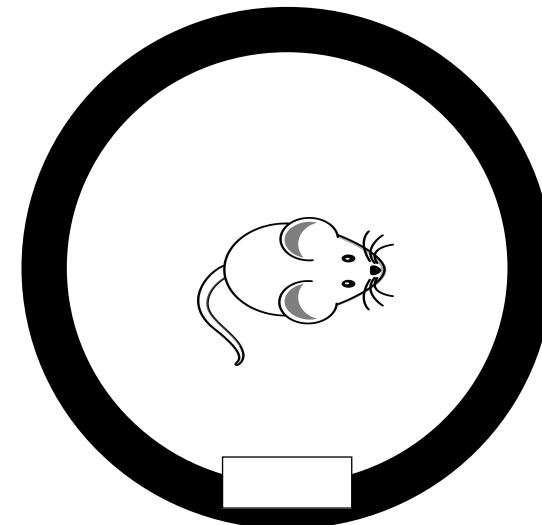
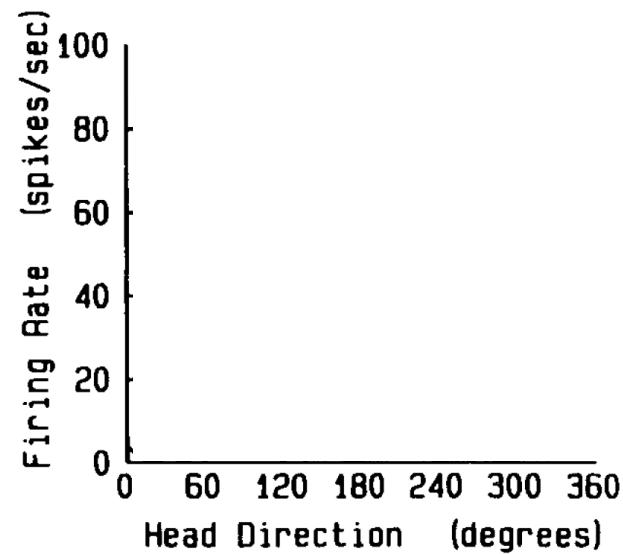


