

Architecture

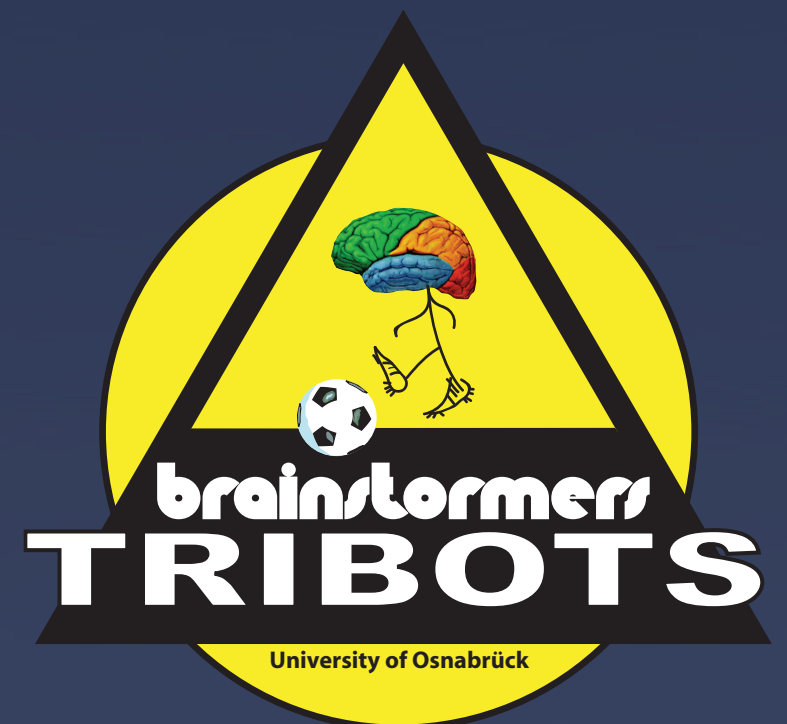
Sascha Lange, Christian Müller, Stefan Welker

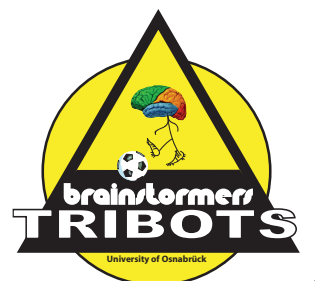
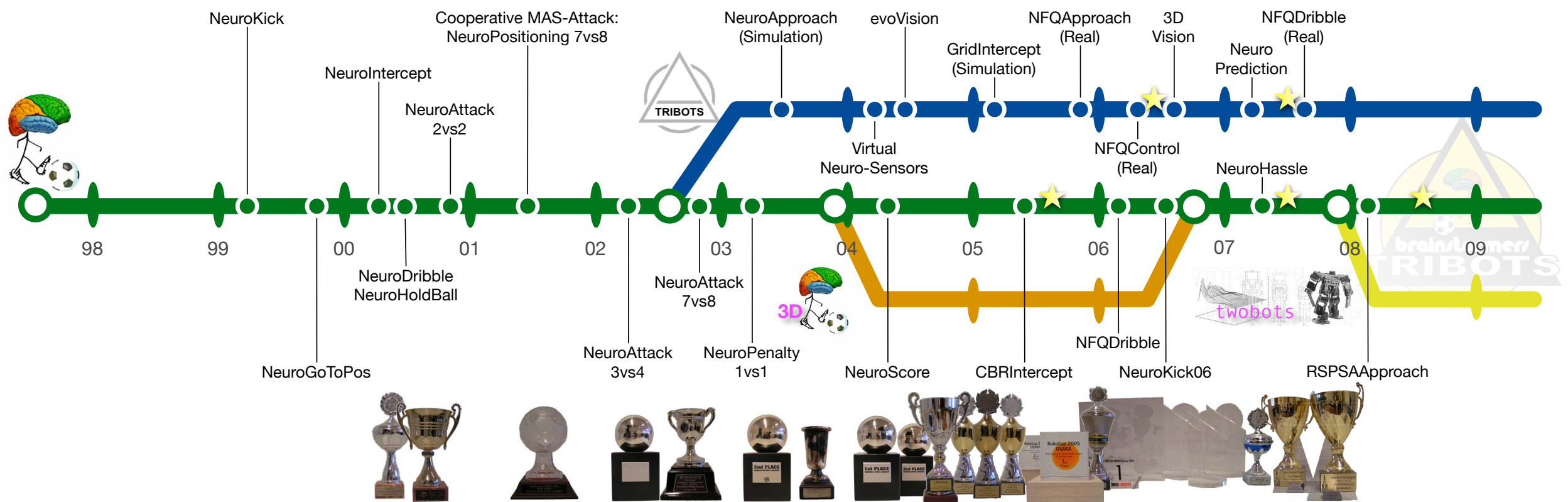
Brainstormers Tribots

