

Motion in flight

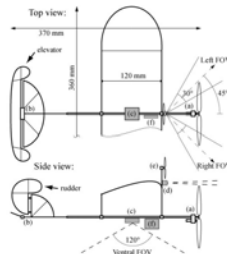
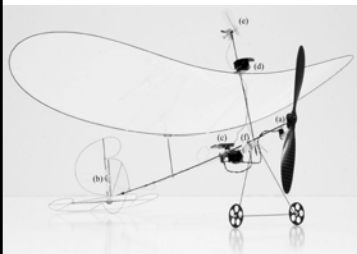
Shih-Chii Liu
Institute for Neuroinformatics
University of Zurich / ETH Zurich
www.ini.uzh.ch/~shih



Outline

- Constraints of micro-aerial vehicles
- Choice of aVLSI motion chips
- Behavioral experiments on a robot using motion sensors
- Prototype motion chip for flyer

MC1 Microflyer



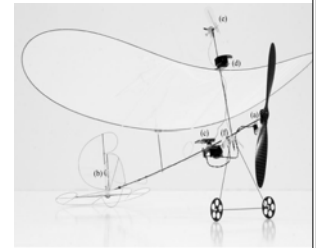
Specs
-65 mAh Li-polymer
-2 linear B/W cameras with 80 pixels covering
120deg FOV, 2 gyros for yaw, pitch, 1 anemometer
-10g, 1.5m/s,
-Optimal latency is 6.4ms, duration is 10mins

(Zufferey, Beyeler, and Floreano, EPFL)



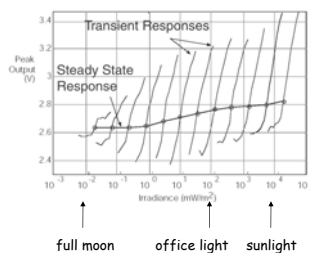
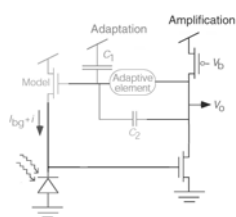
Vision sensor challenges

- Operate
 - at low light levels
 - in wide dynamic range scenes
- Low latency
- Low power
- Small and light



- Plus: Offload computation from on-board microcontroller

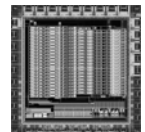
Photoreceptor circuit in most motion chip front-end



(Delbruck, 1993)

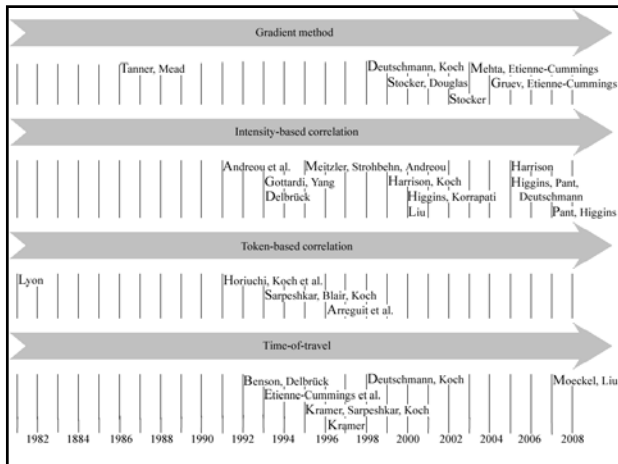
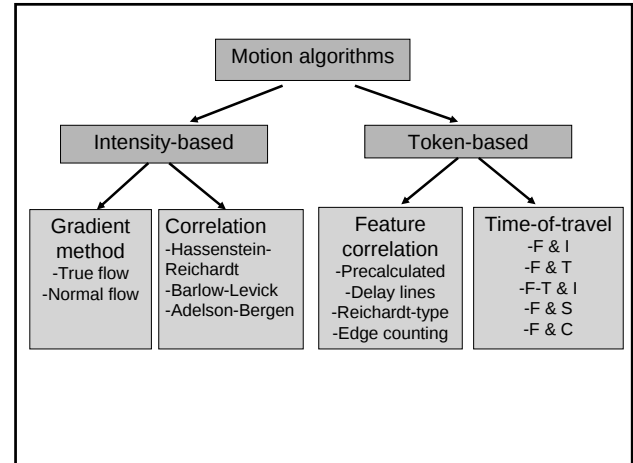
Challenges for neuromorphic vision sensors