Compass in our head

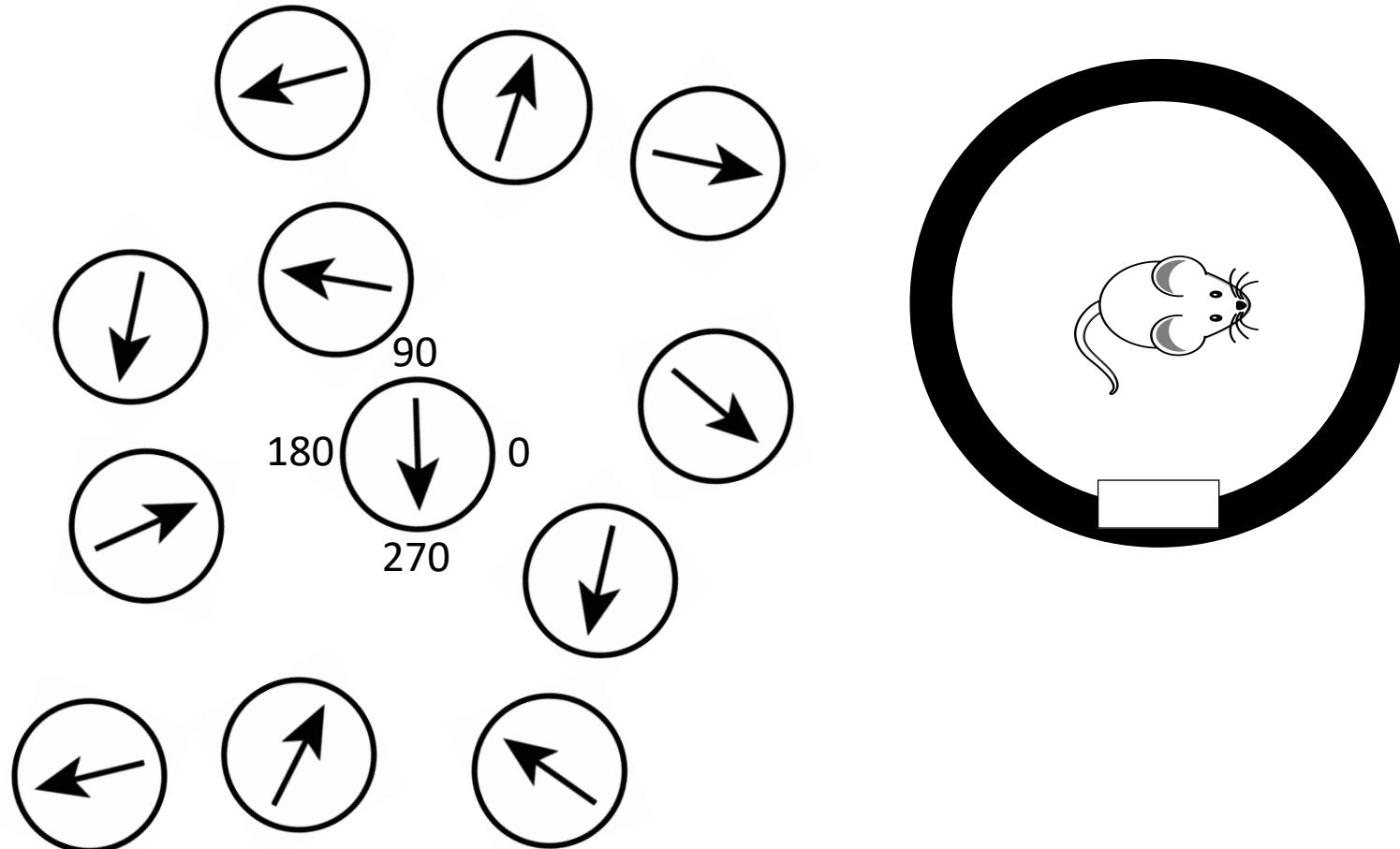
– Head direction system –

Head Direction Cells

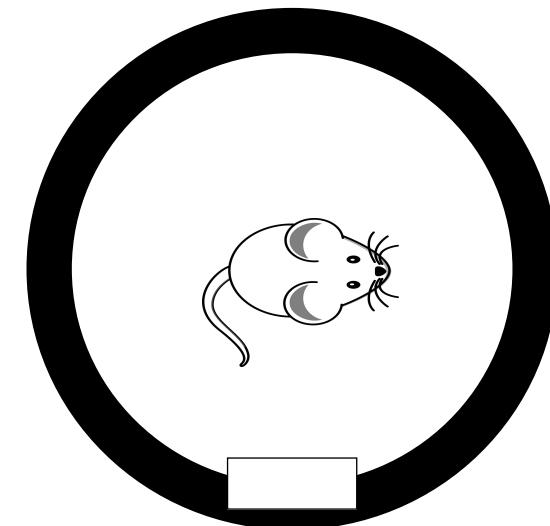
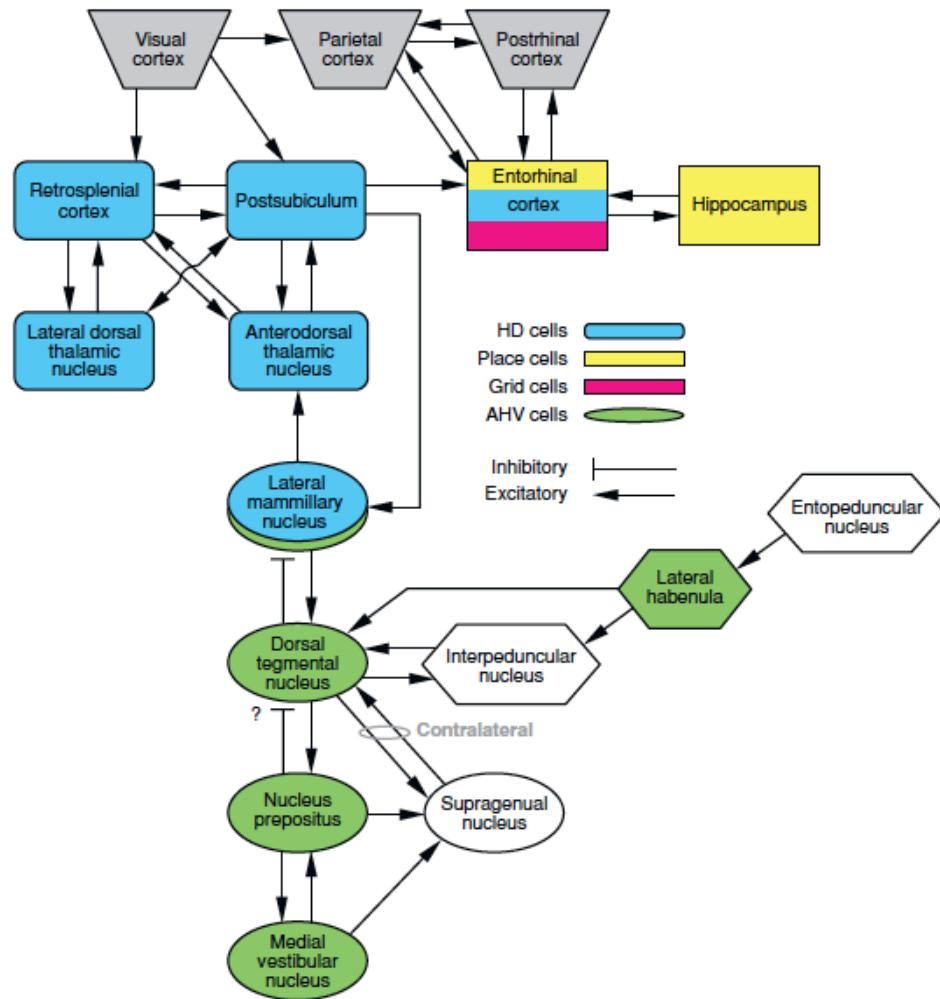
Single Neuron Tuning Curve



