

# **The Lagrangian PT and the integrated PT**

**Takahiko Matsubara**  
(Nagoya U.)

**@Paris**  
**2016/5/25**

# **The Lagrangian Perturbation Theory (LPT)**

# Eulerian vs Lagrangian picture

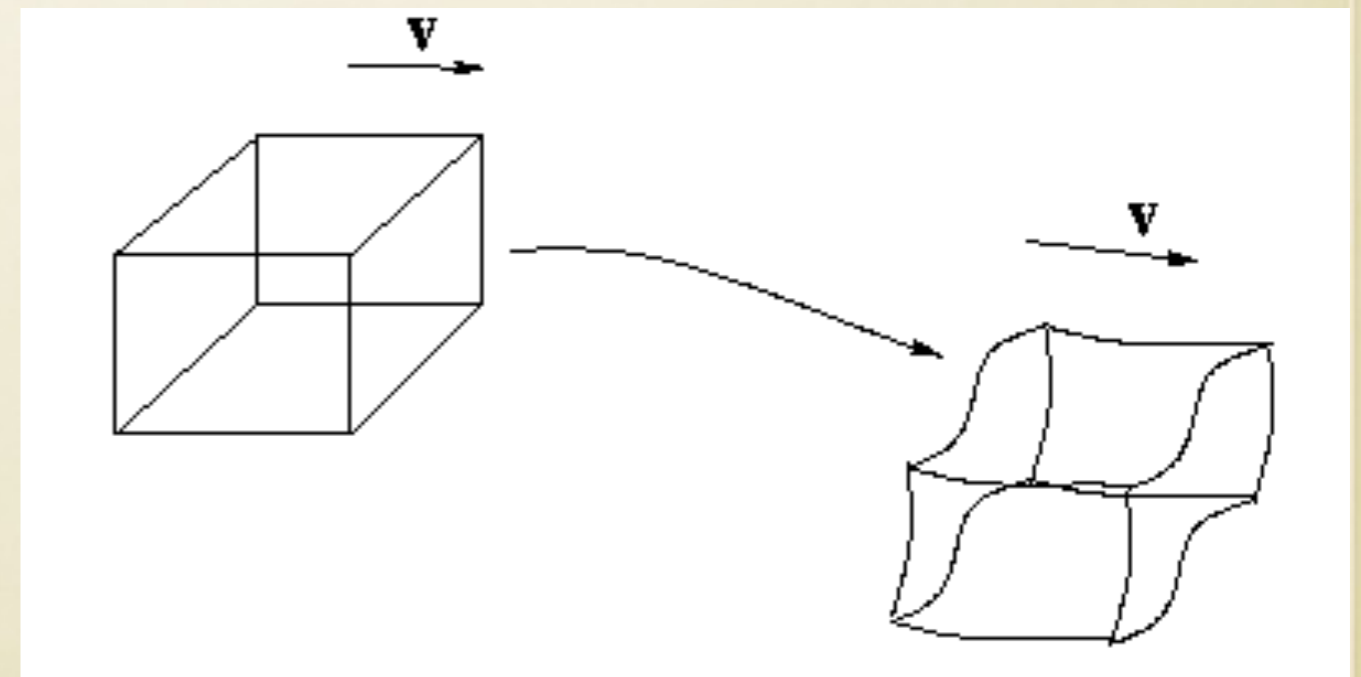
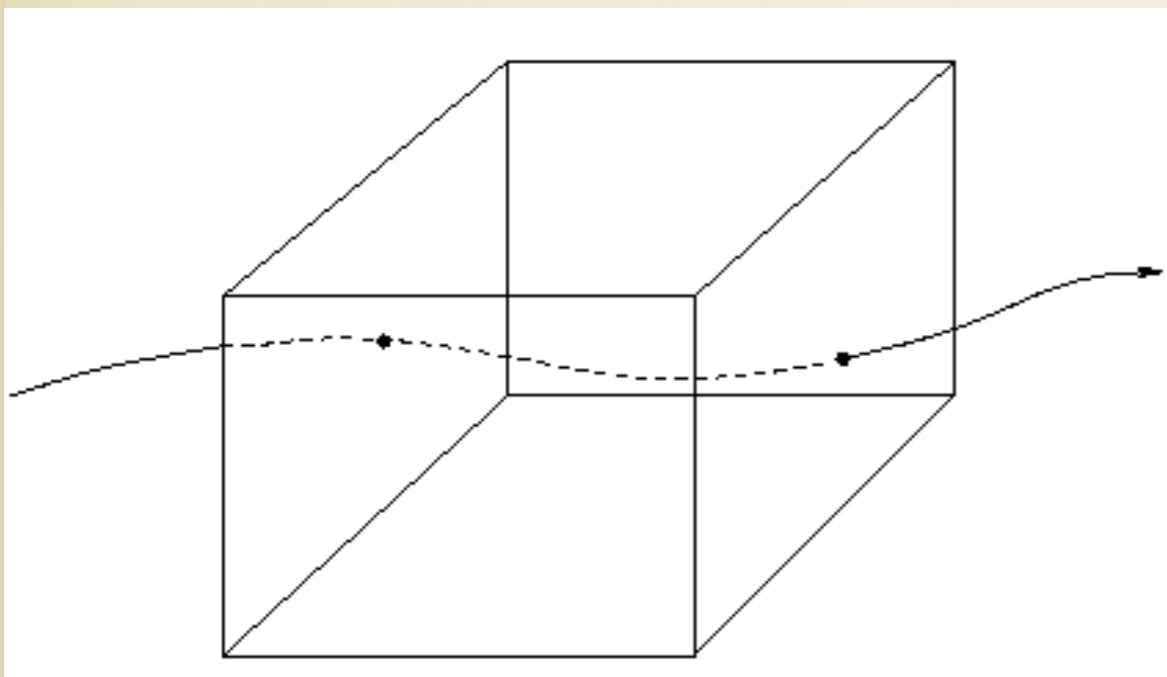
- Eulerian

- density and velocity fields on a fixed space



- Lagrangian

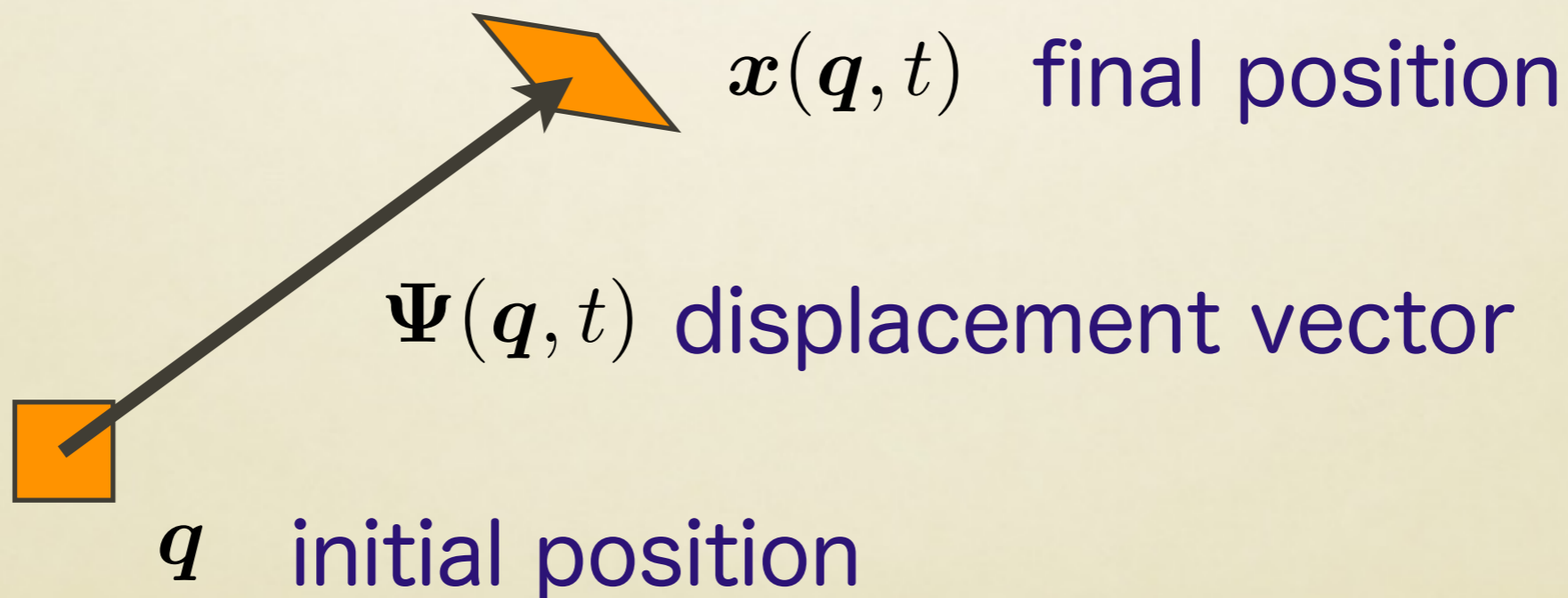
- follows a trajectory of a fluid element



# Variables in the Lagrangian picture

- Fundamental variable in the Lagrangian picture:
  - Displacement of a fluid element from the initial position: Displacement vector field

$$\Psi(\mathbf{q}, t) = \mathbf{x}(\mathbf{q}, t) - \mathbf{q}$$



# Eulerian vs Lagrangian picture

- **Eulerian**

- field labels:  $\boldsymbol{x}$
- fundamental variables:

$$\rho(\boldsymbol{x}, t) = \bar{\rho}[1 + \delta(\boldsymbol{x}, t)]$$

$$\boldsymbol{v}(\boldsymbol{x}, t)$$

- **gravitational potential**

$$\phi(\boldsymbol{x}, t) = 4\pi G \bar{\rho} a^2 \Delta^{-1} \delta(\boldsymbol{x}, t)$$

- **Lagrangian**

- field labels:  $\boldsymbol{q}$
- fundamental variables:

$$\Psi(\boldsymbol{q}, t) = \boldsymbol{x}(\boldsymbol{q}, t) - \boldsymbol{q}$$

- **density**

$$\begin{aligned} \rho(\boldsymbol{q}, t) &= \bar{\rho} \left[ \det \left( \frac{\partial \boldsymbol{x}}{\partial \boldsymbol{q}} \right) \right]^{-1} \\ &= \bar{\rho} \left[ \det \left( \boldsymbol{I} + \frac{\partial \Psi}{\partial \boldsymbol{q}} \right) \right]^{-1} \end{aligned}$$

- **velocity**

$$\boldsymbol{v}(\boldsymbol{q}, t) = a \dot{\boldsymbol{x}} = a \dot{\Psi}(\boldsymbol{q}, t)$$

# Eulerian vs Lagrangian picture

- **Eulerian**

- **EoM**

$$\dot{\delta} + \frac{1}{a} \nabla \cdot [(1 + \delta)\mathbf{v}] = 0$$
$$\dot{\mathbf{v}} + \frac{\dot{a}}{a} \mathbf{v} + \frac{1}{a} (\mathbf{v} \cdot \nabla) \mathbf{v} = -\frac{1}{a} \nabla \Phi$$