- Presence of across-pixel heterogeneity
- Low precision computation
- Longer development time



Outline

- Constraints of micro-aerial vehicles
- **Choice of aVLSI motion chips**
- Behavioral experiments on a robot using motion sensors
- Prototype motion chip for flyer



Optical flow chips

$v = (dl/dt) / (dl/dx)$

Brightness constancy equation

$$\frac{dE(x, y, t)}{dt} = 0$$

$$\Rightarrow \frac{\partial E}{\partial x} u + \frac{\partial E}{\partial y} v + \frac{\partial E}{\partial t} = 0; \quad u = \frac{\partial x}{\partial t}, v = \frac{\partial y}{\partial t}$$

$$\Rightarrow E_x u + E_y v + E_t = 0$$

$$H_{of} = \sum_{ij} \left[\left(E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij} \right)^2 \right]$$

$$\dot{u}_{ij} \propto \left[E_{xij} \left(E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij} \right) \right]$$

$$\dot{v}_{ij} \propto \left[E_{yij} \left(E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij} \right) \right]$$

Horn and Schunk (1981), Fenema and Thompson (1979)

Smoothness constraint

$$H_{of} = \sum_{ij} \left[\left(E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij} \right)^2 \right]$$

$$+ \rho_{ij}^x \left(\left(\Delta^x u_{ij} \right)^2 + \left(\Delta^x v_{ij} \right)^2 \right) + \rho_{ij}^y \left(\left(\Delta^y u_{ij} \right)^2 + \left(\Delta^y v_{ij} \right)^2 \right)$$



Horn and Schunk (1981), Fenema and Thompson (1979)

Bias constraint

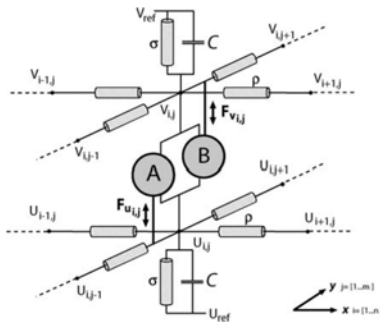
$$H_{of} = \sum_{ij} \left[(E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij})^2 \right] \\ + \rho_{ij}^x \left((\Delta^x u_{ij})^2 + (\Delta^x v_{ij})^2 \right) + \rho_{ij}^y \left((\Delta^y u_{ij})^2 + (\Delta^y v_{ij})^2 \right) \\ + \sigma \left((u_{ij} - u_{ref}^x)^2 + (v_{ij} - v_{ref}^y)^2 \right)$$

Stocker (2002)

Update equations

$$\dot{u}_{ij} = -\frac{1}{C} [E_{xij} (E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij}) \\ - \rho(u_{i+1,j} + u_{i-1,j} + u_{i,j+1} + u_{i,j-1} - 4u_{ij}) \\ + \sigma(u_{ij} - u_{ref})] \\ \dot{v}_{ij} = -\frac{1}{C} [E_{yij} (E_{xij} u_{ij} + E_{yij} v_{ij} + E_{tij}) \\ - \rho(v_{i+1,j} + v_{i-1,j} + v_{i,j+1} + v_{i,j-1} - 4v_{ij}) \\ + \sigma(v_{ij} - v_{ref})]$$

Circuit implementation



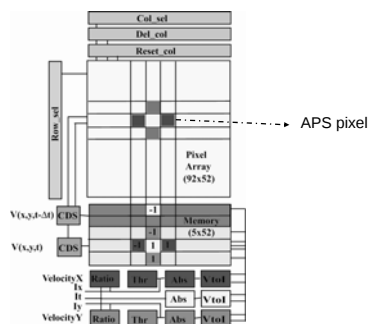
Stocker (2002)

