

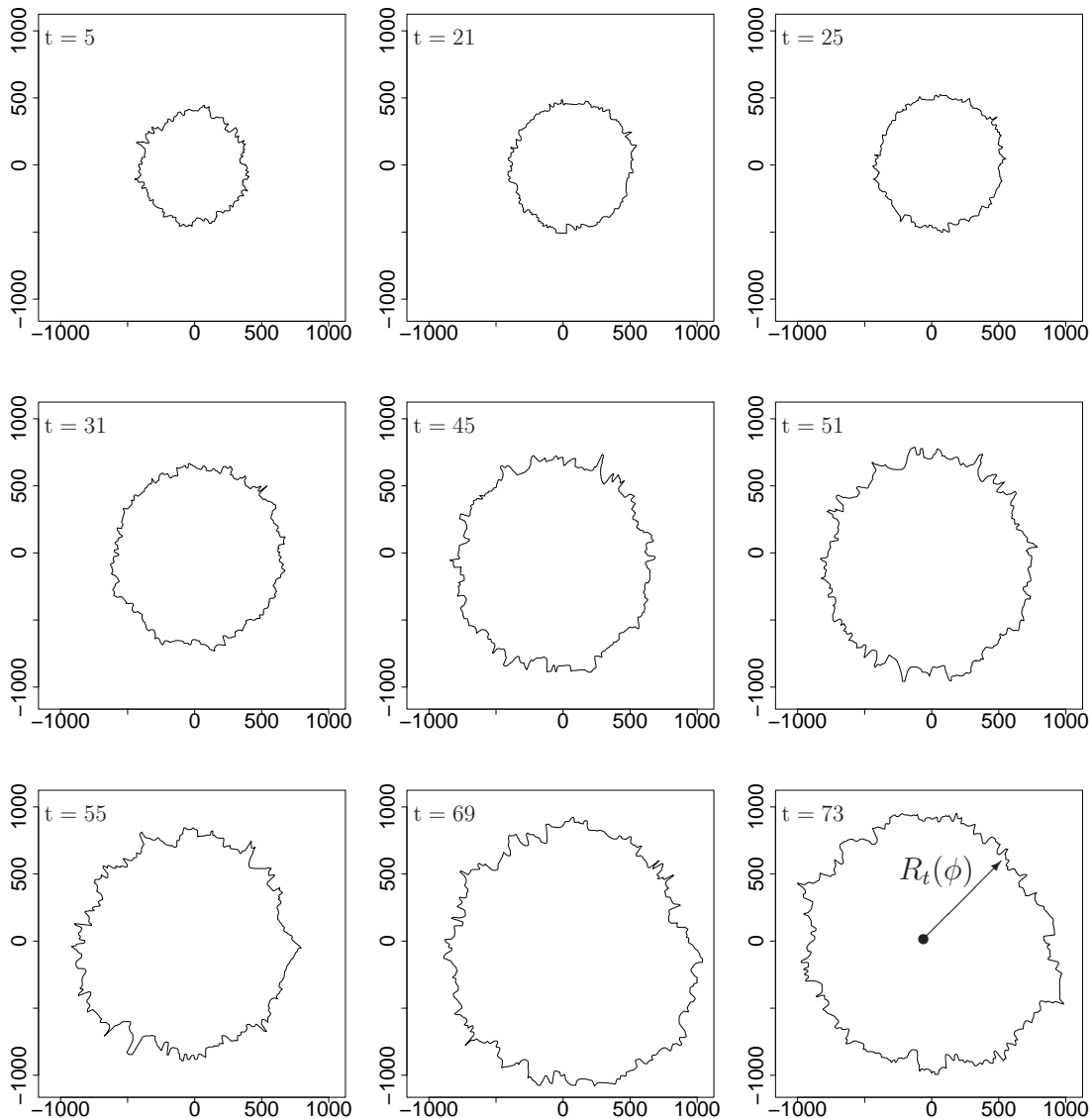
Self-scaling tumor growth

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star-shaped tumor profiles:

$$R_t(\phi) = \max\{R : c_0 + Re_\phi \in Y_t\}$$



spatial correlators:

$$c_{n_1, n_2}(t, \Delta\phi) \equiv \frac{\mathbf{E}\{r_t(\phi)^{n_1} r_t(\phi + \Delta\phi)^{n_2}\}}{\mathbf{E}\{r_t(\phi)^{n_1}\} \mathbf{E}\{r_t(\phi + \Delta\phi)^{n_2}\}}$$