Head Direction Cells



Head Direction Cells





Baird, el Jundi et al.



Reppert, Heinze et al.



von Frisch, Menzel,
Srinivasan et al.

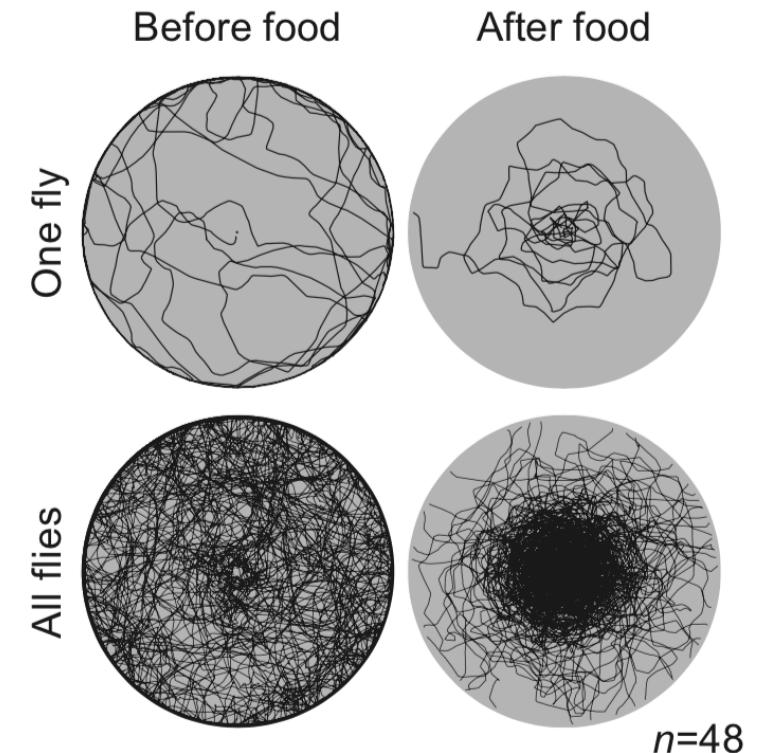


Condition: Oe-

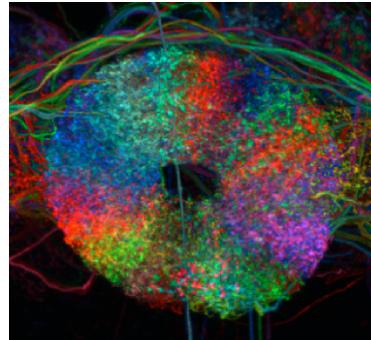
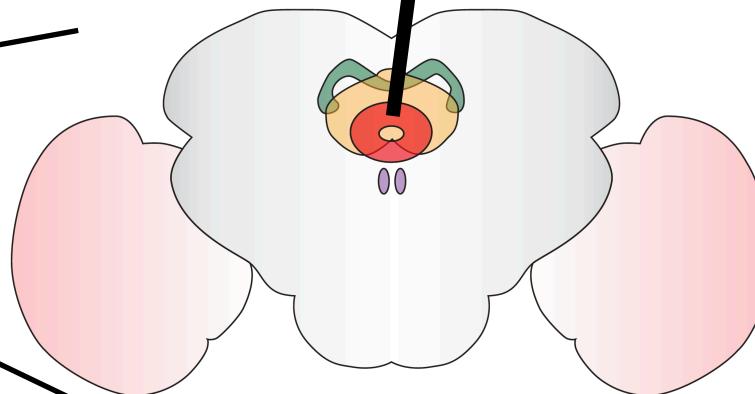
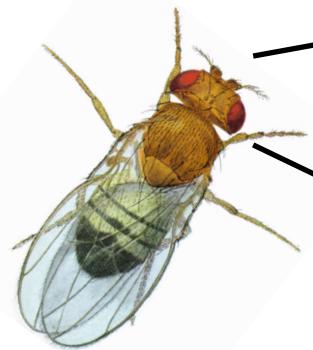
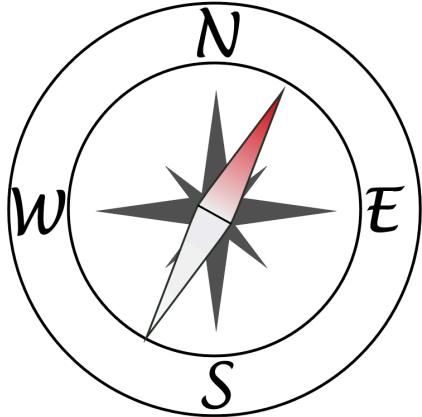
Fly: starved oe- female