Normal flow chips

$$v_x = -\frac{I_t}{I_x}, v_y = -\frac{I_t}{I_y}$$

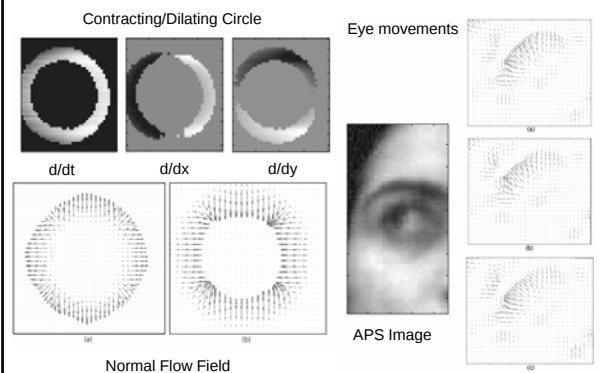
Deutschmann et al (1988), Mehta, Gruev, Etienne-Cummings (2003-)

Visual motion estimation: A normal flow camera

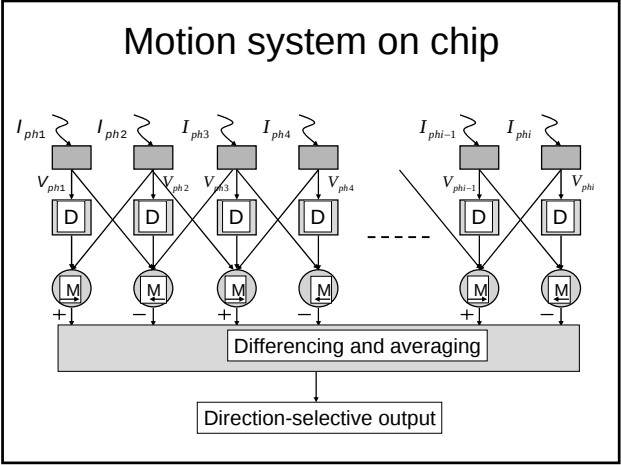
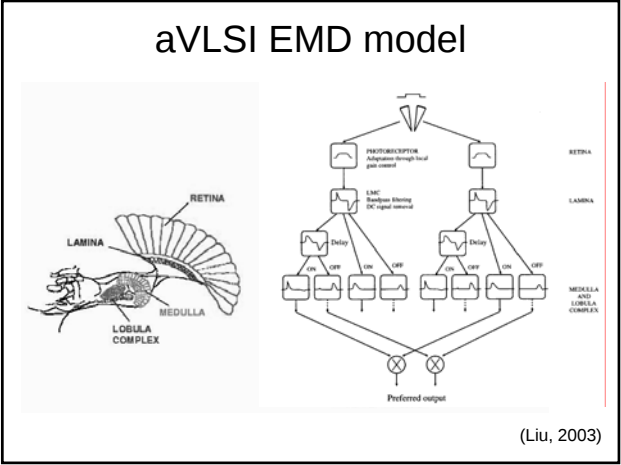
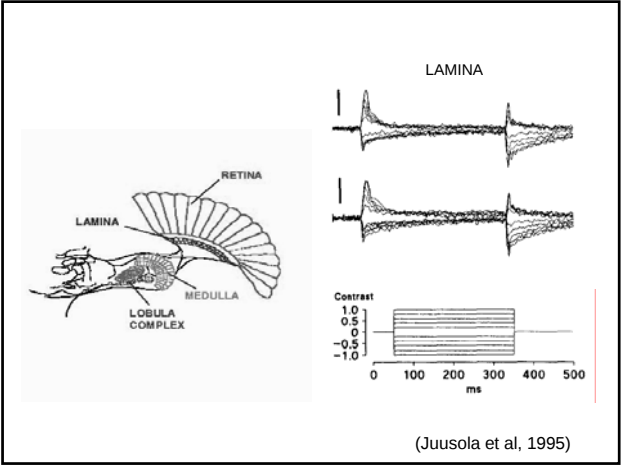
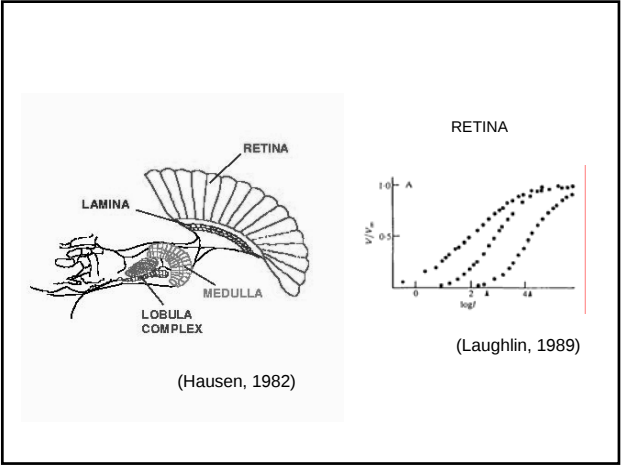
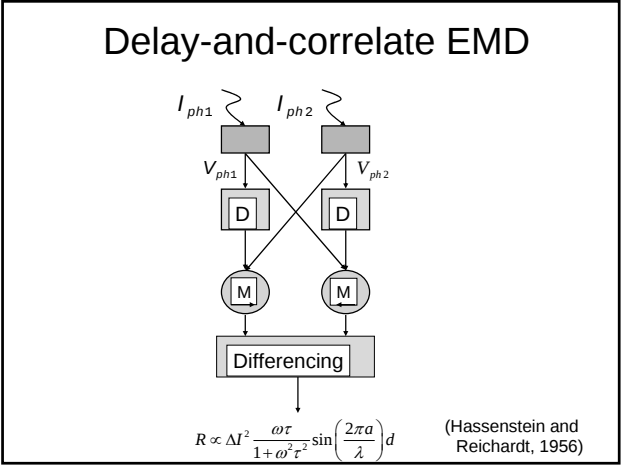
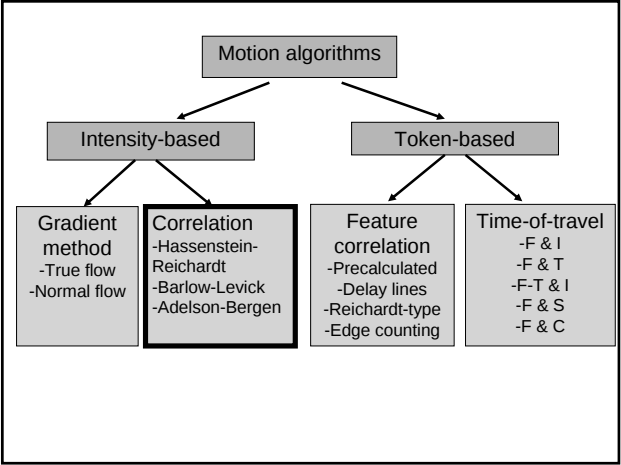