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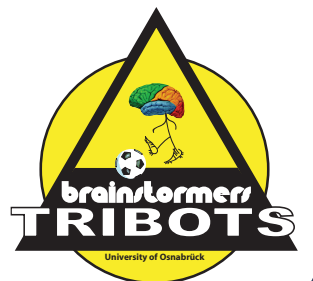
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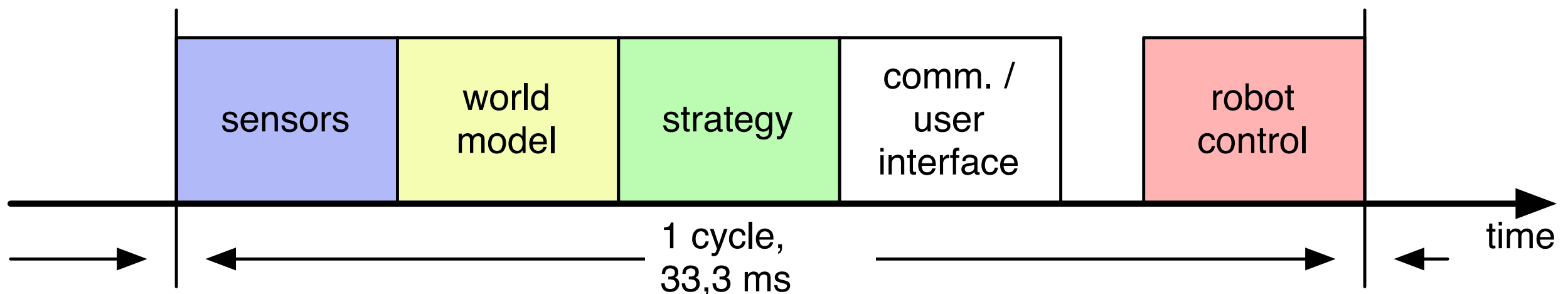
Relevant Design Goals

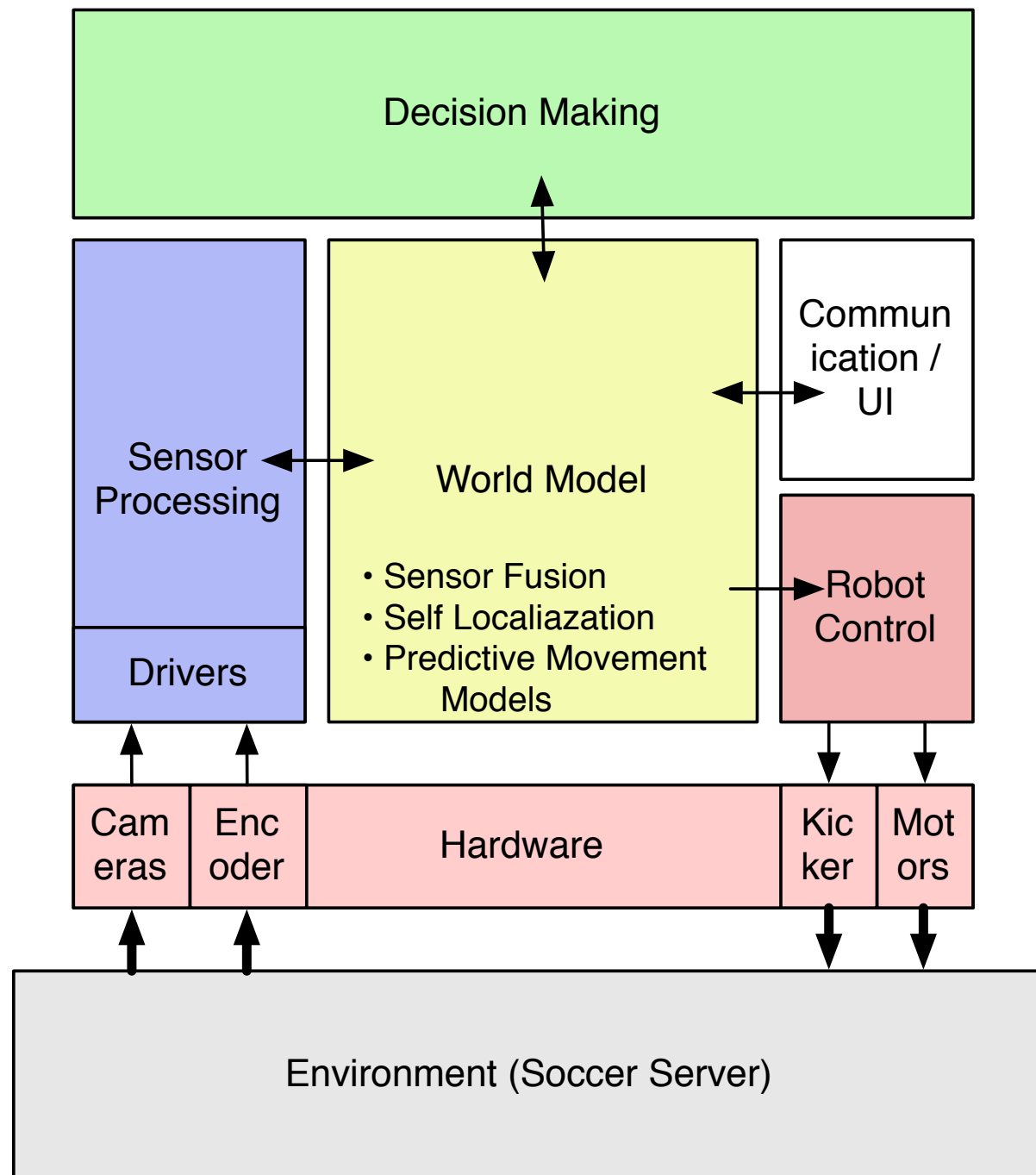
- Research goal: Value-function-based Reinforcement Learning (RL) on real systems
- Markov Decision Process (MDP)
 - discrete time steps
 - state representation with markov property
- As little restrictions as possible on behavior
 - reactive, subsumtion-like behaviors
 - planning