- **Lagrangian**

- **EoM**

$$\ddot{\Psi} + 2\frac{\dot{a}}{a} \dot{\Psi} = -\frac{1}{a^2} \nabla_x \Phi$$

- **Common:**

- **Poisson equation**

$$\Delta_x \Phi = 4\pi G \bar{\rho} a^2 \delta(\mathbf{x}, t)$$

# Zel'dovich approximation

- Linearize the Lagrangian EoM w.r.t. displacement

Zel'dovich (1970)



Equations of motion:

$$\ddot{\Psi} + 2\frac{\dot{a}}{a}\dot{\Psi} = -\frac{1}{a^2}\nabla_x\Phi$$

$$\Delta_x\Phi = 4\pi G\bar{\rho}a^2\delta(\mathbf{x}, t)$$



$$\nabla_q \cdot \left( \ddot{\Psi} + 2\frac{\dot{a}}{a}\dot{\Psi} \right) = -4\pi G\bar{\rho}\nabla_q \cdot \Psi$$

$$\nabla_q \times \left( \ddot{\Psi} + 2\frac{\dot{a}}{a}\dot{\Psi} \right) = 0$$



(Taking a growing mode)

$$\nabla_q \times \Psi \simeq 0, \quad \nabla_q \cdot \Psi \propto D(t)$$



Linearization:

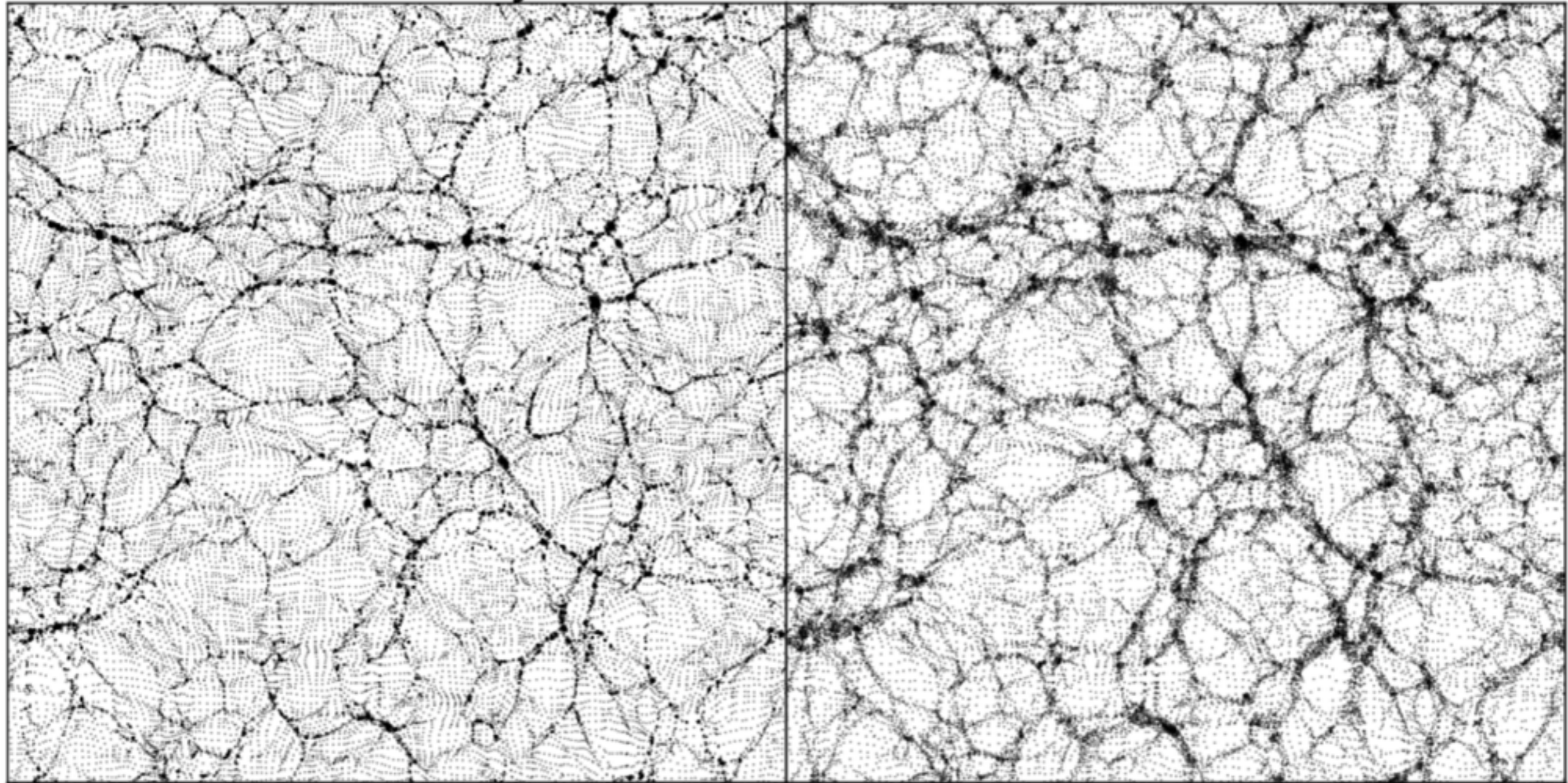
$$\delta(\mathbf{x}, t) = \left[ \det \left( \mathbf{I} + \frac{\partial \Psi}{\partial \mathbf{q}} \right) \right]^{-1} - 1 \simeq -\nabla_q \cdot \Psi$$

$$\nabla_x \cdot \Psi \simeq \nabla_q \cdot \Psi, \quad \nabla_x \times \Psi \simeq \nabla_q \times \Psi$$

$$\Psi \simeq -D(t)\nabla_q\varphi_0(\mathbf{q})$$

Full N-body

Zel'dovich




Neyrinck (2013)



# Lagrangian Perturbation Theory

Buchert (1989); Moutarde+ (1991); Buchert (1992); Buchert & Ehlers (1993); Hivon+ (1995); Catelan (1995); Rampf & Wong (2012); Tatekawa (2013); Zheligovski & Frisch (2014); TM (2015); ...

- Taking into account the higher-order perturbations in the displacement

$$\Psi = \sum_{n=1}^{\infty} \Psi^{(n)} = \Psi^{(1)} + \Psi^{(2)} + \Psi^{(3)} + \dots$$


$$\Psi^{(1)} = -D(t) \nabla \varphi_0(\mathbf{q})$$

(First order: Zel'dovich approx.)

$$\Psi^{(2)} = -\frac{1}{2} D_2(t) \nabla \Delta^{-1} \left[ \Psi_{i,i}^{(1)} \Psi_{j,j}^{(1)} - \Psi_{i,j}^{(1)} \Psi_{i,j}^{(1)} \right]$$

$$\Psi^{(3)} = -\frac{1}{3!} \left[ D_{3a}(t) \nabla \Delta^{-1} \left( \Psi_{i,i}^{(1)} \Psi_{j,j}^{(2)} - \Psi_{i,j}^{(1)} \Psi_{i,j}^{(2)} \right) + D_{3b}(t) \nabla \Delta^{-1} \det \left( \Psi_{i,j}^{(1)} \right) \right. \\ \left. + D_{3c}(t) \Delta^{-1} \left( \Psi_{,j}^{(1)} \Psi_{i,j}^{(2)} - \Psi_{i,j}^{(1)} \Psi_{,j}^{(2)} \right)_{,i} \right]$$

⋮

$$\left( D_2 \simeq \frac{3}{7} D^2, D_{3a} \simeq -\frac{10}{7} D^2, D_{3b} \simeq 2D^2, D_{3c} \simeq -\frac{6}{7} D^2, \dots \right)$$

# Recursive solutions (general)

TM (2015)

$$\ddot{\mathbf{x}} + 2H\dot{\mathbf{x}} = -\frac{1}{a^2}\nabla_x\phi(\mathbf{x}, t),$$

$$\Delta_x\phi(\mathbf{x}, t) = 4\pi G\bar{\rho}a^2\delta(\mathbf{x}, t),$$

$$\hat{\mathcal{T}} \equiv \frac{\partial^2}{\partial t^2} + 2H\frac{\partial}{\partial t},$$



$$\mathbf{x}(\mathbf{q}, t) = \mathbf{q} + \boldsymbol{\Psi}(\mathbf{q}, t).$$

$$\nabla \cdot \boldsymbol{\Psi} = D_+(t)A_+ + D_-(t)A_- - (\hat{\mathcal{T}} - 4\pi G\bar{\rho})^{-1} \left[ \varepsilon_{ijk}\varepsilon_{ipq}\Psi_{j,p} (\hat{\mathcal{T}} - 2\pi G\bar{\rho})\Psi_{k,q} + \frac{1}{2}\varepsilon_{ijk}\varepsilon_{pqr}\Psi_{i,p}\Psi_{j,q} \left( \hat{\mathcal{T}} - \frac{4\pi G}{3}\bar{\rho} \right)\Psi_{k,r} \right],$$

$$\nabla \times \boldsymbol{\Psi} = \mathbf{B}_0 + E_-(t)\mathbf{B}_- + \hat{\mathcal{T}}^{-1} (\nabla\Psi_i \times \hat{\mathcal{T}}\nabla\Psi_i),$$

$$\boldsymbol{\Psi} = \sum_{n=1}^{\infty} \boldsymbol{\Psi}^{(n)} = \boldsymbol{\Psi}^{(1)} + \boldsymbol{\Psi}^{(2)} + \boldsymbol{\Psi}^{(3)} + \dots,$$

$$\boldsymbol{\Psi} = \Delta^{-1} [\nabla(\nabla \cdot \boldsymbol{\Psi}) - \nabla \times (\nabla \times \boldsymbol{\Psi})],$$