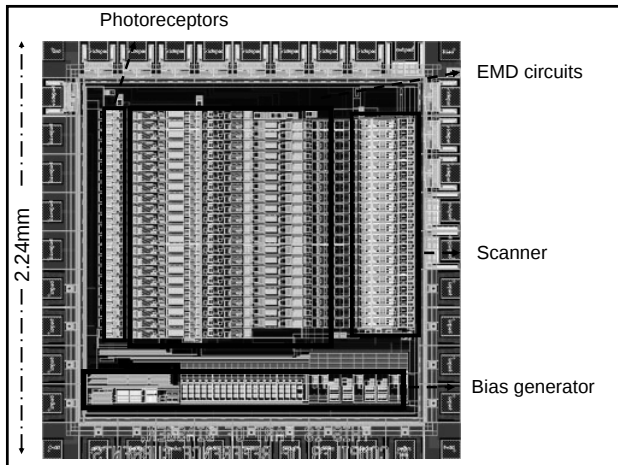
(Mehta and Etienne-Cummings, 2006)

Velocity characterization

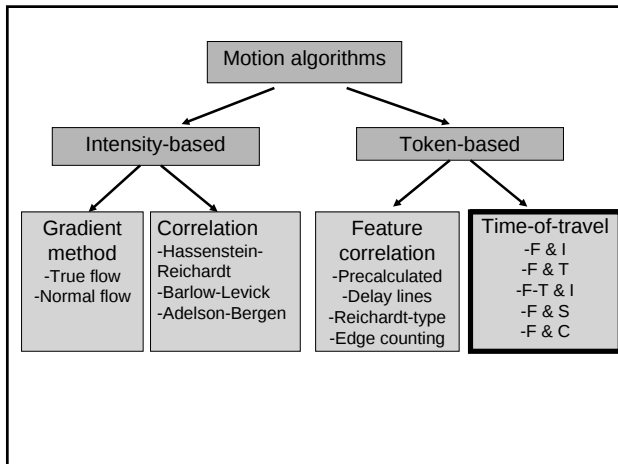
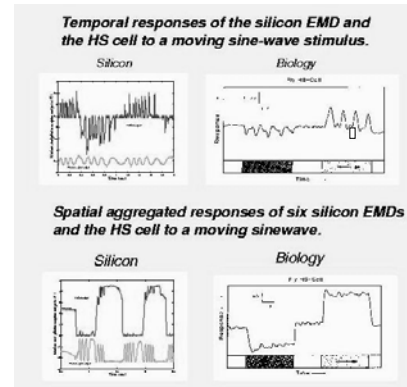


(Mehta and Etienne-Cummings, 2006)

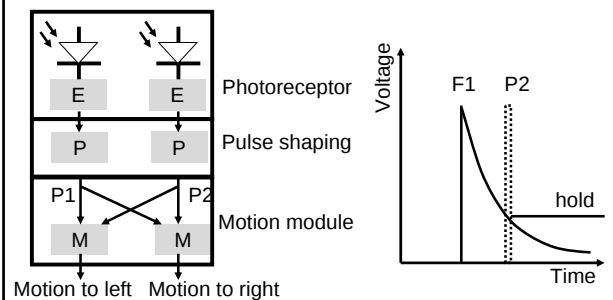




Chip and HS cell responses

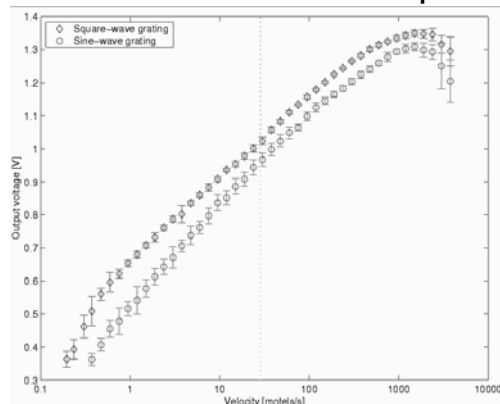


Facilitate-and-sample algorithm



(Kramer, Sarpeshkar, and Koch, 1997)

Results from FS chip



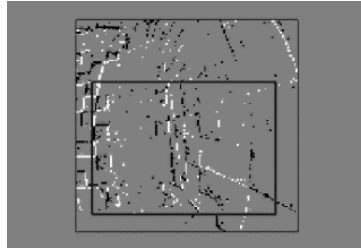
Spike-based approach (more than one chip)

Measuring motion using the temporal contrast silicon retina (Delbruck and Lichtsteiner)



Key specifications: 128x128 pixels, 120dB dynamic range, 2.1%-contrast threshold mismatch, 3kHz pixel bandwidth, 23mW power consumption, 15 μ s-latency

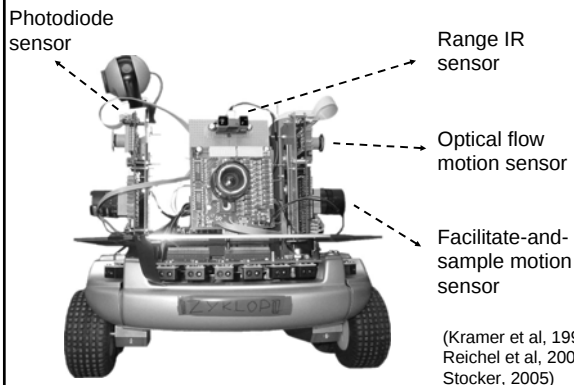
1. Start with raw events that represent temporal contrast
2. Use **spatio-temporal event coincidence** to label events with local edge orientation
3. Label these **OrientationEvents** with local direction and speed
4. Integrate these **DirectionSelectiveEvents** to compute global translational, rotational, and radial motion