where $\mathbf{E}\{ \}$ denotes the expectation and

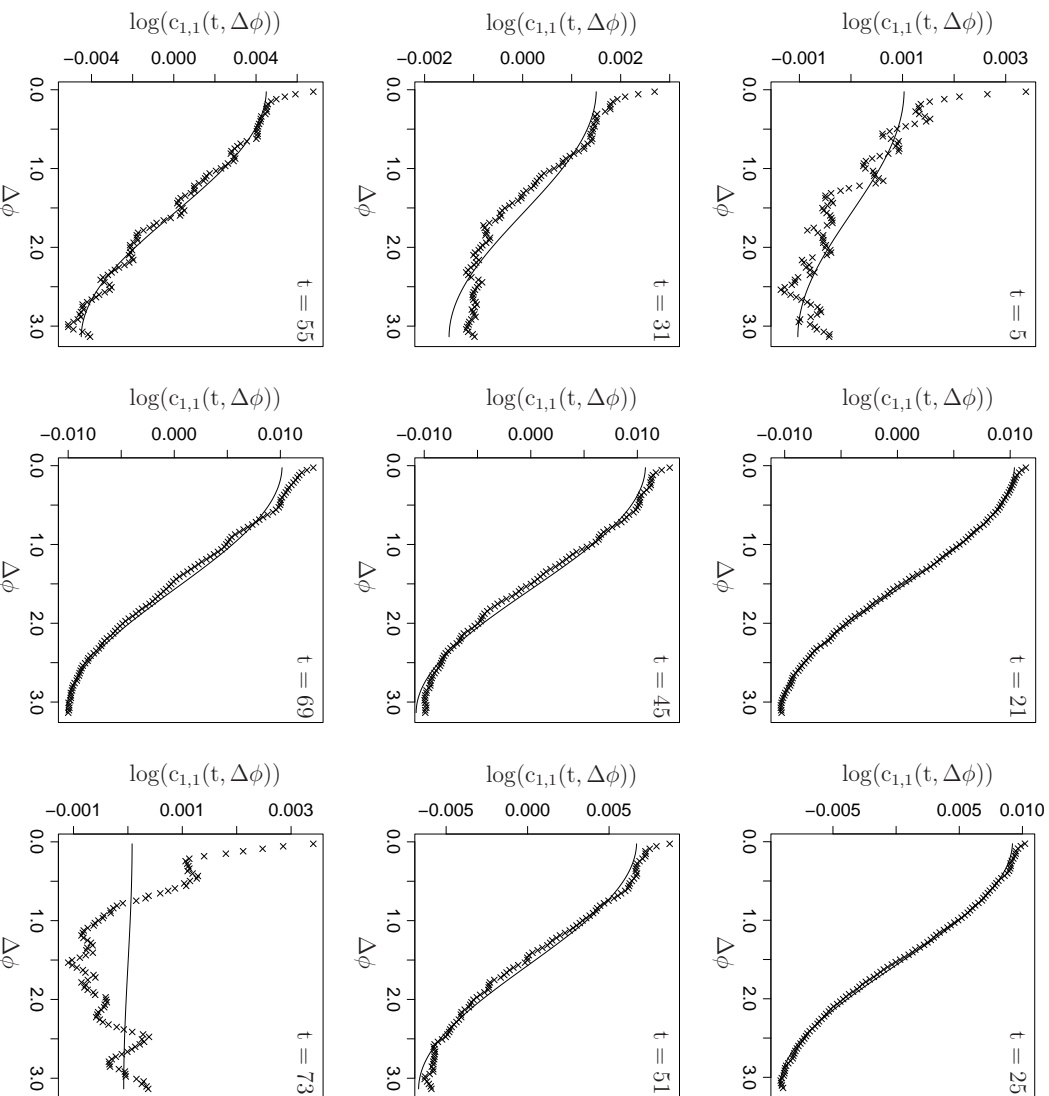
$$r_t(\phi) \equiv \frac{R_t(\phi)}{\mathbf{E}\{R_t(\phi)\}}$$

is the normalized radius function.

spatial correlators: cosine law

$$\log(c_{n_1, n_2}(t, \Delta\phi)) = b_{n_1, n_2}(t) \cos(\Delta\phi)$$

for $\Delta\phi > \phi_0(t)$ (critical angle)