Food source: yeast drop

Walking trajectory before first food encounter

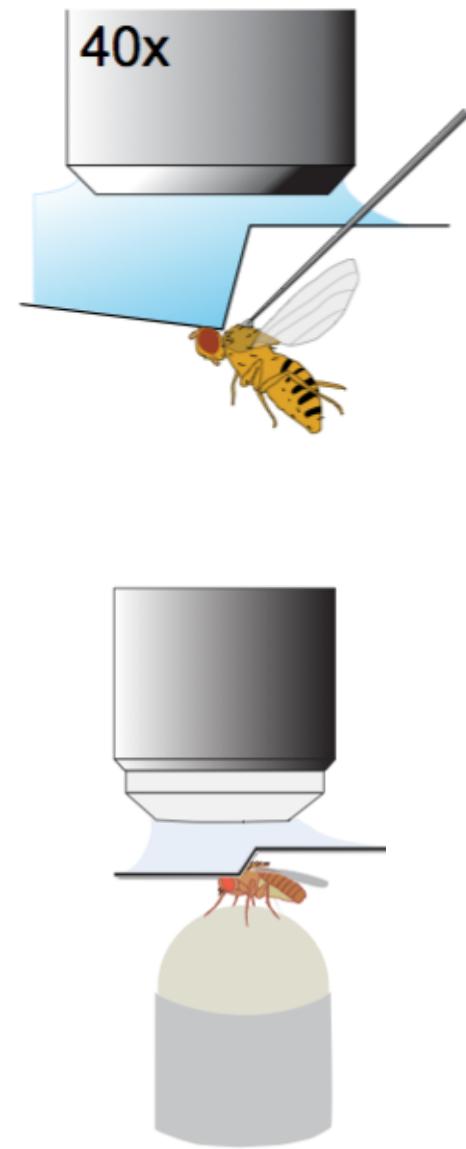
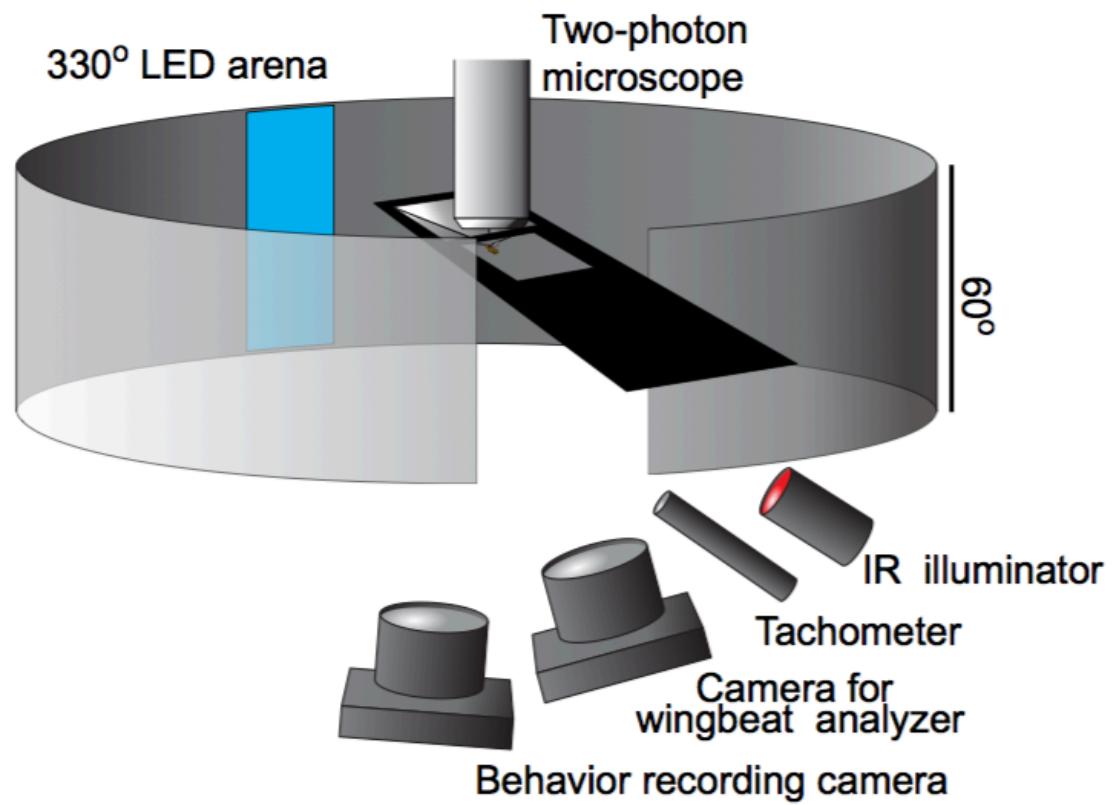