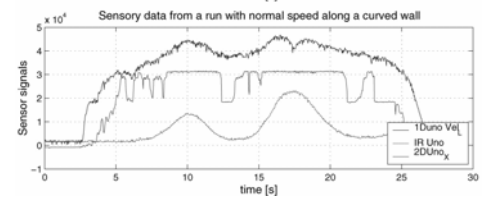
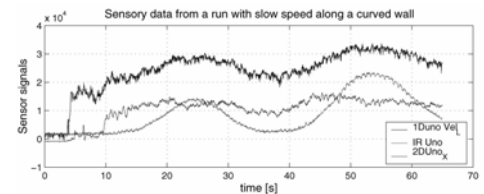
Outline

- Constraints of micro-aerial vehicles
- Choice of aVLSI motion chips
- **Behavioral experiments on a robot using motion sensors**
- Prototype motion chip for flyer

Multi-modal sensors

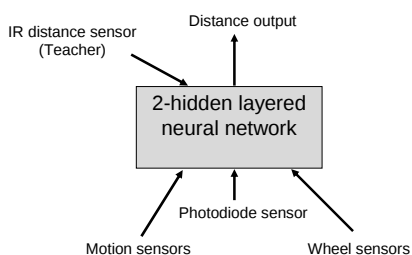


Motion sensor vs IR sensor

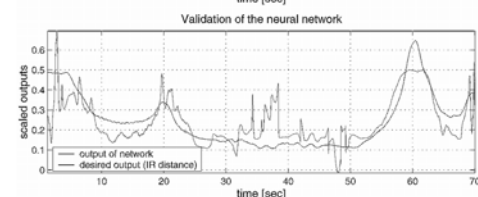
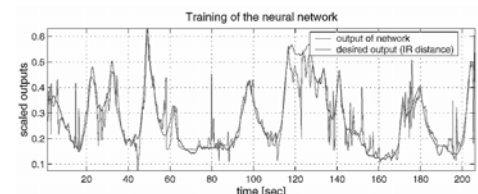


(Reichel et al, 2005)

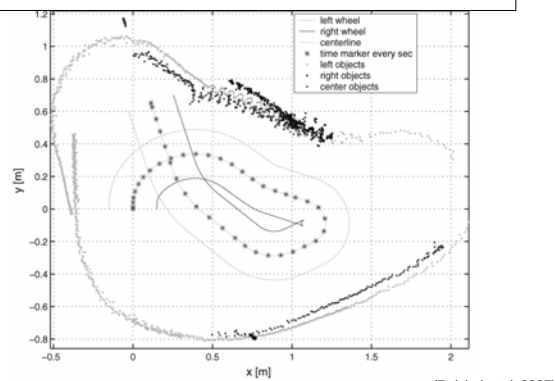
Estimation of distance



Distance estimation with NN



Map of the environment

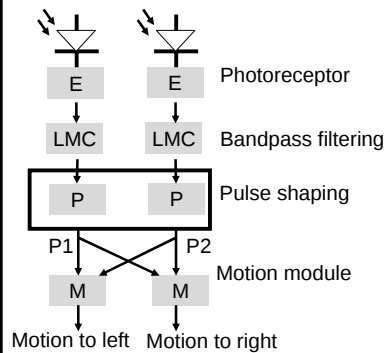


(Reichel et al, 2005)

Outline

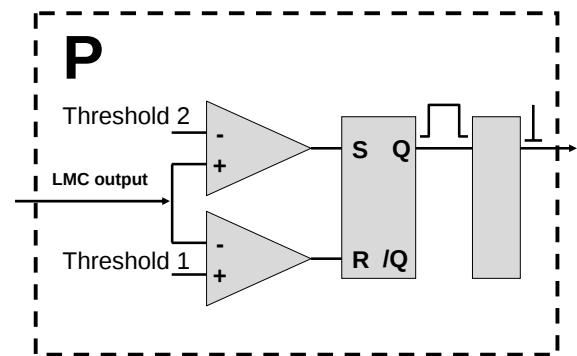
- Constraints of micro-aerial vehicles
- Choice of aVLSI motion chips
- Behavioral experiments on a robot using motion sensors
- **Prototype motion chip for flyer**