Control Loop

- Control Loop like in Closed Loop Control
 - discrete time steps
 - sequential processing of the modules
 - start of cycle synchronized to camera





- the world model (WM) is the central module of the architecture, the single point of storage for all information
- all modules access the world model to get and store information
- no module has access to another module (besides the WM)
- all modules have abstract interfaces, implementations are exchangable at runtime through abstract factory-pattern

Delays



Camera: ~70 ms



WLan: ~300 ms

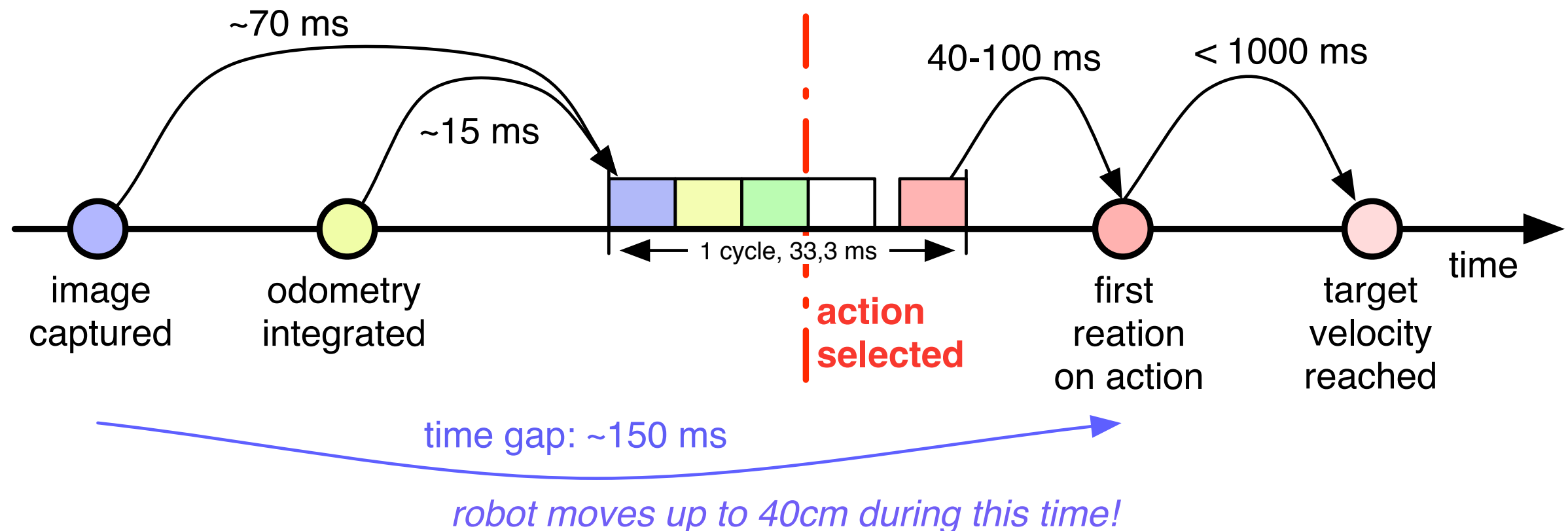


Drive Unit: ~40-100 ms

Encoder / Odometry: ~15 ms

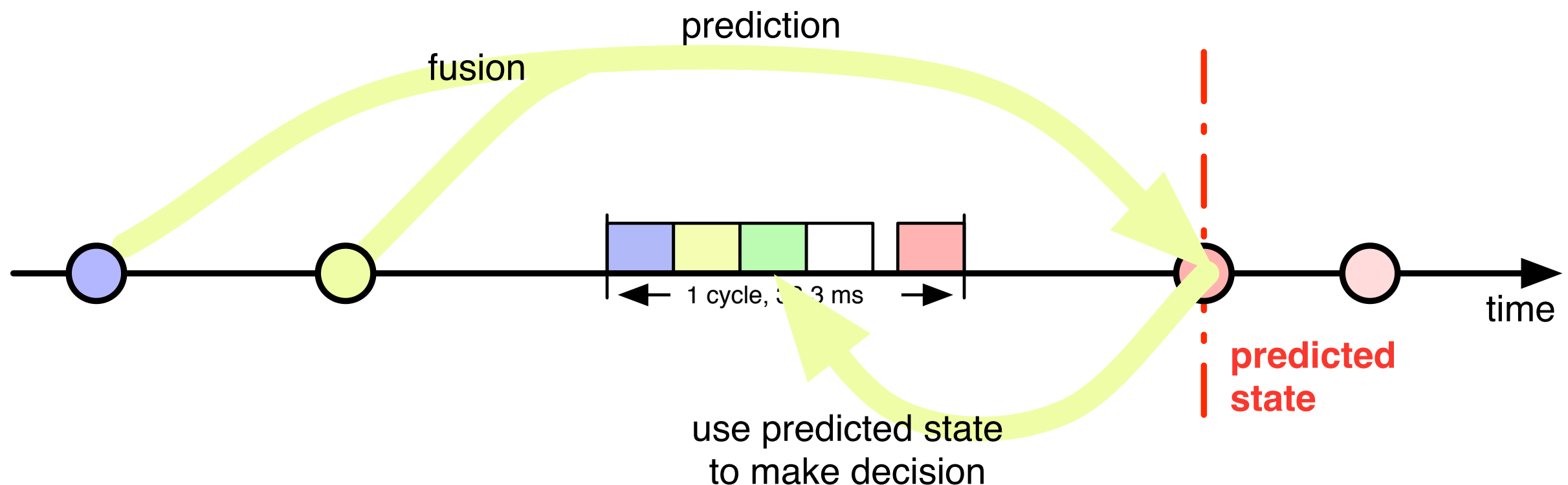
Delays in relation to one cycle

- information is more than 70ms old, when decision is made
- decision won't have any measurable physical effects for up to 100 ms



Prediction

- use predictive models to bridge the time gap
- the extrapolated state resembles markov property as closely as possible
- all decisions are made on this extrapolated state information
- thus, decision making is done „delay free“ (ideally)