Compass neurons

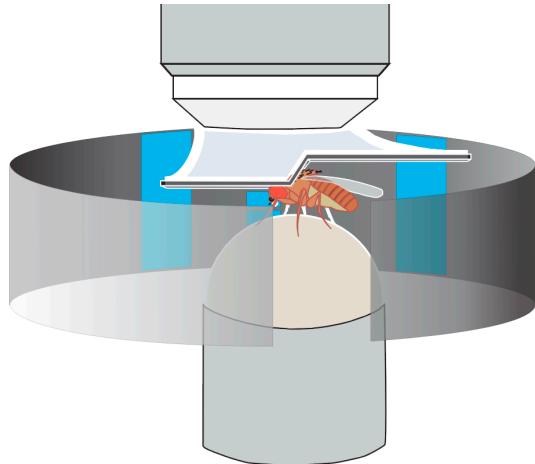


E-PG neuron (a.k.a. compass neurons)
in Ellipsoid Body

Jenett et al. (2012) *Cell Reports*
Wolff et al. (2015) *J Comp Neurol*



An unique representation of orientation



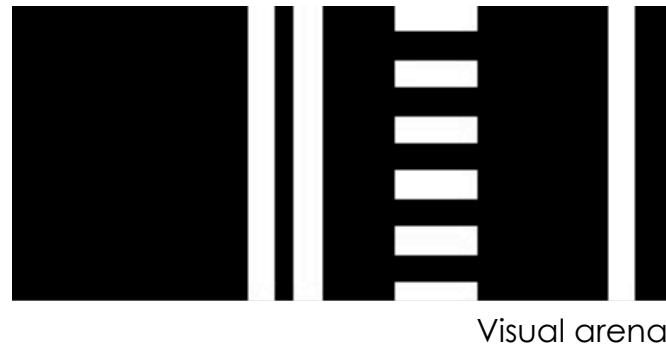
A fly is walking on an air-floated ball



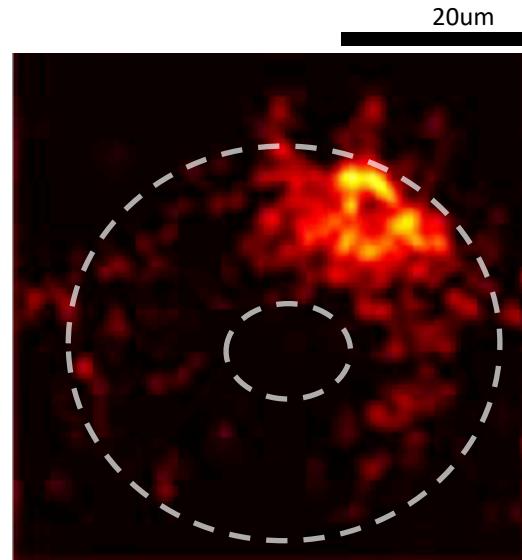
The rotation of the ball is measured to infer the fly's intention of turning