# Recursive solutions (growing modes)

$$\tilde{\Psi}^{(n)}(\mathbf{k}, t) = \frac{iD^n}{n!} \int_{\mathbf{k}_{1\dots n}=\mathbf{k}} L_n(\mathbf{k}_1, \dots, \mathbf{k}_n) \delta_0(\mathbf{k}_1) \cdots \delta_0(\mathbf{k}_n).$$

$$\mathbf{k}_{1\dots n} \equiv \mathbf{k}_1 + \cdots + \mathbf{k}_n$$

$$L_n(\mathbf{k}_1, \dots, \mathbf{k}_n) = \frac{1}{k_{1\dots n}^2} [k_{1\dots n} S_n(\mathbf{k}_1, \dots, \mathbf{k}_n) + \mathbf{k}_{1\dots n} \times \mathbf{T}_n(\mathbf{k}_1, \dots, \mathbf{k}_n)].$$

$$U(\mathbf{k}_1, \mathbf{k}_2) = \frac{|\mathbf{k}_1 \times \mathbf{k}_2|^2}{k_1^2 k_2^2} = 1 - \left( \frac{\mathbf{k}_1 \cdot \mathbf{k}_2}{k_1 k_2} \right)^2,$$

$$V(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) = \frac{|\mathbf{k}_1 \cdot (\mathbf{k}_2 \times \mathbf{k}_3)|^2}{k_1^2 k_2^2 k_3^2} = 1 - \left( \frac{\mathbf{k}_1 \cdot \mathbf{k}_2}{k_1 k_2} \right)^2 - \left( \frac{\mathbf{k}_2 \cdot \mathbf{k}_3}{k_2 k_3} \right)^2 - \left( \frac{\mathbf{k}_3 \cdot \mathbf{k}_1}{k_3 k_1} \right)^2 + 2 \frac{(\mathbf{k}_1 \cdot \mathbf{k}_2)(\mathbf{k}_2 \cdot \mathbf{k}_3)(\mathbf{k}_3 \cdot \mathbf{k}_1)}{k_1^2 k_2^2 k_3^2},$$

$$W(\mathbf{k}_1, \mathbf{k}_2) = \frac{(\mathbf{k}_1 \times \mathbf{k}_2)(\mathbf{k}_1 \cdot \mathbf{k}_2)}{k_1^2 k_2^2}.$$

$$S_1(\mathbf{k}) = 1, \quad \mathbf{T}_1(\mathbf{k}) = \mathbf{0}.$$

$$S_2(\mathbf{k}_1, \mathbf{k}_2) = \frac{3}{7} U(\mathbf{k}_1, \mathbf{k}_2), \quad \mathbf{T}_2(\mathbf{k}_1, \mathbf{k}_2) = \mathbf{0}.$$

$$S_3(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) = \frac{5}{3} U(\mathbf{k}_1, \mathbf{k}_{23}) S_2(\mathbf{k}_2, \mathbf{k}_3) - \frac{1}{3} V(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3),$$

$$\mathbf{T}_3(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) = W(\mathbf{k}_1, \mathbf{k}_{23}) S_2(\mathbf{k}_2, \mathbf{k}_3).$$

$$\begin{aligned}
S_4(\mathbf{k}_1, \dots, \mathbf{k}_4) &= \frac{28}{11} \left[ U(\mathbf{k}_1, \mathbf{k}_{234}) S_3(\mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) - W(\mathbf{k}_1, \mathbf{k}_{234}) \cdot T_3(\mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) \right] \\
&\quad + \frac{17}{11} U(\mathbf{k}_{12}, \mathbf{k}_{34}) S_2(\mathbf{k}_1, \mathbf{k}_2) S_2(\mathbf{k}_3, \mathbf{k}_4) - \frac{26}{11} V(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_{34}) S_2(\mathbf{k}_3, \mathbf{k}_4), \\
T_4(\mathbf{k}_1, \dots, \mathbf{k}_4) &= 2 \left[ W(\mathbf{k}_1, \mathbf{k}_{234}) S_3(\mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) + \frac{\mathbf{k}_1 \times \mathbf{k}_{234}}{k_1^2 k_{234}^2} (\mathbf{k}_1 \times \mathbf{k}_{234}) \cdot T_3(\mathbf{k}_2, \mathbf{k}_3, \mathbf{k}_4) \right].
\end{aligned}$$

$$\begin{aligned}
S_5(\mathbf{k}_1, \dots, \mathbf{k}_5) &= \frac{45}{13} \left[ U(\mathbf{k}_1, \mathbf{k}_{2345}) S_4(\mathbf{k}_2, \dots, \mathbf{k}_5) - W(\mathbf{k}_1, \mathbf{k}_{2345}) \cdot T_4(\mathbf{k}_2, \dots, \mathbf{k}_5) \right] \\
&\quad + \frac{70}{13} S_2(\mathbf{k}_1, \mathbf{k}_2) \left[ U(\mathbf{k}_{12}, \mathbf{k}_{345}) S_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) - W(\mathbf{k}_{12}, \mathbf{k}_{345}) \cdot T_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) \right] \\
&\quad - \frac{60}{13} \left\{ V(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_{345}) S_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) + \frac{(\mathbf{k}_1 \times \mathbf{k}_2) \cdot \mathbf{k}_{345}}{k_1^2 k_2^2 k_{345}^2} [(\mathbf{k}_1 \times \mathbf{k}_2) \times \mathbf{k}_{345}] \cdot T_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) \right\} \\
&\quad - \frac{75}{13} V(\mathbf{k}_1, \mathbf{k}_{23}, \mathbf{k}_{45}) S_2(\mathbf{k}_2, \mathbf{k}_3) S_2(\mathbf{k}_4, \mathbf{k}_5), \\
T_5(\mathbf{k}_1, \dots, \mathbf{k}_5) &= 3 \left[ W(\mathbf{k}_1, \mathbf{k}_{2345}) S_4(\mathbf{k}_2, \dots, \mathbf{k}_5) + \frac{\mathbf{k}_1 \times \mathbf{k}_{2345}}{k_1^2 k_{2345}^2} (\mathbf{k}_1 \times \mathbf{k}_{2345}) \cdot T_4(\mathbf{k}_2, \dots, \mathbf{k}_5) \right] \\
&\quad + 2 S_2(\mathbf{k}_1, \mathbf{k}_2) \left[ W(\mathbf{k}_{12}, \mathbf{k}_{345}) S_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) + \frac{\mathbf{k}_{12} \times \mathbf{k}_{345}}{k_{12}^2 k_{345}^2} (\mathbf{k}_{12} \times \mathbf{k}_{345}) \cdot T_3(\mathbf{k}_3, \mathbf{k}_4, \mathbf{k}_5) \right].
\end{aligned}$$

$$\begin{aligned}
S_6(\mathbf{k}_1, \dots, \mathbf{k}_6) = & \frac{22}{5} \left[ U(\mathbf{k}_1, \mathbf{k}_{23456}) S_5(\mathbf{k}_2, \dots, \mathbf{k}_6) - W(\mathbf{k}_1, \mathbf{k}_{23456}) \cdot T_5(\mathbf{k}_2, \dots, \mathbf{k}_6) \right] \\
& + \frac{43}{5} S_2(\mathbf{k}_1, \mathbf{k}_2) \left[ U(\mathbf{k}_{12}, \mathbf{k}_{3456}) S_4(\mathbf{k}_3, \dots, \mathbf{k}_6) - W(\mathbf{k}_{12}, \mathbf{k}_{3456}) \cdot T_4(\mathbf{k}_3, \dots, \mathbf{k}_6) \right] \\
& + \frac{26}{5} \left[ S_3(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3) [U(\mathbf{k}_{123}, \mathbf{k}_{456}) S_3(\mathbf{k}_4, \mathbf{k}_5, \mathbf{k}_6) - 2W(\mathbf{k}_{123}, \mathbf{k}_{456}) \cdot T_3(\mathbf{k}_4, \mathbf{k}_5, \mathbf{k}_6)] \right. \\
& \quad \left. + \frac{\mathbf{k}_{123} \times \mathbf{k}_{456}}{k_{123}^2 k_{456}^2} \cdot \{ [\mathbf{k}_{123} \times T_3(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_3)] \times [\mathbf{k}_{456} \times T_3(\mathbf{k}_4, \mathbf{k}_5, \mathbf{k}_6)] \} \right] \\
& - \frac{39}{5} \left\{ V(\mathbf{k}_1, \mathbf{k}_2, \mathbf{k}_{3456}) S_4(\mathbf{k}_3, \dots, \mathbf{k}_6) + \frac{(\mathbf{k}_1 \times \mathbf{k}_2) \cdot \mathbf{k}_{3456}}{k_1^2 k_2^2 k_{3456}^2} [(\mathbf{k}_1 \times \mathbf{k}_2) \times \mathbf{k}_{3456}] \cdot T_4(\mathbf{k}_3, \dots, \mathbf{k}_6) \right\} \\
& - \frac{124}{5} S_2(\mathbf{k}_2, \mathbf{k}_3) \left\{ V(\mathbf{k}_1, \mathbf{k}_{23}, \mathbf{k}_{456}) S_3(\mathbf{k}_4, \mathbf{k}_5, \mathbf{k}_6) + \frac{(\mathbf{k}_1 \times \mathbf{k}_{23}) \cdot \mathbf{k}_{456}}{k_1^2 k_{23}^2 k_{456}^2} [(\mathbf{k}_1 \times \mathbf{k}_{23}) \times \mathbf{k}_{456}] \cdot T_3(\mathbf{k}_4, \mathbf{k}_5, \mathbf{k}_6) \right\} \\
& - \frac{27}{5} V(\mathbf{k}_{12}, \mathbf{k}_{34}, \mathbf{k}_{56}) S_2(\mathbf{k}_1, \mathbf{k}_2) S_2(\mathbf{k}_3, \mathbf{k}_4) S_2(\mathbf{k}_5, \mathbf{k}_6), \tag{B1}
\end{aligned}$$

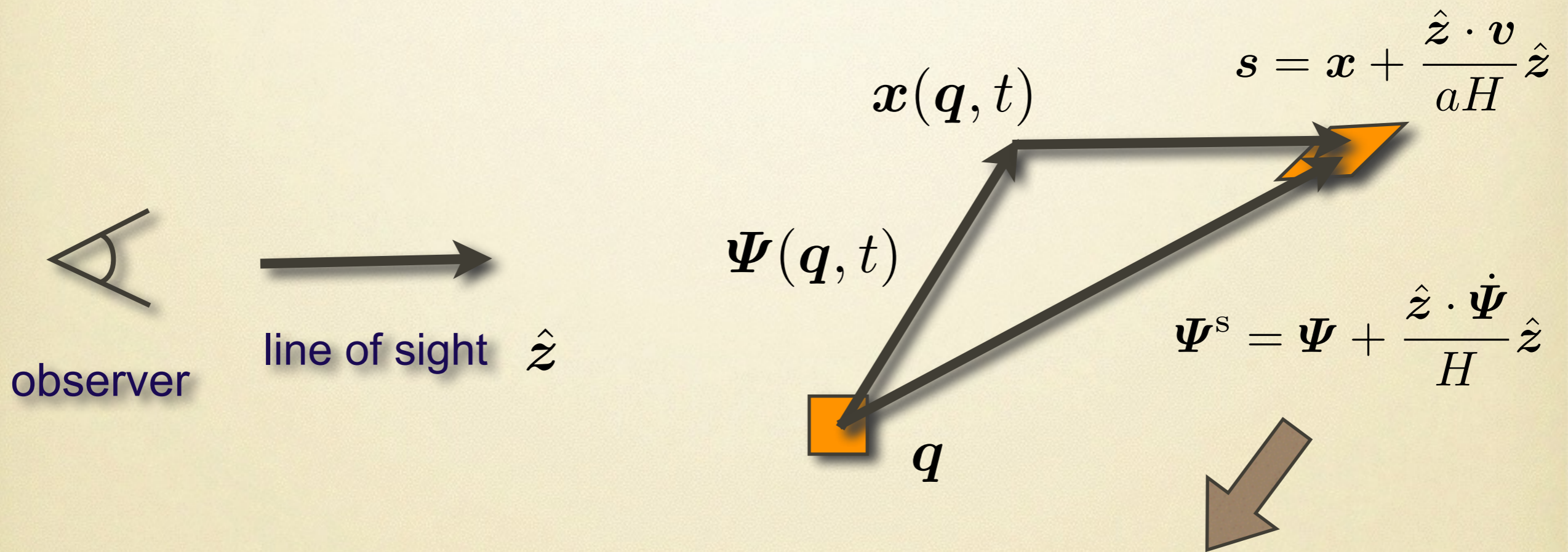
$$\begin{aligned}
T_6(\mathbf{k}_1, \dots, \mathbf{k}_6) = & 4 \left[ W(\mathbf{k}_1, \mathbf{k}_{23456}) S_5(\mathbf{k}_2, \dots, \mathbf{k}_6) + \frac{\mathbf{k}_1 \times \mathbf{k}_{23456}}{k_1^2 k_{23456}^2} (\mathbf{k}_1 \times \mathbf{k}_{23456}) \cdot T_5(\mathbf{k}_2, \dots, \mathbf{k}_6) \right] \\
& + 5 S_2(\mathbf{k}_1, \mathbf{k}_2) \left[ W(\mathbf{k}_{12}, \mathbf{k}_{3456}) S_4(\mathbf{k}_3, \dots, \mathbf{k}_6) + \frac{\mathbf{k}_{12} \times \mathbf{k}_{3456}}{k_{12}^2 k_{3456}^2} (\mathbf{k}_{12} \times \mathbf{k}_{3456}) \cdot T_4(\mathbf{k}_3, \dots, \mathbf{k}_6) \right]. \tag{B2}
\end{aligned}$$

