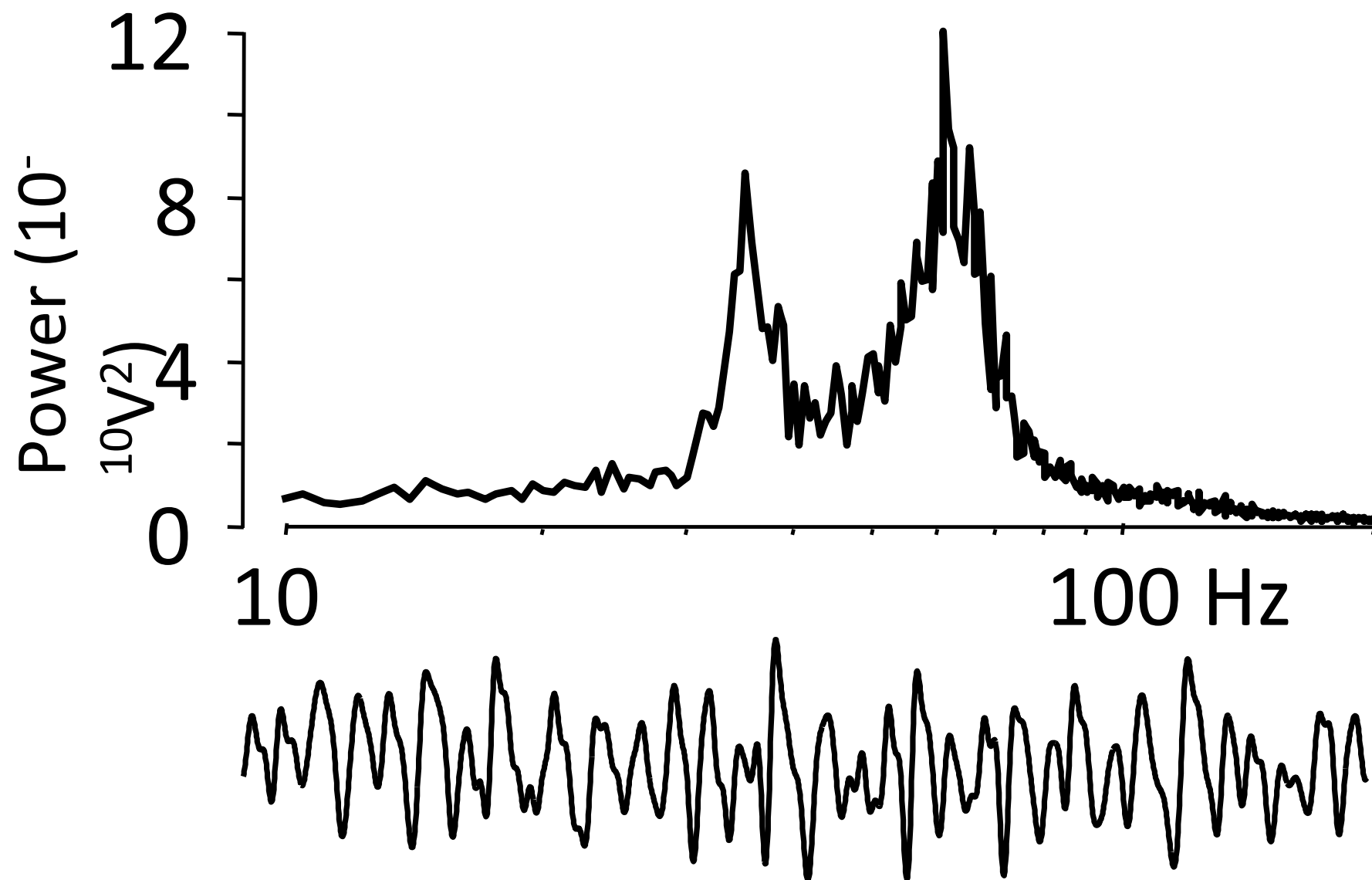
A much simpler representation of brain activity.



## Example: Real world

**Q:** What does the power spectrum of real-world brain signals look like?

**Ex.** From a slice of rat cortex:



[Ainsworth et al, Neuron, 2012]

**Q:** What rhythms are dominant?

# Next

A simple answer to one big rhythm question ...