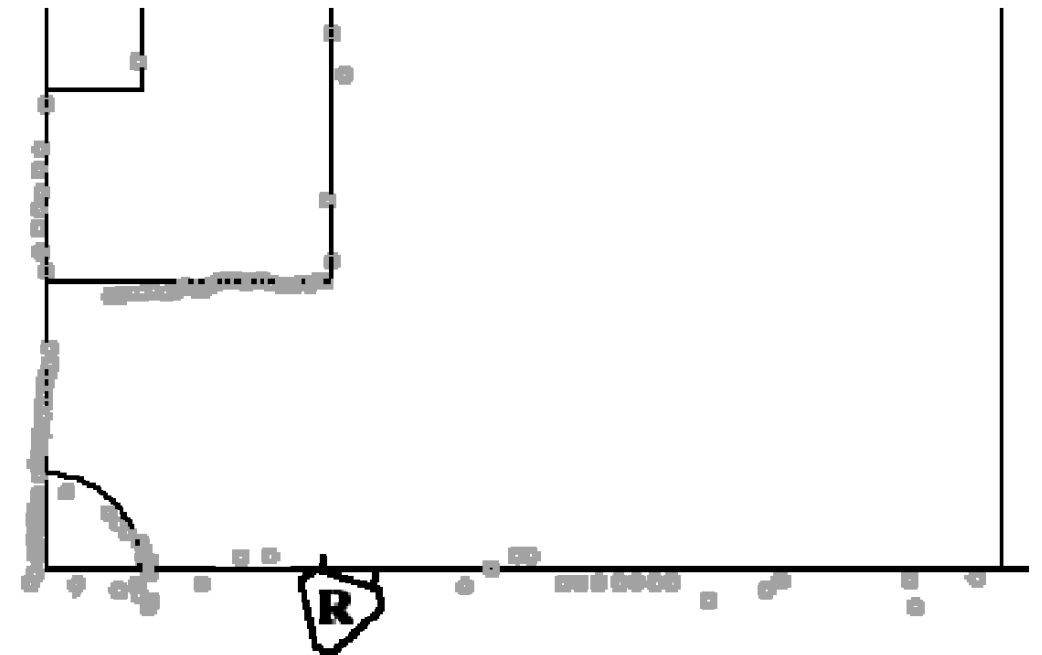
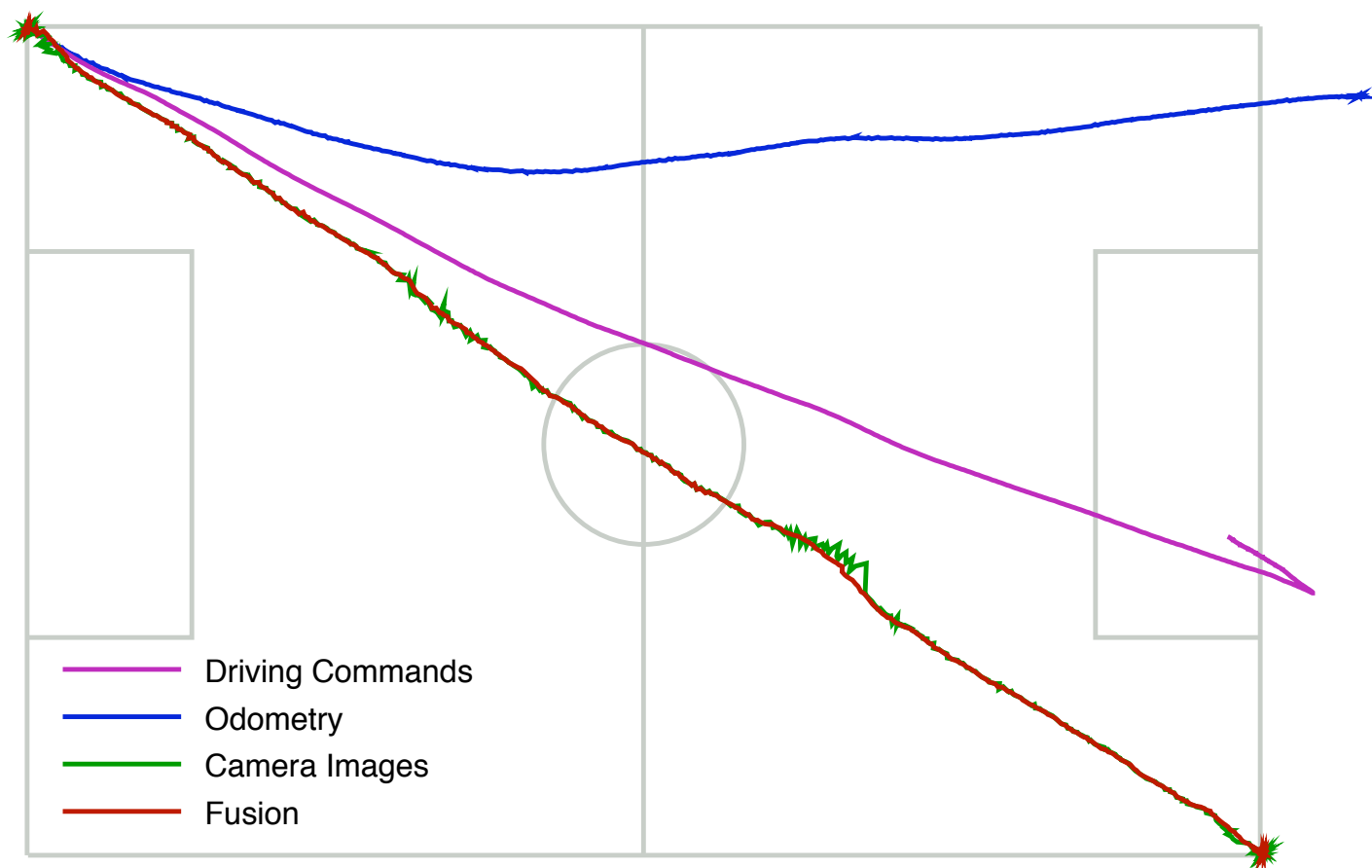


- all prediction is encapsulated in a „predictive world-model“
- all information and all queries are timestamped
- WM is able to „continuously“ extrapolate the state into the future
- Sensor Fusion & Models used in the World-Model
 - Self-localization (minimizing an error using gradient descent)
 - Ball-model (robust regression / multiple hypothesis checking)
 - Self-model (robust regression / MLP)
 - Teammate / Opponent-Model (shared WM, robust regression)