$$\begin{aligned}
S_7(k_1, \dots, k_7) = & \frac{91}{17} \left[ U(k_1, k_{234567}) S_6(k_2, \dots, k_7) - W(k_1, k_{234567}) \cdot T_6(k_2, \dots, k_7) \right] \\
& + \frac{217}{17} S_2(k_1, k_2) \left[ U(k_{12}, k_{34567}) S_5(k_3, \dots, k_7) - W(k_{12}, k_{34567}) \cdot T_5(k_3, \dots, k_7) \right] \\
& + \frac{315}{17} \left[ U(k_{123}, k_{4567}) S_3(k_1, k_2, k_3) S_4(k_4, \dots, k_7) \right. \\
& \quad - W(k_{123}, k_{4567}) \cdot [S_3(k_1, k_2, k_3) T_4(k_4, \dots, k_7) - T_3(k_1, k_2, k_3) S_4(k_4, \dots, k_7)] \\
& \quad \left. + \frac{k_{123} \times k_{4567}}{k_{123}^2 k_{4567}^2} \cdot \{ [k_{123} \times T_3(k_1, k_2, k_3)] \times [k_{4567} \times T_4(k_4, \dots, k_7)] \} \right] \\
& - \frac{203}{17} \left\{ V(k_1, k_2, k_{34567}) S_4(k_3, \dots, k_7) + \frac{(k_1 \times k_2) \cdot k_{34567}}{k_1^2 k_2^2 k_{34567}^2} [(k_1 \times k_2) \times k_{34567}] \cdot T_5(k_3, \dots, k_7) \right\} \\
& - \frac{805}{17} S_2(k_2, k_3) \left\{ V(k_1, k_{23}, k_{4567}) S_4(k_4, \dots, k_7) + \frac{(k_1 \times k_{23}) \cdot k_{4567}}{k_1^2 k_{23}^2 k_{4567}^2} [(k_1 \times k_{23}) \times k_{4567}] \cdot T_4(k_4, \dots, k_7) \right\} \\
& - \frac{490}{17} \left[ V(k_1, k_{234}, k_{567}) S_3(k_2, k_3, k_4) S_3(k_5, k_6, k_7) \right. \\
& \quad + 2S_3(k_2, k_3, k_4) \frac{(k_1 \times k_{234}) \cdot k_{567}}{k_1^2 k_{234}^2 k_{567}^2} [(k_1 \times k_{234}) \times k_{567}] \cdot T_3(k_5, k_6, k_7) \\
& \quad \left. + \frac{(k_1 \times k_{234}) \cdot k_{567}}{k_1^2 k_{234}^2 k_{567}^2} k_1 \cdot \{ [k_{234} \times T_3(k_2, k_3, k_4)] \times [k_{567} \times T_3(k_5, k_6, k_7)] \} \right] \\
& - \frac{665}{17} S_2(k_1, k_2) S_2(k_3, k_4) \left\{ V(k_{12}, k_{34}, k_{567}) S_3(k_5, k_6, k_7) \right. \\
& \quad \left. + \frac{(k_{12} \times k_{34}) \cdot k_{567}}{k_{12}^2 k_{34}^2 k_{567}^2} [(k_{12} \times k_{34}) \times k_{567}] \cdot T_3(k_5, k_6, k_7) \right\}, \tag{B17}
\end{aligned}$$

$$\begin{aligned}
T_7(k_1, \dots, k_7) = & 5 \left[ W(k_1, k_{234567}) S_6(k_2, \dots, k_7) + \frac{k_1 \times k_{234567}}{k_1^2 k_{234567}^2} (k_1 \times k_{234567}) \cdot T_6(k_2, \dots, k_7) \right] \\
& + 9S_2(k_1, k_2) \left[ W(k_{12}, k_{34567}) S_5(k_3, \dots, k_7) + \frac{k_{12} \times k_{34567}}{k_{12}^2 k_{34567}^2} (k_{12} \times k_{34567}) \cdot T_5(k_3, \dots, k_7) \right] \\
& + 5 \left[ W(k_{123}, k_{4567}) S_3(k_1, k_2, k_3) S_4(k_4, \dots, k_7) \right. \\
& \quad + \frac{k_{123} \times k_{4567}}{k_{123}^2 k_{4567}^2} \{ (k_{123} \times k_{4567}) \cdot [S_3(k_1, k_2, k_3) T_4(k_4, \dots, k_7) - T_3(k_1, k_2, k_3) S_4(k_4, \dots, k_7)] \\
& \quad \left. + [k_{123} \times T_3(k_1, k_2, k_3)] \cdot [k_{4567} \times T_4(k_4, \dots, k_7)] \} \right]. \tag{B18}
\end{aligned}$$

# Redshift-space distortions in Lagrangian PT

- Lagrangian variables and redshift space distortions



$$\Psi^{s(n)} = \Psi^{(n)} + \frac{\hat{z} \cdot \dot{\Psi}^{(n)}}{H} \hat{z} \simeq \Psi^{(n)} + n f(\hat{z} \cdot \Psi^{(n)}) \hat{z}$$

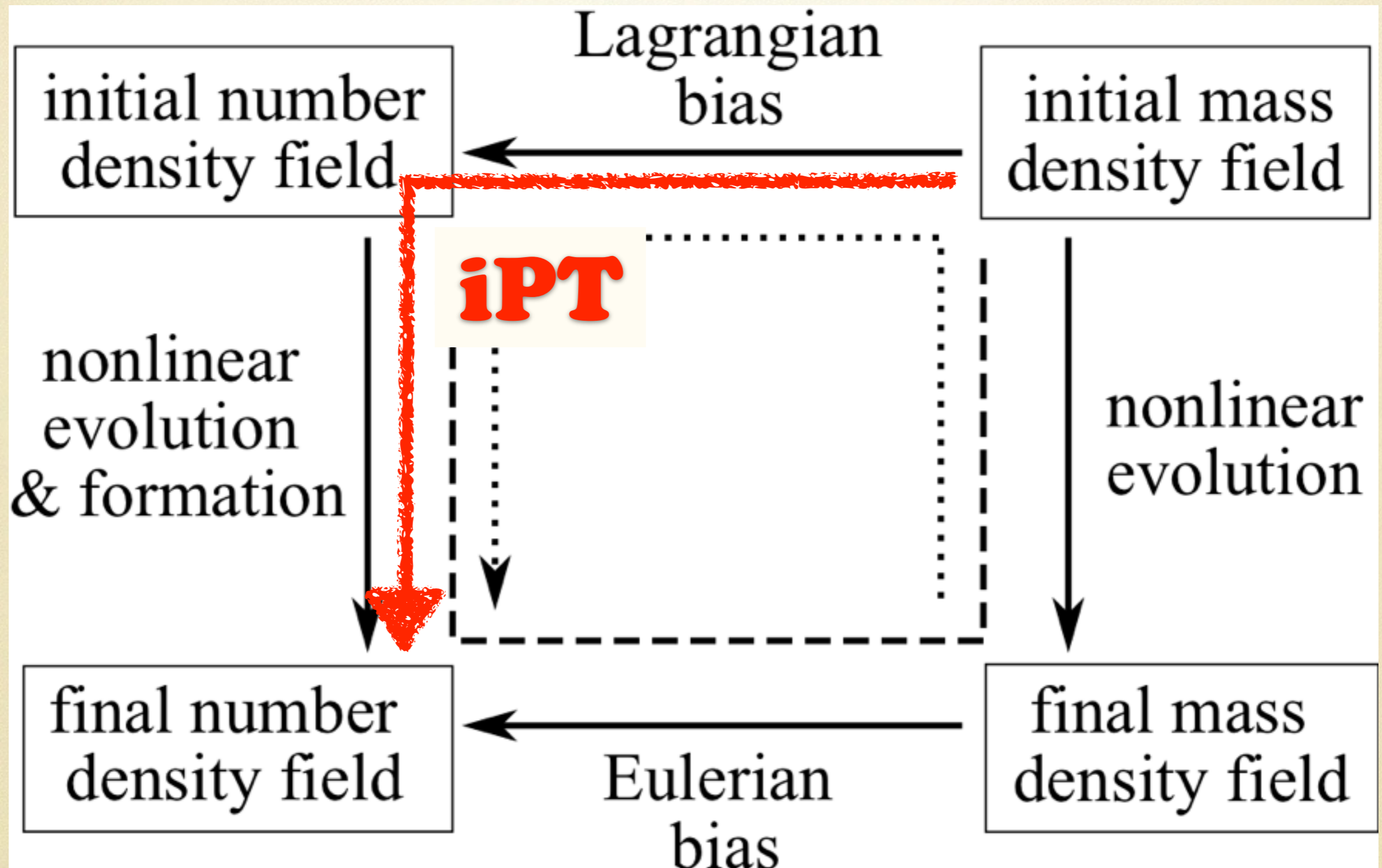
# The integrated Perturbation Theory (iPT)



# iPT

- **integrated Perturbation Theory**
  - **integration of:**
    - **nonlinear perturbation theory**
    - **nonlinear redshift-space distortion**
    - **nonlinear bias (nonlocal in general)**
    - **primordial non-Gaussianity**
  - **iPT does not primarily intend to extrapolate PT into strongly nonlinear regime**
  - **It provides a basic framework to predict observable quantities from the LPT**

# Dynamical evolution of Lagrangian bias



# iPT in a nutshell

- The relation between Eulerian density fluctuations and Lagrangian variables

$$1 + \delta_X(\mathbf{x}) = \int d^3q \left[ 1 + \delta_X^L(\mathbf{q}) \right] \delta_D^3[\mathbf{x} - \mathbf{q} - \boldsymbol{\Psi}(\mathbf{q})]$$