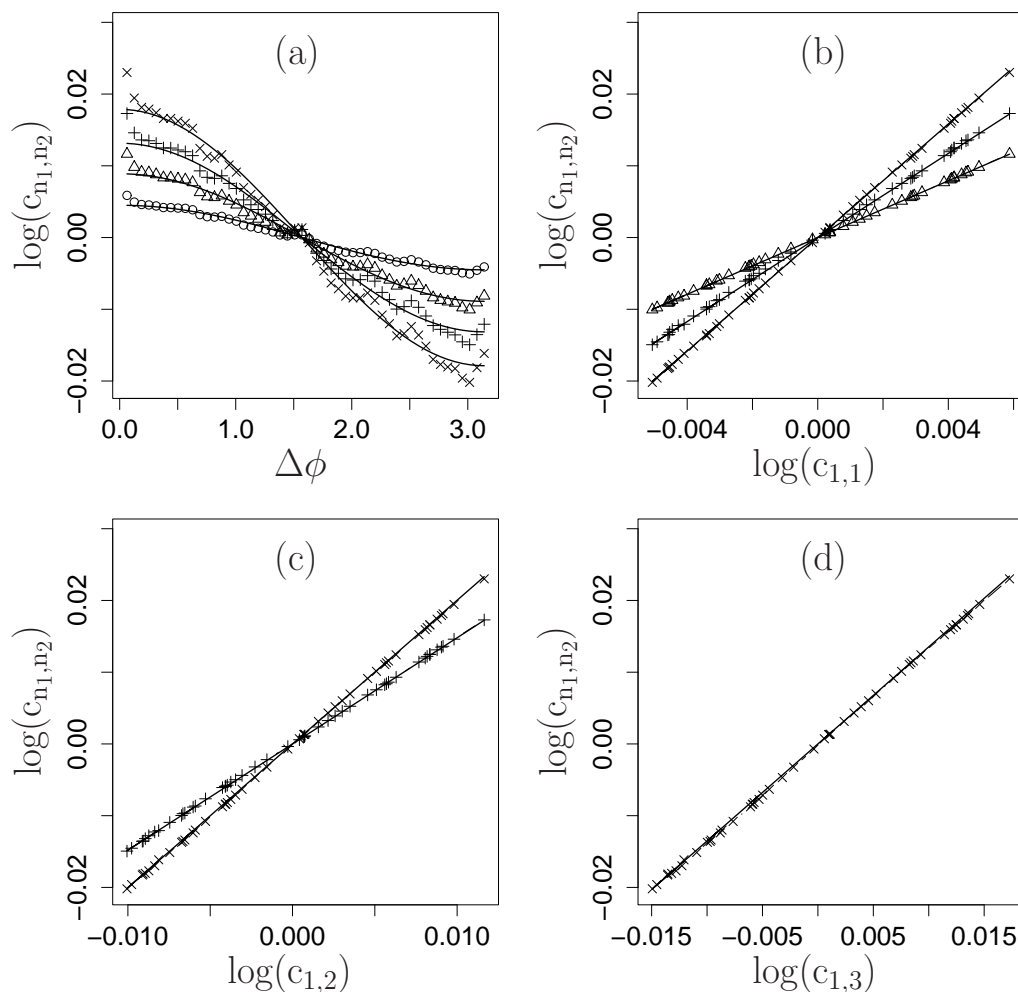


spatial correlators: self-scaling

$$c_{n_1, n_2}(t, \Delta\phi) = (c_{m_1, m_2}(t, \Delta\phi))^{k_t[m_1, m_2; n_1, n_2]}$$

The self-scaling exponents $k_t[m_1, m_2; n_1, n_2]$ are independent of $\Delta\phi$ and given by

$$k_t[m_1, m_2; n_1, n_2] = \frac{b_{n_1, n_2}(t)}{b_{m_1, m_2}(t)}$$



spatial correlators: self-scaling

Self-scaling holds for all angular distances $\Delta\phi$, including the deviations from the cosine law below the critical angle $\phi_0(t)$. Therefore, we may expect the deviations at small angular distances to be of the form

