# **DFT** embodied

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### Goal

#### show how

- I) neural dynamic fields can be coupled to time-varying sensory inputs
- 2) neural dynamic fields can control motor systems in closed loop

# Driving fields from sensory signals

#### robot that orients toward sound sources



[from Bicho, Mallet, Schöner, Int J Rob Res,2000]



### Sensory surface

#### each microphone samples heading direction



# Sensory input

each microphone provides input to the field = loudness \* sensitivity cone



# Detection instability as intensity of sound source increases





[from Bicho, Mallet, Schöner: Int. J. Rob. Res., 2000]



# Target selection in the presence of two sources



# Robust estimation in the presence of outliers



#### Tracking moving sound source 60 time $\psi^{350}$ 60 time $\psi^{350}$

# Working memory



[from Bicho, Mallet, Schöner: Int J Rob Res 19:424(2000)]

## Conclusion I

- Neural dynamic fields can be directly driven from timevarying sensory inputs
- the attractor states and their bifurcations enable

detection

l robust estimation, tracking

selection

working memory



# How may neural dynamic fields generate the behavior?

 challenge I: generating behavior
involves
"behavioral dynamics"
(Braitenberg)



# How may neural dynamic fields generate the behavior?

note the mere "reading out" of a peak's location...



### Generating behavior entails dynamics

behavioral dynamics of a vehicle

with an attractor at desired heading



Human locomotion described by dynamics of heading direction

humans walking in virtual reality under the influence of targets and obstacles



[Warren, Fajen et al, 2003]





## Heading direction

- Neural evidence for head-orientation cells... that function as heading direction representation
- Neural attractor dynamics (neural field) for heading direction



[McNaughton et al., Nature reviews neuroscience 2006]

## Neural dynamics of path integration



[McNaughton et al., Nature reviews neuroscience 2006]

# From field to behavioral dynamics

- standard idea:  $\sigma(u)$ ~ probability density
- but: normalization!
- => problem when there is no peak: divide by zero!

$$\phi_{\text{peak}} = \frac{\int d\phi \ \phi \ \sigma(u(\phi, t))}{\int d\phi' \ \sigma(u(\phi', t))}$$



#### Erect an attractor rather than "read out"



$$\begin{split} \dot{\phi} &= -\left[\int d\phi' \sigma(u(\phi',t))\right](\phi - \phi_{\text{peak}}) \\ &= -\int d\phi' \; (\phi - \phi') \; \sigma(u(\phi',t)) \end{split}$$





# Challenge

In organism (rather than vehicles), movement is generated by muscles which receive descending activation that is graded (rate code)... and temporally structured (timing)

### Rate code in motor neurons

<sup>=</sup>orce (mN)

a single action potential of<sup>A</sup> a motor neuron elicits a transient force pulse in a "fast twitch" motor unit of a muscle

higher firing rates of the motor neuron induce higher peak force of the motor unit



[Kandel, Schwartz et al, 2021]

### So movement involves

- generation of a neural activity pattern that rises and falls to generate transient muscle activity...
- => "translation into rate code"

### Motor cortical neural timers



[Moran, Schwartz, J Neurophys 1999]

## Neural oscillator model of timing

- standard excitatoryinhibitory neural population dynamics => oscillations/active transients
- field of such oscillators for different peak velocities/ amplitudes

[Zibner, Tekülve, Schöner, ICDL 2015; Schöner, Tekülve, Zibner, 2019]]



#### Localized input triggers transients/ oscillations in such fields



[J-S Jokeit dissertation 2022]



# Muscles generate force in closed loop: behavioral dynamics



[Ramadan et al, 2022]

#### Conclusion 2

#### Neural dynamics drive behavior by steering a behavioral dynamics