

Event-By-Event Prediction, Gating, and Cancellation notes for me to refer

Goal and setting. Given a real event $e_i = (\mathbf{x}_i, t_i, p_i)$ and a local rotational ego-motion model around \mathbf{c} with angular velocity ω , we predict the event Δt seconds into the future and, when the clock reaches $t_i + \Delta t$, *match* the real and predicted events within spatio-temporal tolerances. If matched (plus a polarity rule), we treat the pair as ego-motion and cancel both; if not, we keep the event as likely scene motion. This mechanism follows the event-warping/contrast-maximization view of motion compensation [??] and the event sensor timing model [??].

1 Rigid rotation & forward prediction

For a 2-D point \mathbf{x} rotating about \mathbf{c} , the rigid rotation is

$$\mathcal{R}(\mathbf{x}; \mathbf{c}, \theta) = \mathbf{c} + R(\theta)(\mathbf{x} - \mathbf{c}), \quad R(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}. \quad (1)$$

This is the same transform used when *warping* events to a reference time in contrast-maximization methods [??]. Setting $\theta = \omega \Delta t$ gives the forward prediction

$$\mathbf{x}'_i = \mathcal{R}(\mathbf{x}_i; \mathbf{c}, \omega \Delta t), \quad t'_i = t_i + \Delta t, \quad \bar{p}_i = -p_i, \quad (2)$$

where the opposite polarity follows the short-horizon sign flip needed to suppress redundant contrast (consistent with event formation and polarity conventions summarized in [??]).

2 Estimating (\mathbf{c}, ω)

Center by algebraic circle LS (Kåsa family). Fit $x^2 + y^2 + ax + by + c = 0$ to points (x_k, y_k) via least squares. With $A = [x \ y \ 1]$ and $b = -(x^2 + y^2)$, solve $A[a, b, c]^T = b$ (in LS sense), then

$$c_x = -\frac{a}{2}, \quad c_y = -\frac{b}{2}, \quad r = \sqrt{c_x^2 + c_y^2 - c}. \quad (3)$$

This standard derivation appears in classic circle-fitting treatments (Kåsa/Delogne families).

Angular velocity from unwrapped phase slope. Let $\theta_k = \text{unwrap}\{\text{atan2}(y_k - c_y, x_k - c_x)\}$. Regress

$$\theta_k \approx \theta_0 + \omega (t_k - t_0) \quad \Rightarrow \quad \hat{\omega} = \text{slope}(\theta \text{ vs. } t), \quad (4)$$

which is the rotation-only specialization consistent with event-based motion estimation frameworks [??].

3 Temporal gate (decision-time test)

Definition. For each real event e_i at t_i , define the *decision time* $t^* = t_i + \Delta t$ and accept only real events whose timestamps lie in a tight band around t^* :

$$|t_j - t^*| \leq \epsilon_t. \quad (5)$$

Derivation and rationale. Event warping/transport [??] treats motion compensation as aligning events to a reference time. Here, the reference for e_i is precisely $t_i + \Delta t$, where the predicted anti-event arrives. Due to microsecond-scale latency/jitter and timestamp noise in DVS-type sensors [??], the true arrival falls in a narrow interval of width $2\epsilon_t$. Choosing $\epsilon_t \gtrsim 3\sigma_t$ (sensor timing stdev) captures the bulk of on-model arrivals while rejecting unrelated events. This implements exactly the intuitive rule: *wait until the clock reaches $t_i + \Delta t$, then only look nearby in time.*

4 Spatial gate (error-budget proof)

Definition. Among temporally valid candidates, accept matches whose positions are close to the forward prediction:

$$\|\mathbf{x}_j - \mathbf{x}'_i\|_2 \leq \epsilon_{xy}. \quad (6)$$

Error budget \Rightarrow sufficiency condition. Let $r = \|\mathbf{x}_i - \mathbf{c}\|_2$. Three first-order contributors dominate the spatial prediction error (consistent with the kinematics used by [5] and the timing model of [??]):

- (i) **Angular-velocity bias** $\Delta\omega = \hat{\omega} - \omega^*$ over horizon Δt causes an angle error $\delta\theta = \Delta\omega \Delta t$, producing chordal error

$$\varepsilon_\omega(r, \Delta t) = 2r |\sin(\delta\theta/2)| \approx r |\Delta\omega| \Delta t \quad (\text{small-angle}). \quad (7)$$

- (ii) **Center bias** $\Delta\mathbf{c}$ perturbs the rotation center, giving a first-order bound

$$\varepsilon_c(r) \lesssim \|\Delta\mathbf{c}\|_2. \quad (8)$$

- (iii) **Timing uncertainty** σ_t yields phase std. $\sigma_\theta \approx |\omega^*| \sigma_t$, hence spatial std. $\sigma_x \approx r |\omega^*| \sigma_t$.

$$\sigma_x \approx r |\omega^*| \sigma_t. \quad (9)$$

A conservative acceptance radius is therefore

$$\varepsilon_\omega(r, \Delta t) + \varepsilon_c(r) + \sigma_x \leq \epsilon_{xy}, \quad (10)$$

which directly justifies the fixed spatial gate in (6). Intuitively: *only keep matches whose spatial discrepancy is explainable by model rate/center error and sensor timing noise.*

5 Matching & cancellation policy

Polarity rule. Use opposite polarity by default ($p_j = -p_i$), consistent with short-horizon compensation and polarity conventions [??]. (Other modes: equal/ignore for sensitivity analyses.)

One-to-one matching. Within the gated set, pick the nearest neighbor in ℓ_2 and mark both predicted & real events as matched (dedup). This mirrors the greedy association used in practical warping-based pipelines ([?]; see also predictive suppression rationale below).

Cancellation. If the temporal, spatial, and polarity gates pass, treat the pair as ego-motion and remove both from residuals. Otherwise, retain the real event—likely scene motion (e.g., a hand) that violates the circular model.

Predictive rationale. The eliminate-on-match behavior is the event-domain analogue of predictive coding: the model emits an *anti-event* that cancels expected input; only *surprise* remains [??].

References

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How each source verifies the math

- **Warping/prediction** (section 1): identical rigid warp as in [[5]] and [[2]]; (1)–(2) are rotation-only versions of event warping.
- **Temporal gate** (section 3): decision-time test matches alignment to reference time; ϵ_t grounded in sensor latency/jitter models [??].
- **Spatial gate** (section 4): error budget (7)–(9) combines small-angle kinematics and timing noise [??]; sufficiency (10) mirrors robust bands in [??].
- **Angular velocity** (section 2): phase-slope (4) agrees with rotational estimation in [??].
- **Polarity/cancellation** (section 5): predictive suppression aligns with biological predictive coding [??].