Tethered fly

Seelig & Jayaraman (2015) *Nature*



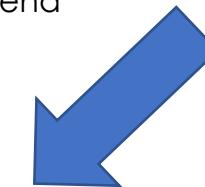
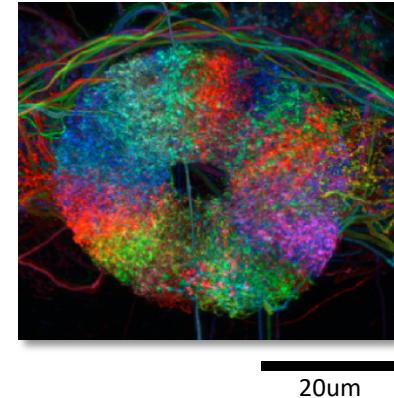
Visual scene on the VR screen that the fly is watching



Calcium imaging of compass neurons with GCaMP

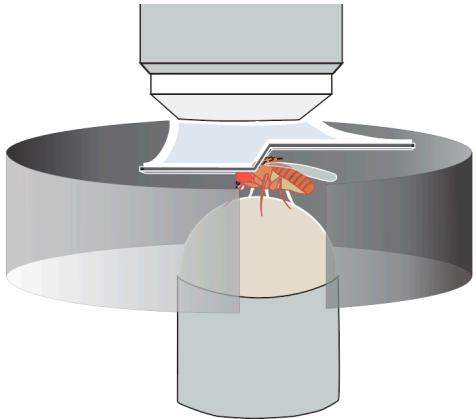


GCaMP is expressed in the entire population of compass neurons

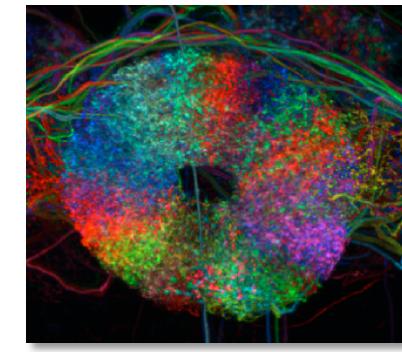
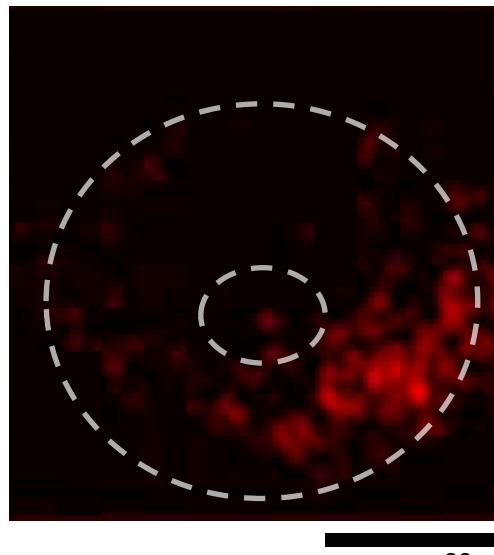
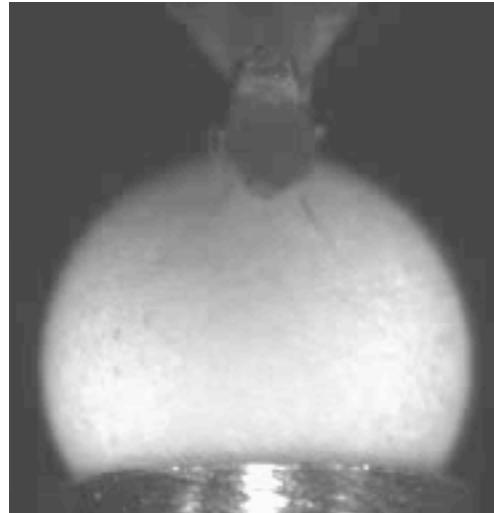


We are looking at the neural activity of the entire population of fly compass neurons