Algorithm on MDC2

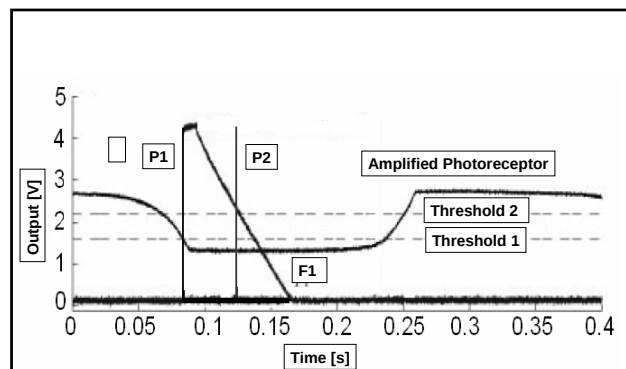
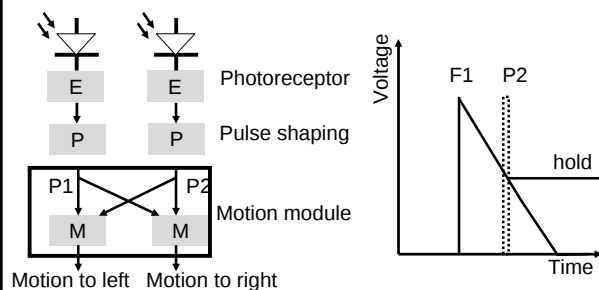


(Moeckel and Liu, 2006)

Pulse-shaping circuit

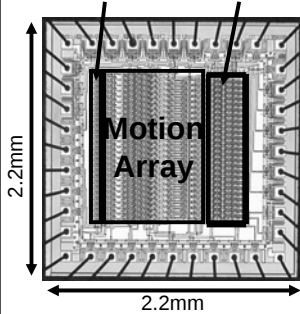


Facilitate-and-sample algorithm



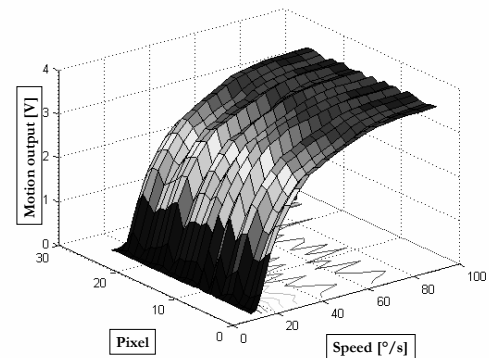
MDC2 Chip

28 photodiodes Scanner circuit

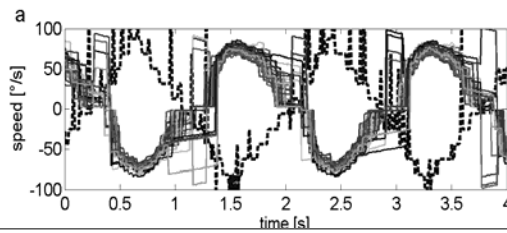


Process	1.5 μ m 2-poly 2-metal 5V CMOS
Number of motion pixels:	26
Pitch size	$68\lambda = 54.4\mu\text{m}$
Minimum contrast:	2.5%
Linear motion output:	2 decades
Illumination range:	>3 decades
Power consumption:	5mW

Average response to pattern with different speeds



MDC2 and gyro



Summary

- Micro-aerial vehicles provide a great platform for showing advantages of motion chips.
- Choices for aVLSI motion chips that match the constraints.
- Challenges ahead – how we use reliably the optical flow information on the flyer.

Acknowledgments

Rico Moeckel
Tobi Delbruck

EPFL

J. C. Zufferey
Antoine Beyeler
Adam Klaptocz
Dario Floreano

Joerg Kramer
Samuel Glauser
Karl Presser
David Liechti
Lukas Reichel

- Thanks to Swiss National Foundation, INI

THE END