Self-Localization

- Gradient descent with Rprop:

$$\underset{\vec{p}, \phi}{\text{minimize}} \quad E := \sum_{i=1}^n \text{err}(d(\vec{p} + \begin{pmatrix} \cos \phi & -\sin \phi \\ \sin \phi & \cos \phi \end{pmatrix} \vec{s}_i))$$



Ego-Motion

$$\phi_{t+\tau} = \phi_t + \omega_t \cdot \tau \mod 2\pi$$

$$\vec{p}_{t+\tau} = \begin{cases} \vec{p}_t + \vec{u} \cdot \tau & \text{if } \omega_t = 0 \\ \vec{p}_t + \frac{1}{\omega_t} R_{\phi_t} \begin{pmatrix} \sin(\omega_t \tau) & \cos(\omega_t \tau) - 1 \\ 1 - \cos(\omega_t \tau) & \sin(\omega_t \tau) \end{pmatrix} R_{-\phi_t} \vec{u}_t & \text{if } \omega_t \neq 0 \end{cases}$$

$$\omega_{t+\tau} = \omega_t$$

$$\vec{u}_{t+\tau} = R_{\omega_t \tau} \vec{u}_t$$

estimating the velocity from observations $(p_{i,x}, p_{i,y}, \phi_i, t_i)$:

$$\omega = \frac{n \sum_{i=1}^n (\phi_i t_i) - \sum_{i=1}^n t_i \sum_{i=1}^n \phi_i}{n \sum_{i=1}^n t_i^2 + (\sum_{i=1}^n t_i)^2}$$

$$\begin{pmatrix} n & 0 & \sum s_i & \sum c_i \\ 0 & n & -\sum c_i & \sum s_i \\ \sum s_i & -\sum c_i & \sum (s_i^2 + c_i^2) & 0 \\ \sum c_i & \sum s_i & 0 & \sum (s_i^2 + c_i^2) \end{pmatrix} \begin{pmatrix} x_0 \\ y_0 \\ u_x \\ u_y \end{pmatrix} = \begin{pmatrix} \sum p_{i,x} \\ \sum p_{i,y} \\ \sum (s_i p_{i,y} - c_i p_{i,x}) \\ \sum (c_i p_{i,x} + s_i p_{i,y}) \end{pmatrix}$$

$$\text{with } s_i = \frac{\sin(\omega t_i)}{\omega} \text{ and } c_i = \frac{\cos(\omega t_i) - 1}{\omega}$$

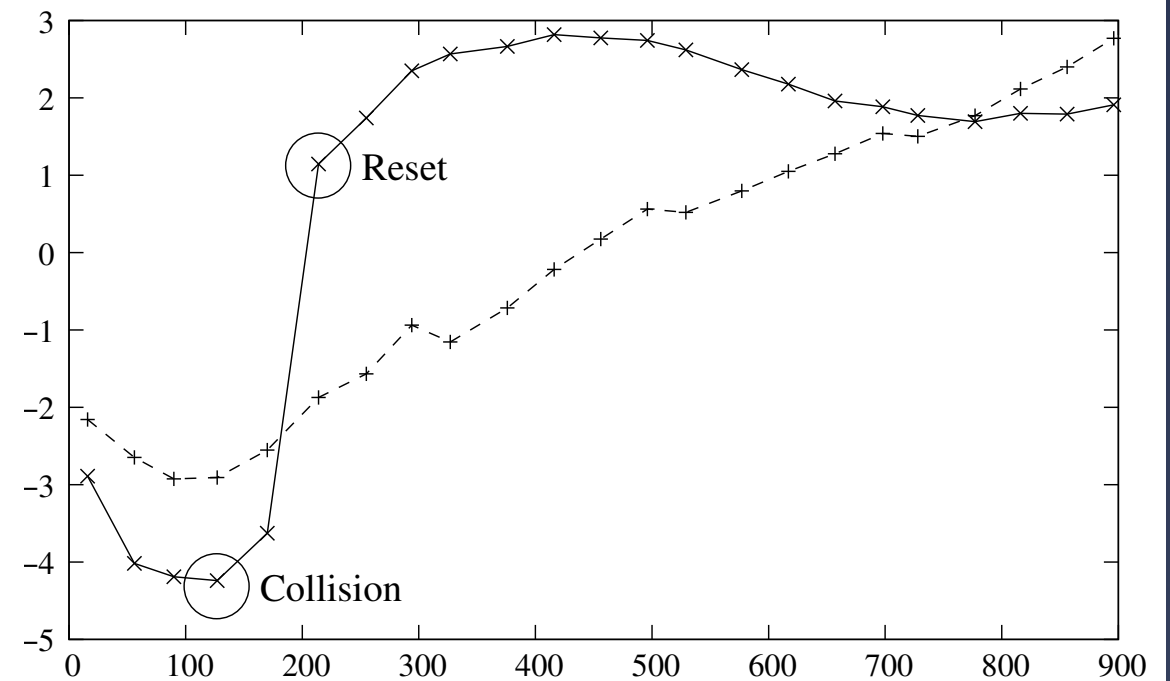
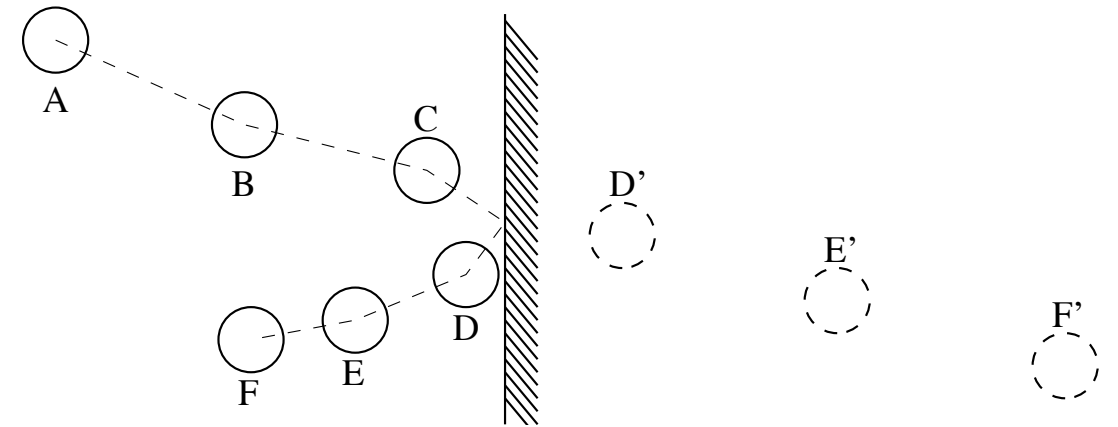
Solving this equation with respect to u_x, u_y yields the velocity estimate.