Eulerian  
density field

Biased field in  
Lagrangian space

displacement  
(& redshift distortions)

- Perturbative expansion in Fourier space

$$\delta_X^L(\mathbf{k}) = \sum_{n=1}^{\infty} \frac{1}{n!} \int \frac{d^3k_1}{(2\pi)^3} \cdots \frac{d^3k_n}{(2\pi)^3} (2\pi)^3 \delta_D^3(\mathbf{k}_{1\dots n} - \mathbf{k}) b_n^L(\mathbf{k}_1, \dots, \mathbf{k}_n) \delta_L(\mathbf{k}_1) \cdots \delta_L(\mathbf{k}_n)$$

$$\tilde{\boldsymbol{\Psi}}(\mathbf{k}) = \sum_{n=1}^{\infty} \frac{i}{n!} \int \frac{d^3k_1}{(2\pi)^3} \cdots \frac{d^3k_n}{(2\pi)^3} (2\pi)^3 \delta_D^3(\mathbf{k}_{1\dots n} - \mathbf{k}) \mathbf{L}_n(\mathbf{k}_1, \dots, \mathbf{k}_n) \delta_L(\mathbf{k}_1) \cdots \delta_L(\mathbf{k}_n)$$

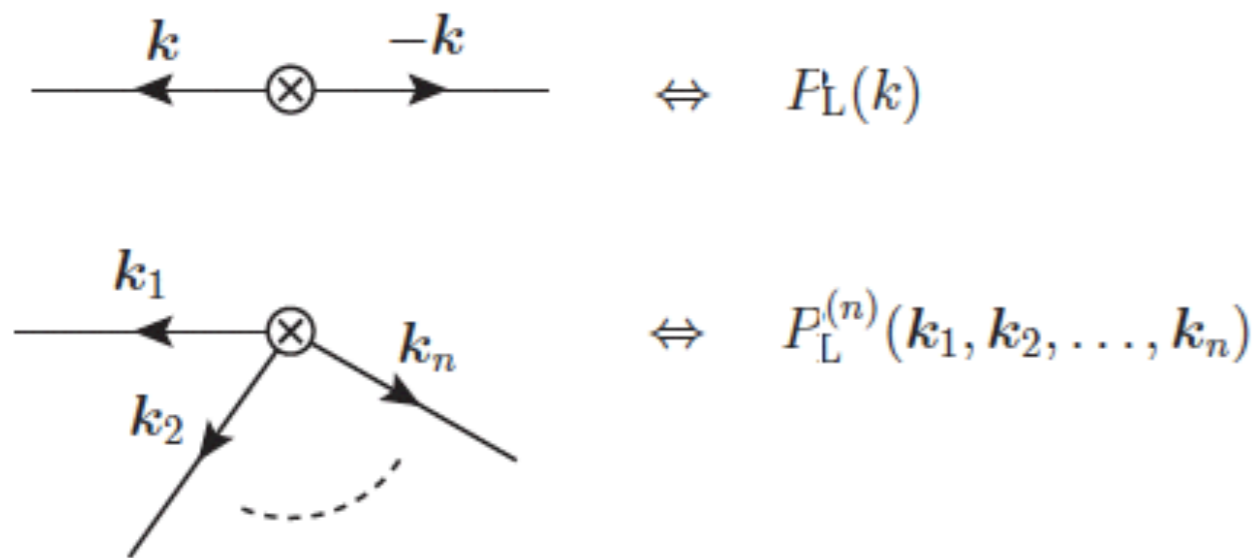
Kernel of the Lagrangian bias

Kernel of the displacement field (& redshift distortions)

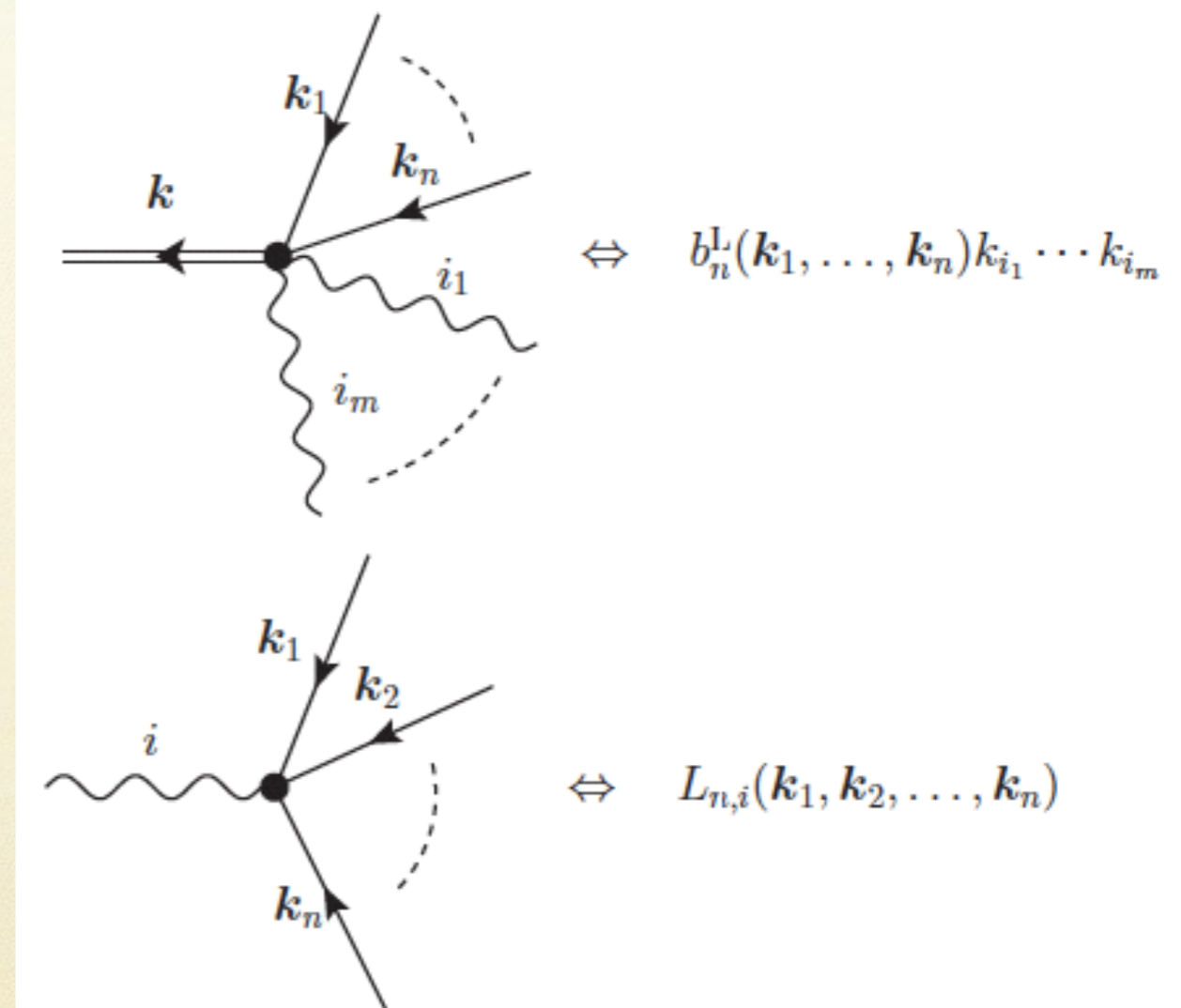
$$\mathbf{k}_{1\dots n} \equiv \mathbf{k}_1 + \cdots + \mathbf{k}_n$$

# Diagrams in iPT

Primordial spectra



Vertices in Lagrangian PT



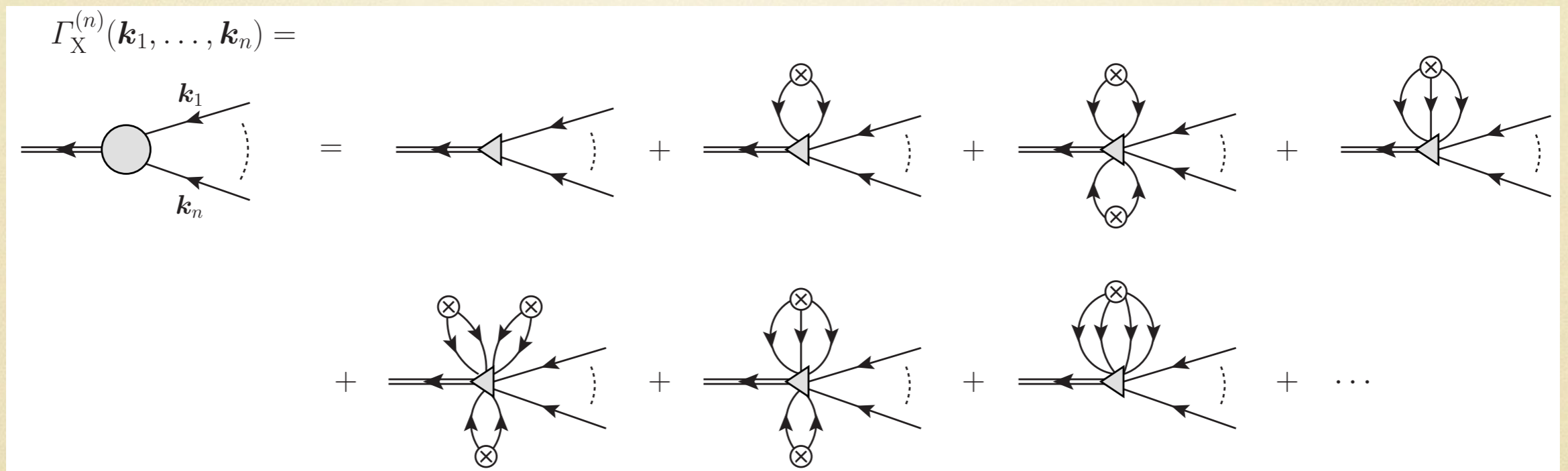
can naturally deal with RSD and nG

# Multi-point propagator

TM (1995); Crocce & Scoccimarro (2006); Bernardeau et al. (2008); TM (2011)