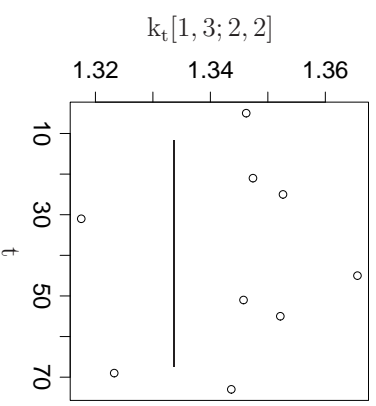
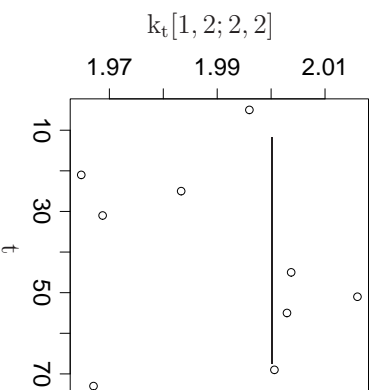
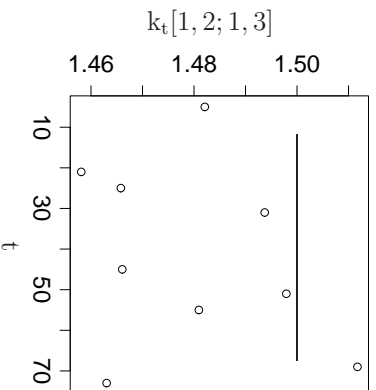
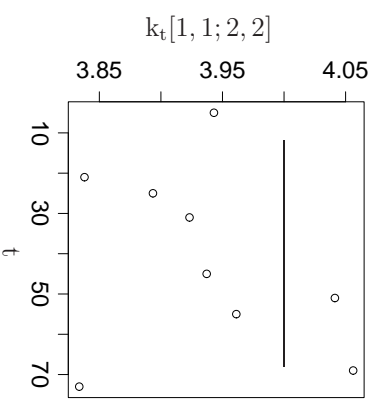
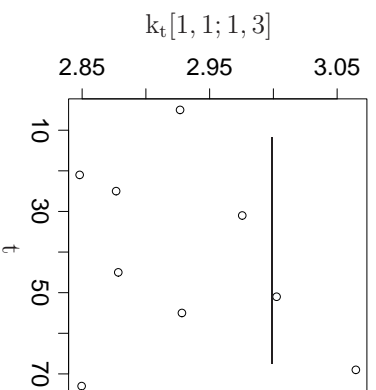
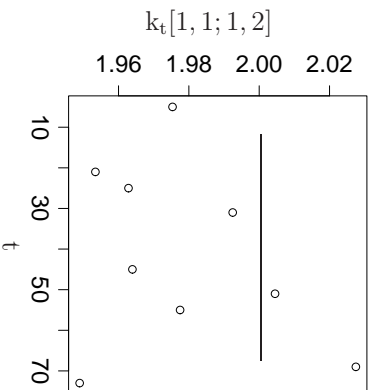
$$\begin{aligned}\log(c_{n_1, n_2}(t, \Delta\phi)) - b_{n_1, n_2}(t) \cos(\Delta\phi) \\ = d_{n_1, n_2}(t) f_t(\Delta\phi)\end{aligned}$$

for $\Delta\phi < \phi_0(t)$ and

$$\frac{d_{n_1, n_2}(t)}{d_{m_1, m_2}(t)} = \frac{b_{n_1, n_2}(t)}{b_{m_1, m_2}(t)} = k_t[m_1, m_2; n_1, n_2].$$

self-scaling exponents:

$$k_t[m_1, m_2; n_1, n_2] = \frac{n_1 n_2}{m_1 m_2}$$



modelling framework: Lévy bases

Let Z be a factorisable and homogeneous Lévy basis on $\mathbb{R} \times \mathbb{R}$ such that $Z(A)$ is infinitely divisible for any $A \subset \mathbb{R} \times \mathbb{R}$. Then we have the fundamental relation

$$\mathbf{E} \left\{ \exp \left\{ \int_A h(a) Z(da) \right\} \right\} = \exp \left\{ \int_A \mathbf{K}[h(a)] da \right\}$$

where h is any integrable deterministic function, and \mathbf{K} denotes the cumulant function of $Z(da)$, defined by