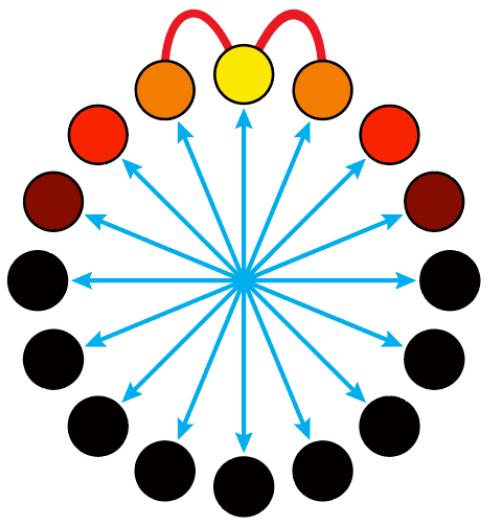
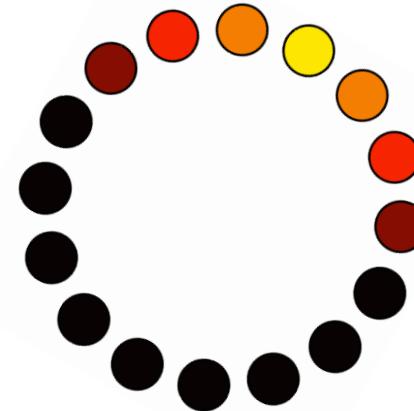
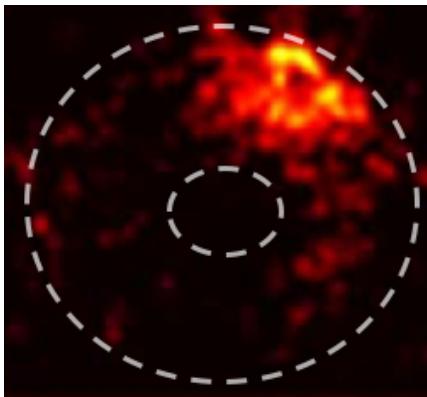
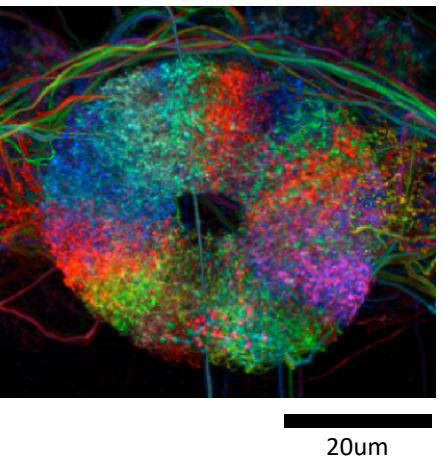
Representation is maintained in darkness



The fly is in a complete darkness.

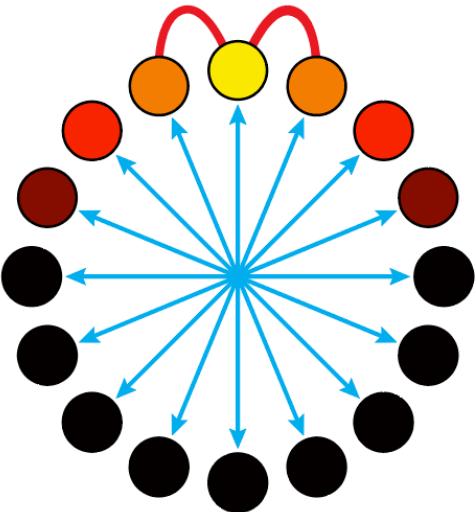


20um



Ring Attractor

$$\tau \frac{\partial f_n}{\partial t} = -f_n + \left[\alpha f_n + D(f_{n+1} + f_{n-1} - 2f_n) - \beta \sum_{m=0}^{N-1} f_m + 1 \right]_+$$



Ring Attractor

$$\tau \frac{\partial f_n}{\partial t} = -f_n + \left[\alpha f_n + D(f_{n+1} + f_{n-1} - 2f_n) - \beta \sum_{m=0}^{N-1} f_m + 1 \right]_+$$

f_n is the activity of a compass neuron with an index, n

$N = 32$ is the number of compass neurons (i.e, there are 32 similar equations)

$[x]_+$ is a rectification function (that is, if $x < 0$, it becomes zero).

$\tau = 0.05$ is a decay time constant (if there is no input, the neuron's activity decays)

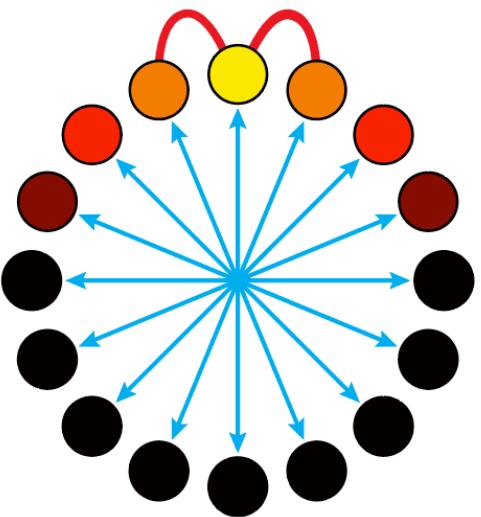
$\alpha = 2.62$ is the strength of self-recurrent connection

$\beta = 1.96$ is the strength of uniform inhibition

$D = 5.19$ is the strength of recurrent excitation between neighboring compass neurons

The constant term 1 is a uniform excitatory drive that generates activity in the system.