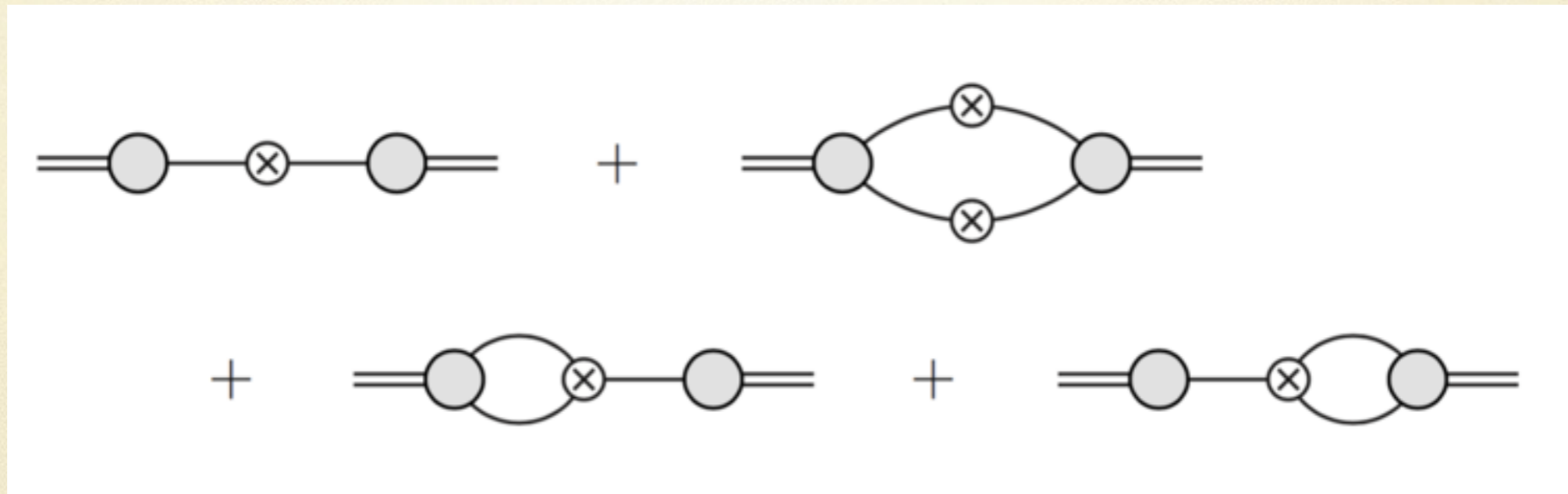
- Density sector of multi-point propagator with nonlocal bias and RSD

$$\left\langle \frac{\delta^n \delta_X(\mathbf{k})}{\delta\delta_L(\mathbf{k}_1) \cdots \delta\delta_L(\mathbf{k}_n)} \right\rangle = (2\pi)^{3-3n} \delta_D^3(\mathbf{k} - \mathbf{k}_{1\dots n}) \Gamma_X^{(n)}(\mathbf{k}_1, \dots, \mathbf{k}_n)$$



# One-loop power spectrum

- In the formalism of multi-point propagator,

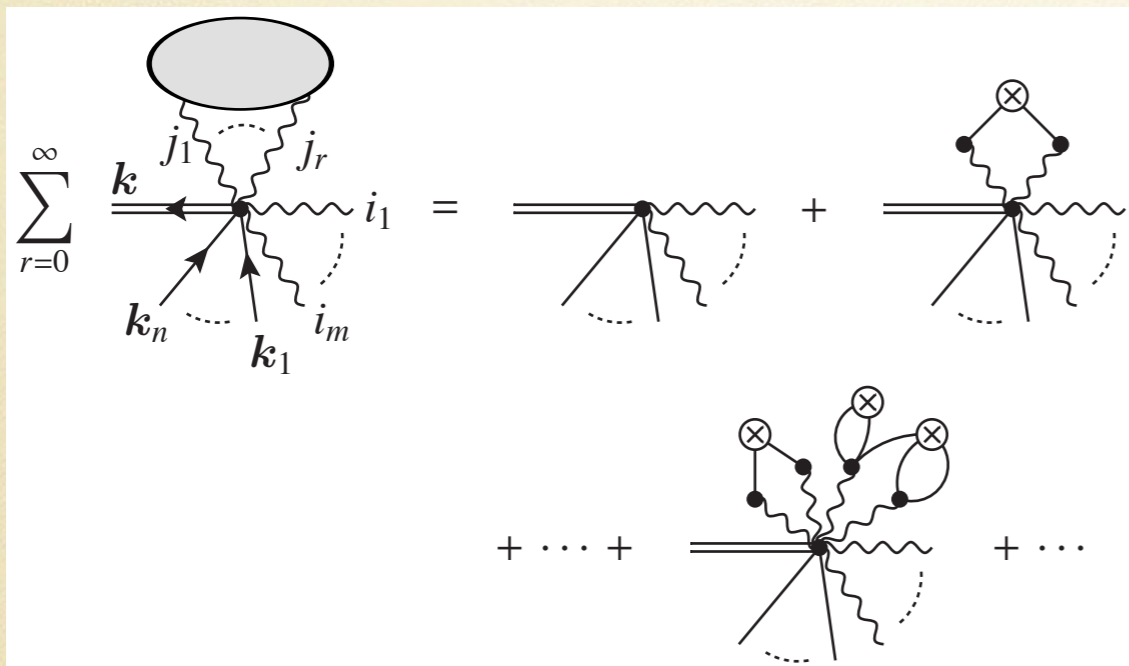


$$\begin{aligned}
 P_X(\mathbf{k}) &= \left[ \Gamma_X^{(1)}(\mathbf{k}) \right]^2 P_L(k) \\
 &+ \frac{1}{2} \int_{\mathbf{k}_{12}=\mathbf{k}} \left[ \Gamma_X^{(2)}(\mathbf{k}_1, \mathbf{k}_2) \right]^2 P_L(k_1) P_L(k_2) \\
 &+ \Gamma_X^{(1)}(\mathbf{k}) \int_{\mathbf{k}_{12}=\mathbf{k}} \Gamma_X^{(2)}(\mathbf{k}_1, \mathbf{k}_2) B_L(k, k_1, k_2),
 \end{aligned}$$

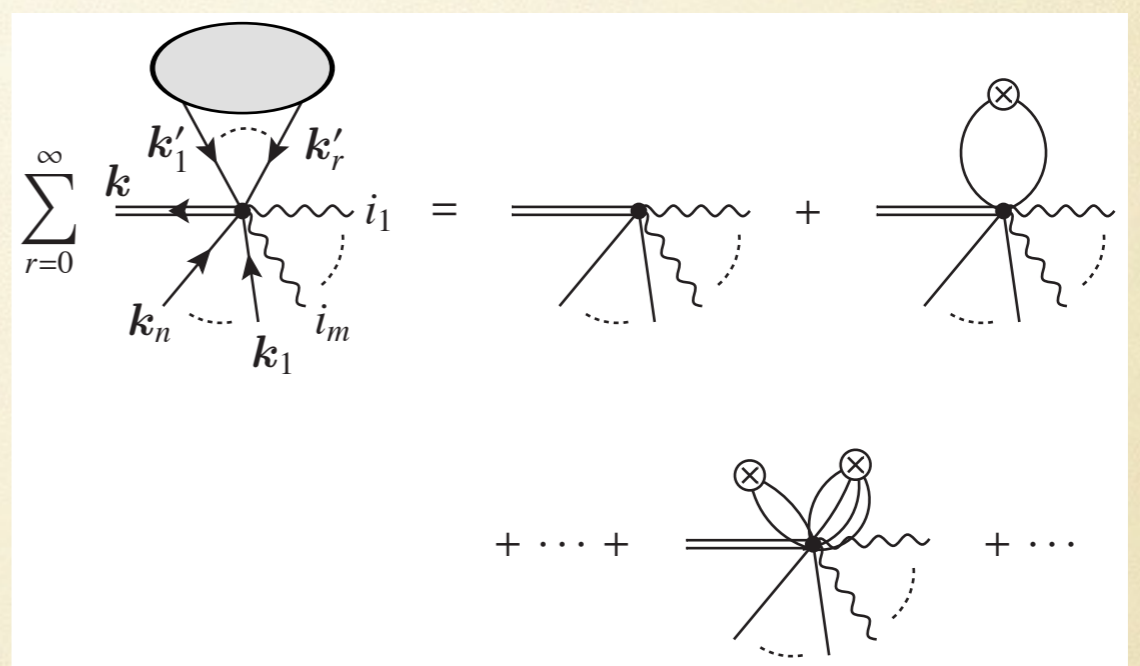
# Multi-point propagator

- Full evaluations of MP propagator are difficult
- Partial resummations in the Lagrangian PT

## Lagrangian vertex resummation



## Lagrangian bias renormalization

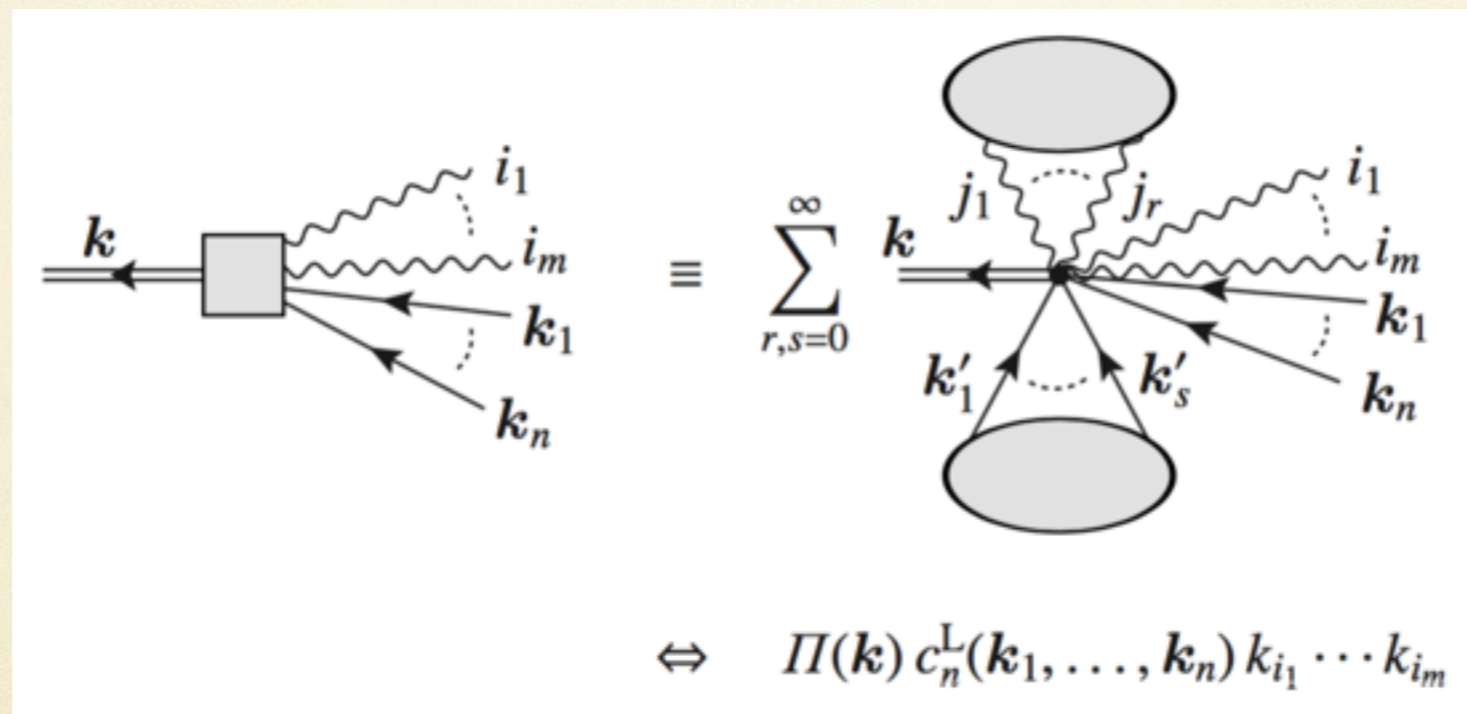


$$\begin{aligned} \Pi(\mathbf{k}) &= \langle e^{-i\mathbf{k} \cdot \Psi} \rangle \\ &= \exp \left[ \sum_{n=1}^{\infty} \frac{(-i)^n}{n!} \langle (\mathbf{k} \cdot \Psi)^n \rangle_c \right] \end{aligned}$$