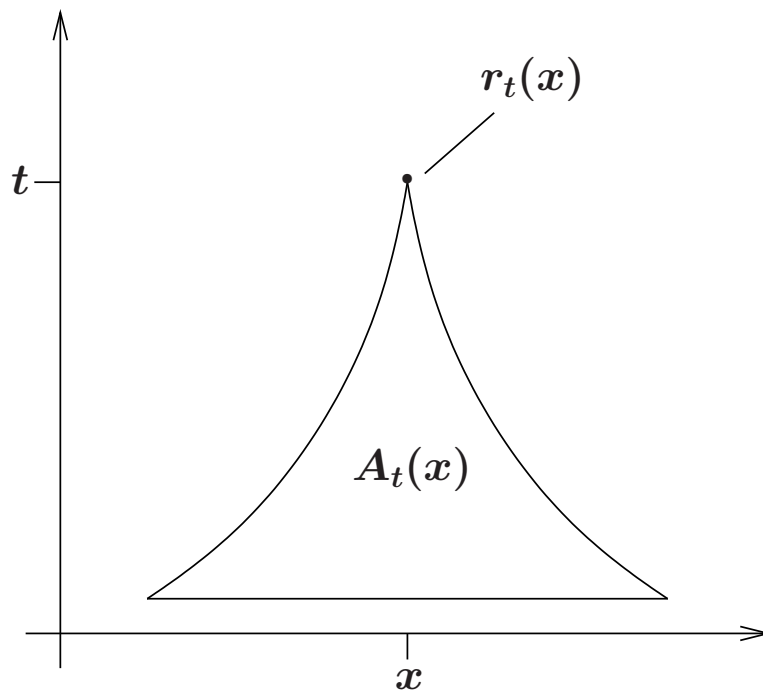
$$\mathbf{K}[\xi] da = \ln \mathbf{E} \{ \exp \{ \xi Z(da) \} \} .$$

exponential ambit process:

Let

$$r_t(x) = \exp \left\{ \int_{A_t(x)} h(t - s, x - \rho) Z(ds \times d\rho) \right\}$$

where $A_t(x) \in \mathbb{R}^2$ is called the associated *ambit set*.



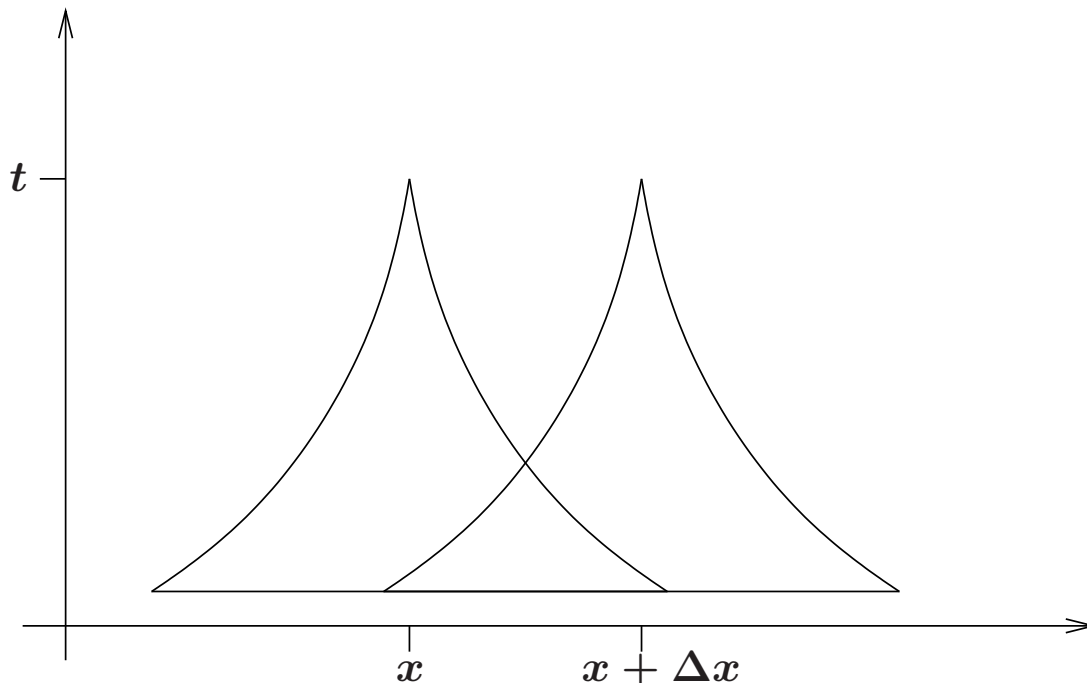
ambit process: correlator

$$c_{n_1, n_2}(t, \Delta x) = \exp \left\{ \int_{A(\Delta x)} \bar{K}_{n_1, n_2}[\Delta x, \rho, s] ds d\rho \right\}$$