Alternative: Time series prediction, MLP trained to the transition model.
(similar setup as in „Predicting away the delay“, FU-Fighters)

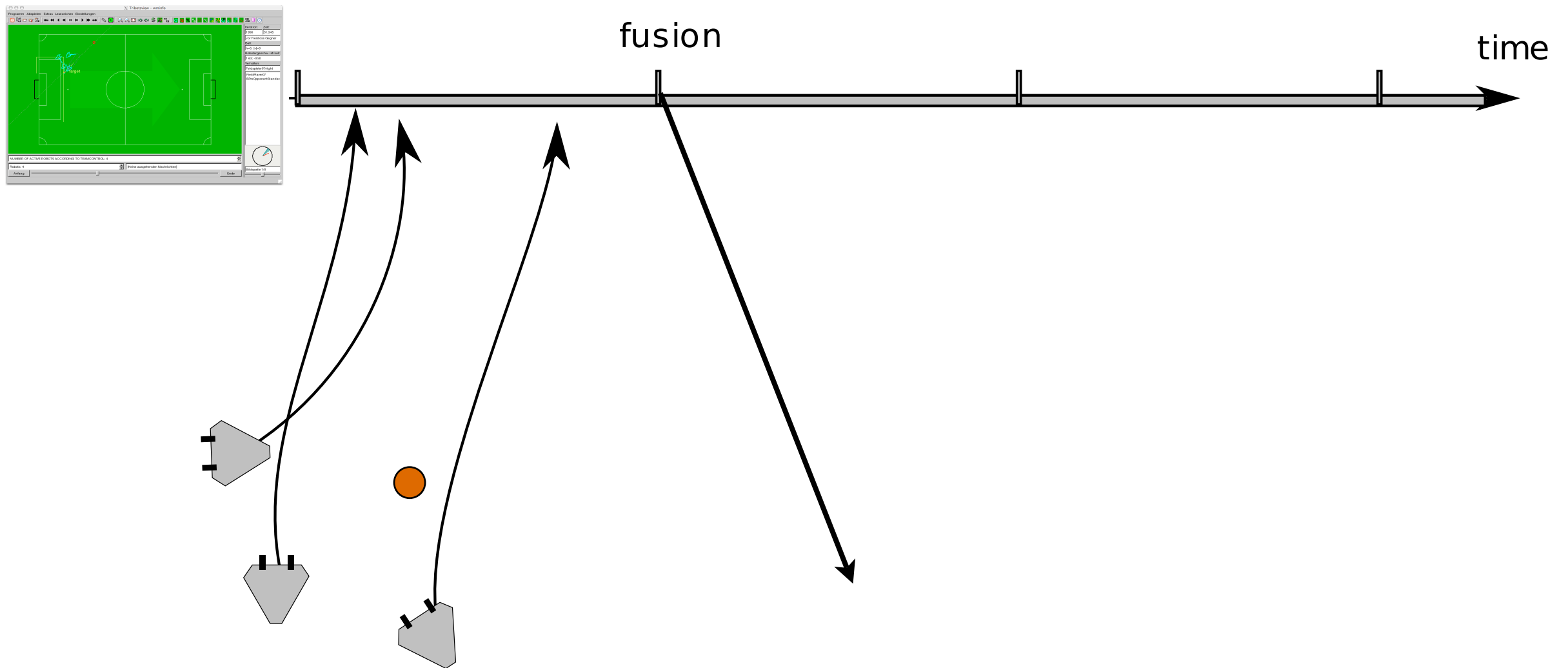
Ball Motion

- sequence of linear motion models
- ridge regression to estimate parameters
- collision detection
- detection of the ball being chipped
- switch to 3D-motion model, if 2nd camera is available

$$\underset{\mathbf{p}_0, \mathbf{v}}{\text{minimize}} \quad \frac{1}{2} \sum_{i=1}^n \|\mathbf{p}_0 + \mathbf{v}(t_i - t_0) - \mathbf{p}_i\|^2 + \frac{\lambda}{2} \|\mathbf{v}\|^2$$



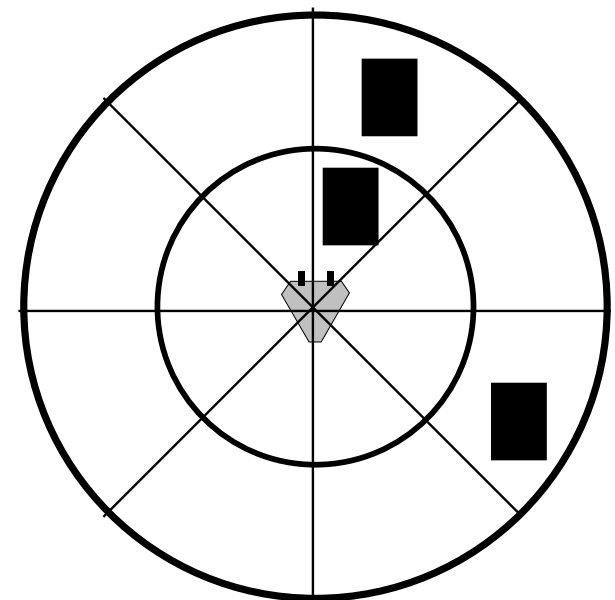
Shared WM



- client-server architecture
- full communication cycle up to 300ms

Shared WM

- information shared with the individual robots
 - position of the ball
 - for each of the teammates
 - position, velocity
 - occupancy grid
 - whiteboard (plain text, unstructured, unrestricted)



- we can not discriminate obstacles visually
- tracking by establishing correspondences between frames minimizing some criterion on the movement of the indistinguishable objects
- same model as for the ego-motion
- communicated positions of teammates are matched to closest obstacles (iff non-ambiguous) and used for identification only