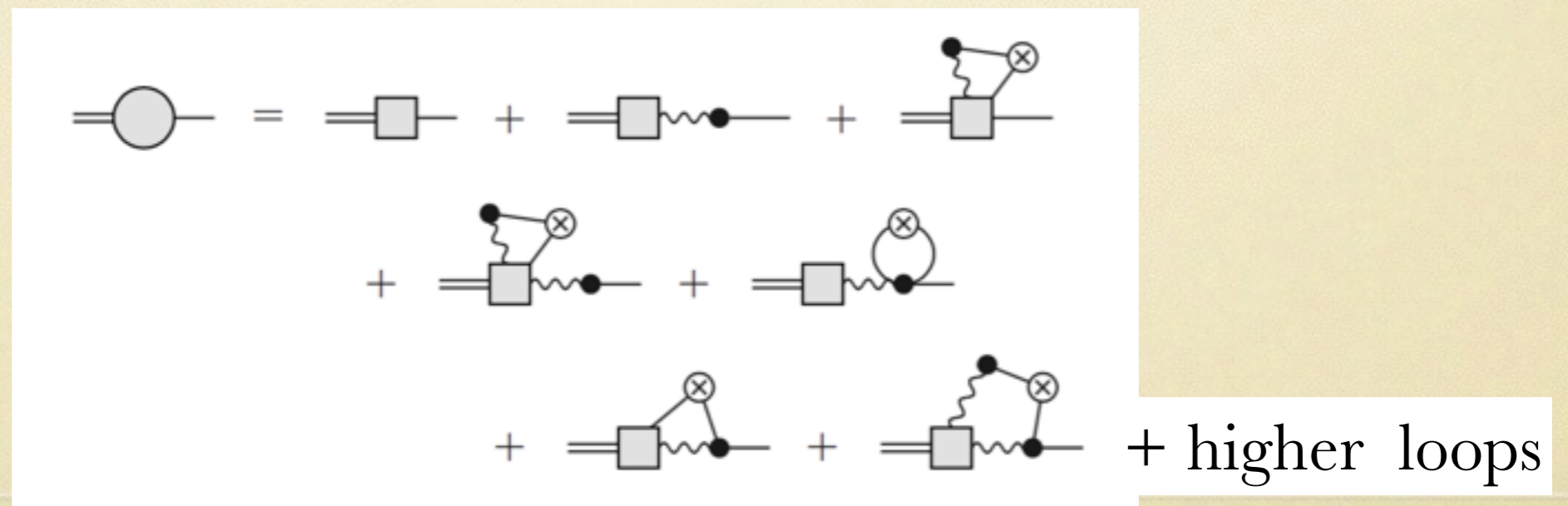
$$\begin{aligned} b_n^L(\mathbf{k}_1, \dots, \mathbf{k}_n) &= (2\pi)^{3n} \int \frac{d^3 k'}{(2\pi)^3} \frac{\delta^n \delta_X^L(\mathbf{k}')}{\delta \delta_L(\mathbf{k}_1) \cdots \delta \delta_L(\mathbf{k}_n)} \Big|_{\delta_L=0} \\ \Rightarrow c_n^L(\mathbf{k}_1, \dots, \mathbf{k}_n) &= (2\pi)^{3n} \int \frac{d^3 k'}{(2\pi)^3} \left\langle \frac{\delta^n \delta_X^L(\mathbf{k}')}{\delta \delta_L(\mathbf{k}_1) \cdots \delta \delta_L(\mathbf{k}_n)} \right\rangle \end{aligned}$$

# Vertex in iPT

- Combine the two resummations



- Ex.)

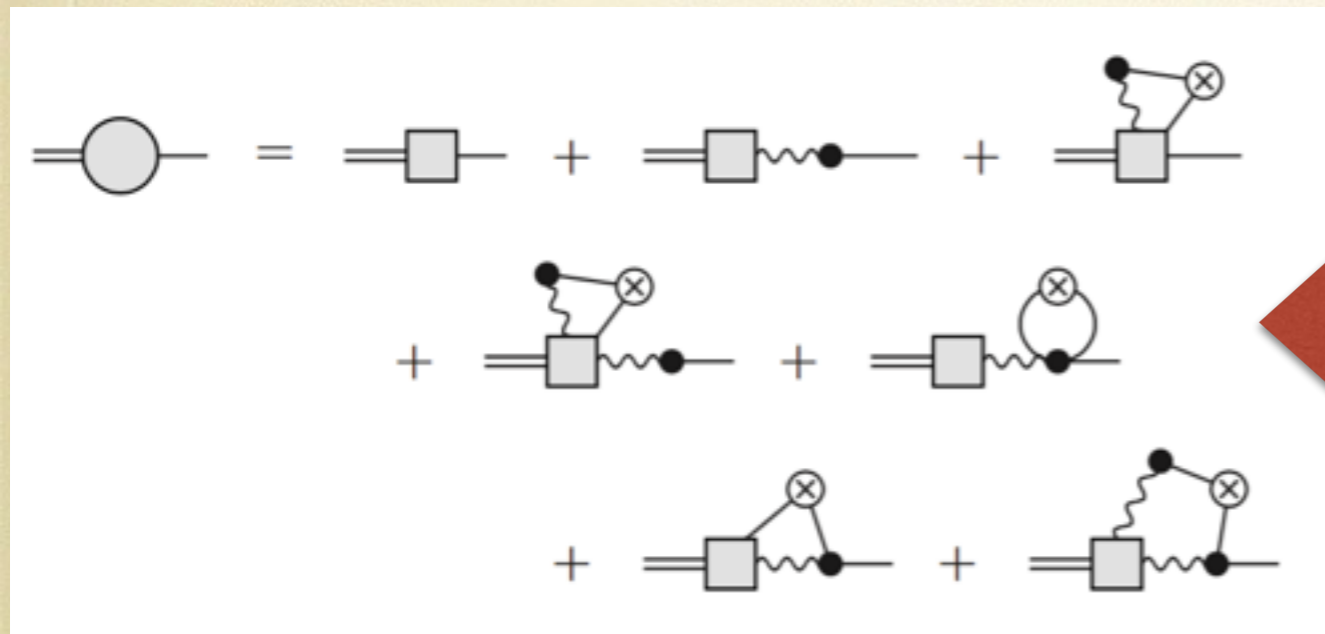




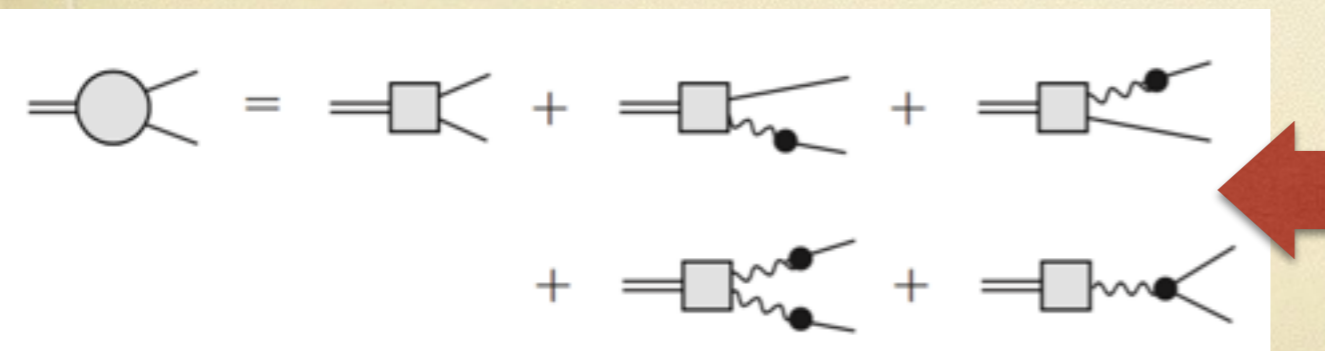
# One-loop propagators in iPT

$$\Gamma_X^{(n)}(\mathbf{k}_1, \dots, \mathbf{k}_n) = \Pi(\mathbf{k}_{1\dots n}) \hat{\Gamma}_X^{(n)}(\mathbf{k}_1, \dots, \mathbf{k}_n),$$

$$\Pi(\mathbf{k}) = \exp \left\{ -\frac{1}{2} \int \frac{d^3 p}{(2\pi)^3} \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{p}) \right]^2 P_L(p) \right\},$$