where

$$A(\Delta x) = A_t(x) \cap A_t(x + \Delta x)$$

$$\begin{aligned} \bar{K}_{n_1, n_2}[\Delta x, \rho, s] = \\ \mathbf{K}[n_1 h(t-s, x-\rho) + n_2 h(t-s, x+\Delta x-\rho)] - \\ \mathbf{K}[n_1 h(t-s, x-\rho)] - \mathbf{K}[n_2 h(t-s, x+\Delta x-\rho)] \end{aligned}$$



simple example: $h \equiv 1$

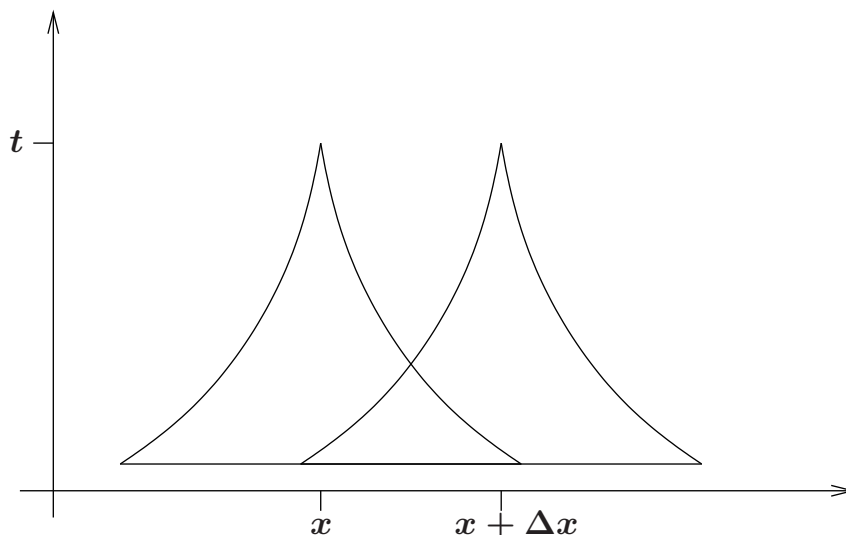
Let $A_t(x) = (x, t) + A_0(0)$ and

$$r_t(x) = \exp \left\{ \int_{A_t(x)} Z(ds \times d\rho) \right\}.$$

Then

$$c_{n_1, n_2}(t, \Delta x) = \exp \left\{ \bar{K}_{n_1, n_2} V(\Delta x) \right\}$$

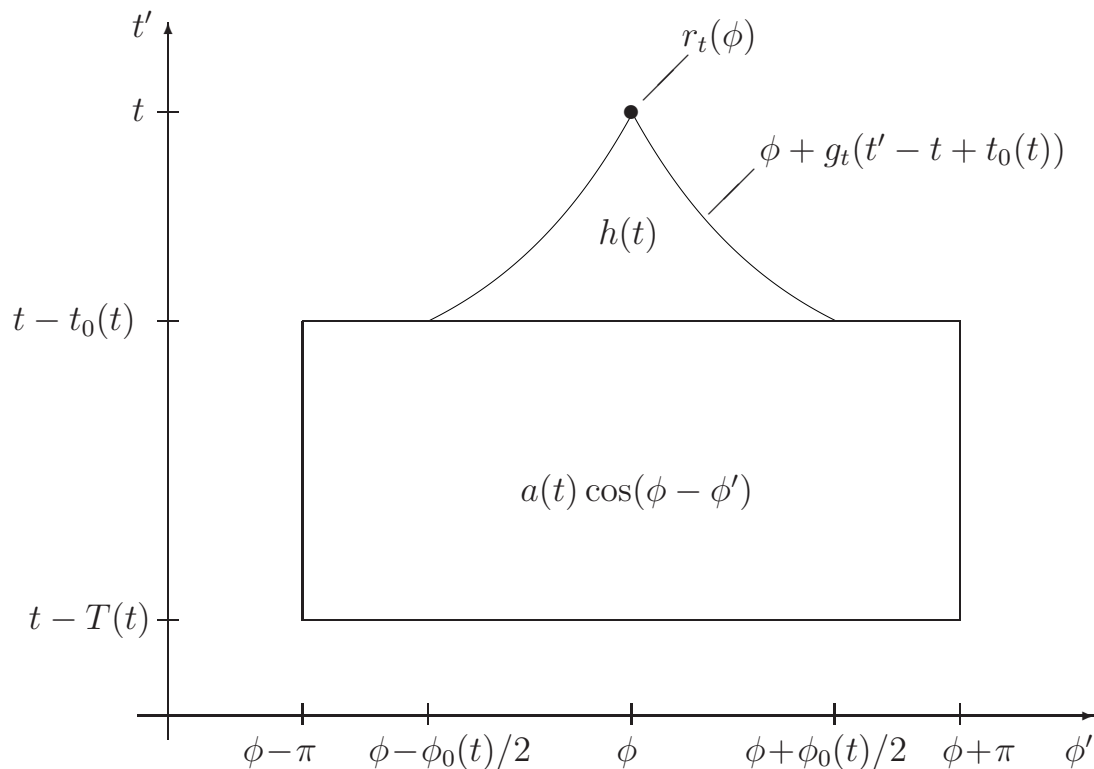
where $V(\Delta x) = \text{Vol}(A_t(x) \cap A_t(x + \Delta x))$
 and $\bar{K}_{n_1, n_2} = K[n_1 + n_2] - K[n_1] - K[n_2]$.