history of object positions

new measurement

standard criterion from
feature point tracking
for matching
indistinguishable objects

$$\delta \mathbf{p}_{t-2,i}, \mathbf{p}_{t-1,j}, \mathbf{p}_{t,k} = \quad (2.22)$$

$$w_1 \left(1 - \frac{(\mathbf{p}_{t-2,i} - \mathbf{p}_{t-1,j}) \cdot (\mathbf{p}_{t-1,j} - \mathbf{p}_{t,k})}{\|\mathbf{p}_{t-2,i} - \mathbf{p}_{t-1,j}\| \cdot \|\mathbf{p}_{t-1,j} - \mathbf{p}_{t,k}\|} \right) + w_2 \left(1 - \frac{2 [\|\mathbf{p}_{t-2,i} - \mathbf{p}_{t-1,j}\| \cdot \|\mathbf{p}_{t-1,j} - \mathbf{p}_{t,k}\|]^{\frac{1}{2}}}{\|\mathbf{p}_{t-2,i} - \mathbf{p}_{t-1,j}\| \cdot \|\mathbf{p}_{t-1,j} - \mathbf{p}_{t,k}\|} \right)$$

Tools

- Tribotsview
- 2 Simulators (team behavior / physically more adequate simulation of drive unit)
- Teamcontrol
- ColorTool / Camera Calibration
- Tacticsboard - modifiable tactics
- iPhone Control
- If you're interested in one tool, please ask...

Conclusion

- Sequential processing in control loop
- All information stored in World-Model
- No direct communication between other modules besides World-Model
- Prediction to „remove“ delay
- Markov Decision Process