The initial value of f_n can be set `yinit = np.random.random(size=(32))*0.05`



Ring Attractor

$$\tau \frac{\partial f_n}{\partial t} = -f_n + \left[\alpha f_n + D(f_{n+1} + f_{n-1} - 2f_n) - \beta \sum_{m=0}^{N-1} f_m + 1 \right]_+$$

```

import numpy as np
from scipy.integrate import solve_ivp

def RingAttractorODEs(t,f):

    tau = 0.05
    alpha = 2.633
    beta = 1.935
    D = 5.187
    n_neurons = 32;

    f0, f1, f2, f3, f4, f5, f6, f7, f8, f9, f10, f11, f12, f13, f14, ..., f31 = f

    alpha*f + D*(np.roll(f,-1) + np.roll(f,1) - 2*f) - beta*f.sum() + 1

```

```

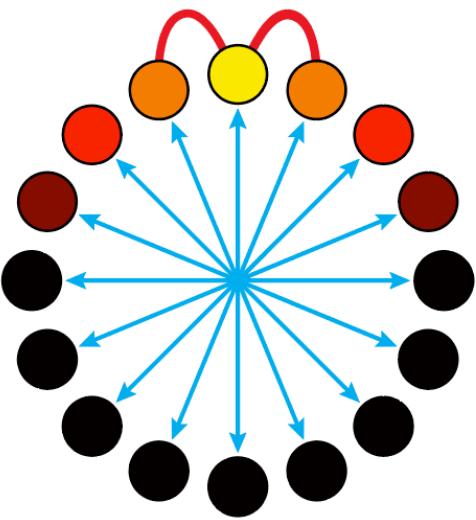
f:      array([100, 200, 300, 400, 500, 600])
f_{n+1}: array([200, 300, 400, 500, 600, 100])
f_{n-1}: array([600, 100, 200, 300, 400, 500])

```

```

f_np1 = f[ list(range(1,n_neurons)) + [0] ]
f_nm1 = f[ [-1] + list(range(n_neurons-1)) ]
D*(f_np1+f_nm1-2*f)

```



Ring Attractor

$$\tau \frac{\partial f_n}{\partial t} = -f_n + \left[\alpha f_n + D(f_{n+1} + f_{n-1} - 2f_n) - \beta \sum_{m=0}^{N-1} f_m + 1 \right]_+$$

```

import numpy as np
from scipy.integrate import solve_ivp

def RingAttractorODEs(t,f):

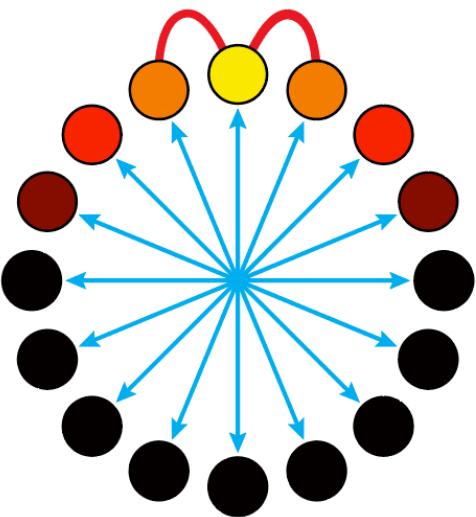
    tau = 0.05
    alpha = 2.6183
    beta = 1.9635
    D = 5.1876
    n_neurons = 32;

    tmp = alpha*f + D*(np.roll(f,-1) + np.roll(f,1) - 2*f) - beta*f.sum() + 1
    tmp[tmp<0] = 0

    dfdt = ( -f + tmp )/tau

    return dfdt

```



Ring Attractor

$$\tau \frac{\partial f_n}{\partial t} = -f_n + \left[\alpha f_n + D(f_{n+1} + f_{n-1} - 2f_n) - \beta \sum_{m=0}^{N-1} f_m + 1 \right]_+$$

```

import numpy as np
from scipy.integrate import solve_ivp

def RingAttractorODEs(t,f):
    .....
    tspan = np.linspace(0, 1, 100)
    f_init = np.random.random(size=(32,))*0.08

    sol = solve_ivp(RingAttractorODEs,
                    [tspan[0], tspan[-1]], f_init, t_eval=tspan)

import matplotlib.pyplot as plt
plt.figure(1, figsize=(12,10))
plt.imshow(sol.y, extent=[0,tspan[-1],32.5,0.5], aspect=tspan[-1]/32/8)

```