tumor growth: log normal model

$$r_t(\phi) = \exp \left\{ a(t) \int_{t-T(t)}^{t-t_0(t)} \int_{\phi-\pi}^{\phi+\pi} \cos(\phi - \phi') Z(dt' \times d\phi') \right. \\
 \left. + h(t) \int_{t-t_0(t)}^t \int_{\phi-g_t(t'-t+t_0(t))}^{\phi+g_t(t'-t+t_0(t))} Z(dt' \times d\phi') \right\}$$

where Z is a normal Lévy basis with mean $\mu dt d\phi$ and variance $\sigma^2 dt d\phi$.



log normal model: correlators

Spatial correlators follow the observed cosine law with deviations at small scales

$$\log(c_{n_1, n_2}(t, \Delta\phi)) = d_{n_1, n_2}(t) f_t(\Delta\phi) \mathbf{1}_{[0, \phi_0(t)]}(\Delta\phi) + b_{n_1, n_2}(t) \cos(\Delta\phi),$$

where

$$b_{n_1, n_2}(t) = n_1 n_2 a(t)^2 \pi (T(t) - t_0(t))$$

$$d_{n_1, n_2}(t) = n_1 n_2 h(t)^2$$

$$f_t(\Delta\phi) = \int_0^{g_t^{(-1)}(\Delta\phi/2)} (2g_t(s) - \Delta\phi) ds.$$

The critical angle is given by

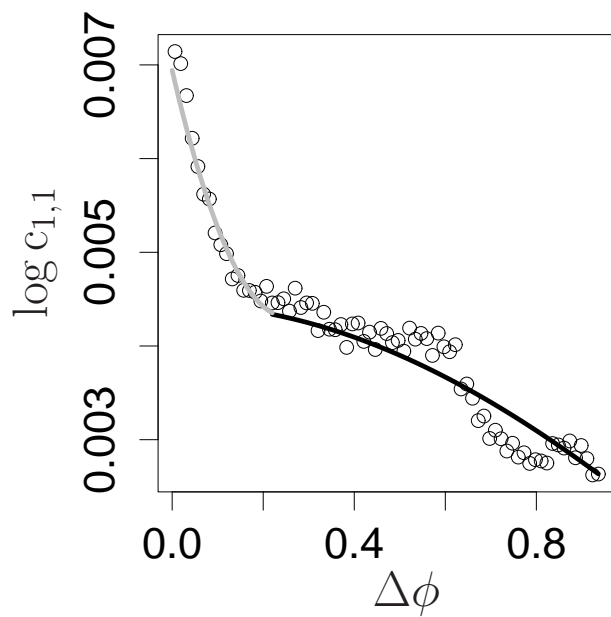
$$\phi_0(t) = 2g_t(0)$$

and

$$k_t[m_1, m_2; n_1, n_2] = \frac{n_1 n_2}{m_1 m_2}.$$

log normal model: simulation

$$g_t(s) = \frac{\phi_0(t)}{2} - \frac{\phi_0(t)}{2t_0(t)}s, \quad s \in [0, t_0(t)].$$



log normal model: simulation