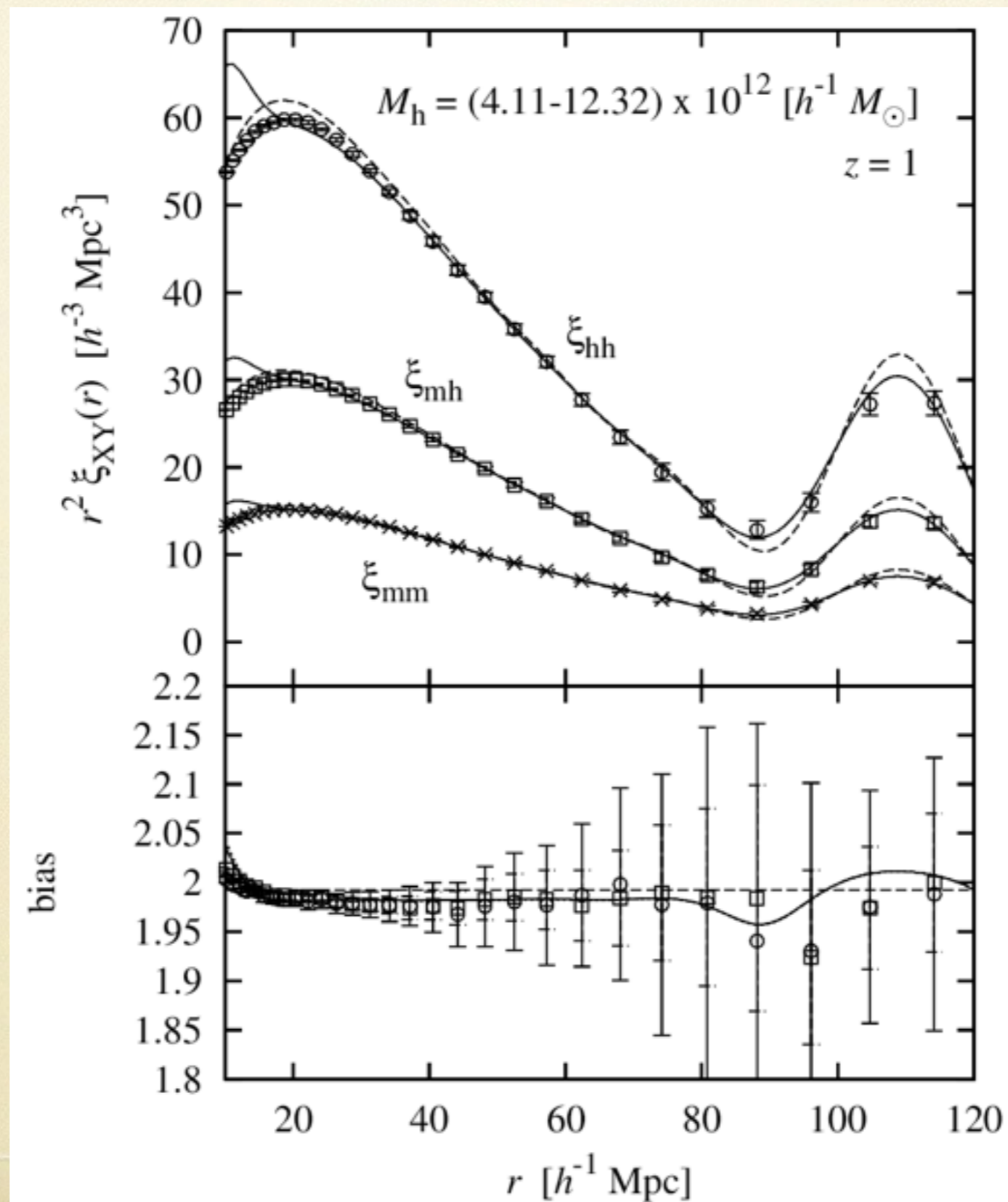


$$\begin{aligned} \hat{\Gamma}_X^{(1)}(\mathbf{k}) &= c_X^{(1)}(\mathbf{k}) + \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}) \\ &+ \int \frac{d^3 p}{(2\pi)^3} P_L(p) \left\{ c_X^{(2)}(\mathbf{k}, \mathbf{p}) \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(-\mathbf{p}) \right] \right. \\ &+ c_X^{(1)}(\mathbf{p}) \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(-\mathbf{p}) \right] \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}) \right] \\ &+ \frac{1}{2} \mathbf{k} \cdot \mathbf{L}^{(3)}(\mathbf{k}, \mathbf{p}, -\mathbf{p}) \\ &+ c_X^{(1)}(\mathbf{p}) \left[ \mathbf{k} \cdot \mathbf{L}^{(2)}(\mathbf{k}, -\mathbf{p}) \right] \\ &\left. + \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{p}) \right] \left[ \mathbf{k} \cdot \mathbf{L}^{(2)}(\mathbf{k}, -\mathbf{p}) \right] \right\}, \end{aligned}$$



$$\begin{aligned} \hat{\Gamma}_X^{(2)}(\mathbf{k}_1, \mathbf{k}_2) &= c_X^{(2)}(\mathbf{k}_1, \mathbf{k}_2) + c_X^{(1)}(\mathbf{k}_1) \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}_2) \right] \\ &+ c_X^{(1)}(\mathbf{k}_2) \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}_1) \right] + \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}_1) \right] \left[ \mathbf{k} \cdot \mathbf{L}^{(1)}(\mathbf{k}_2) \right] \\ &+ \mathbf{k} \cdot \mathbf{L}^{(2)}(\mathbf{k}_1, \mathbf{k}_2), \end{aligned}$$

# Halo clustering: slight scale-dependence of bias around BAO



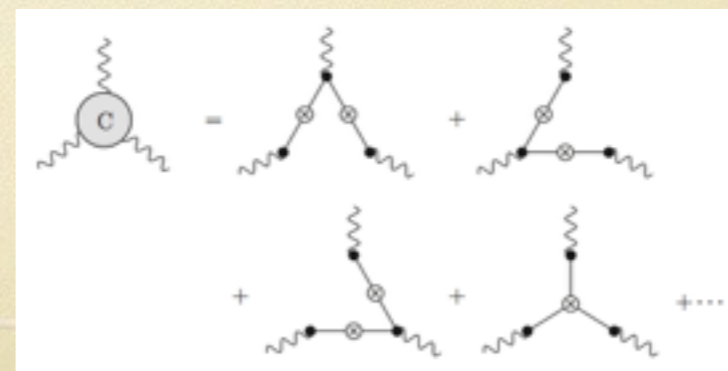
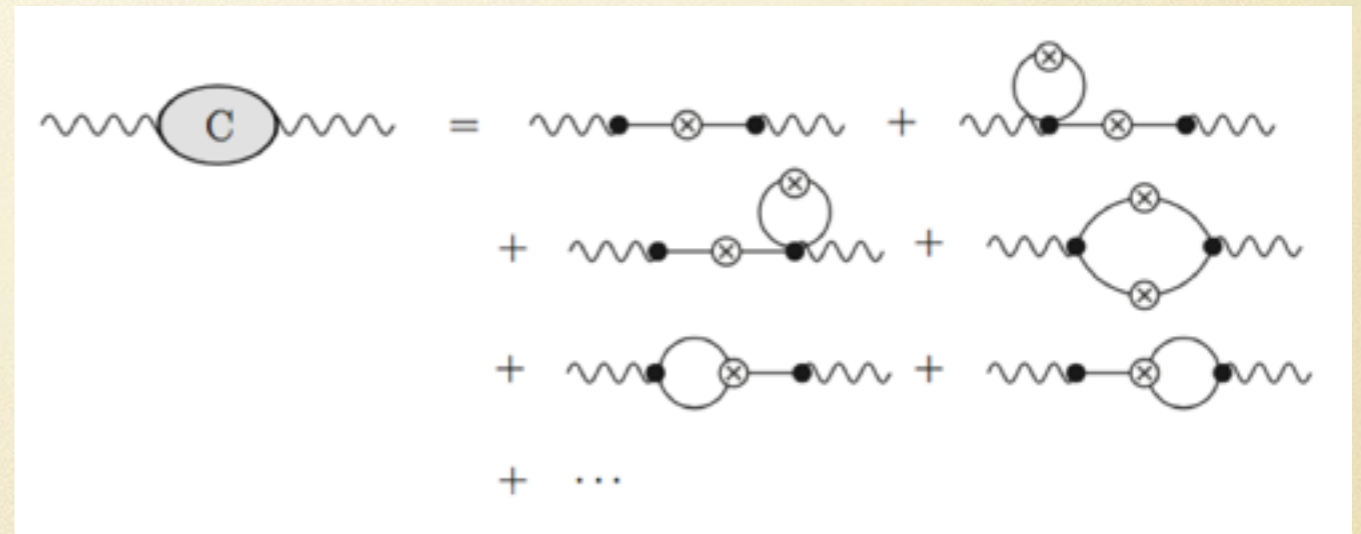
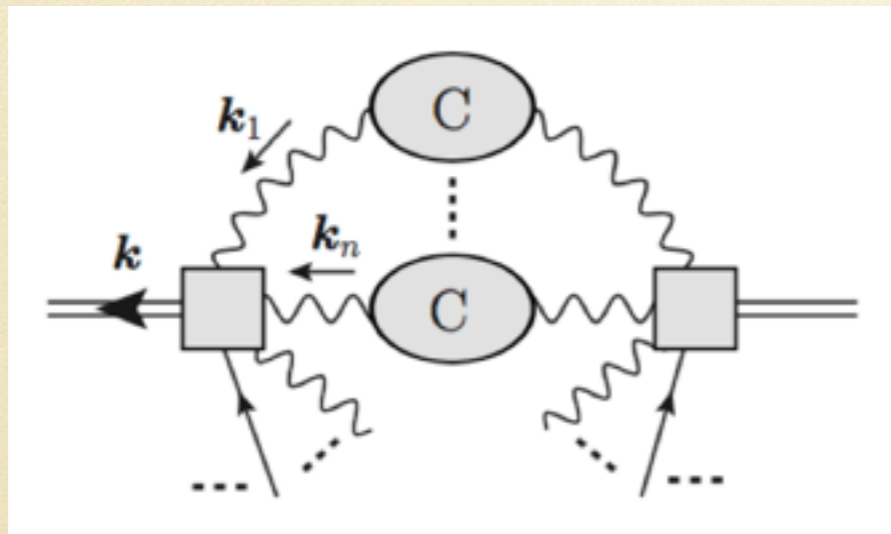
TM (2014)

# CLPT

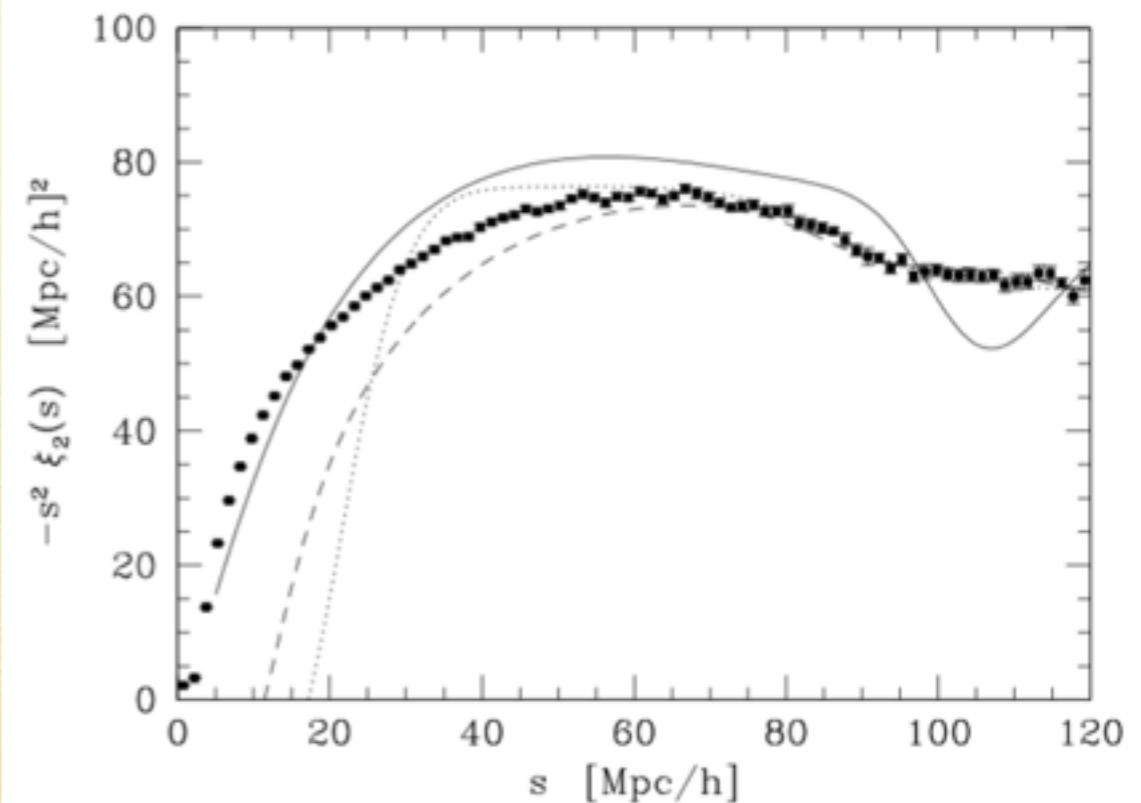
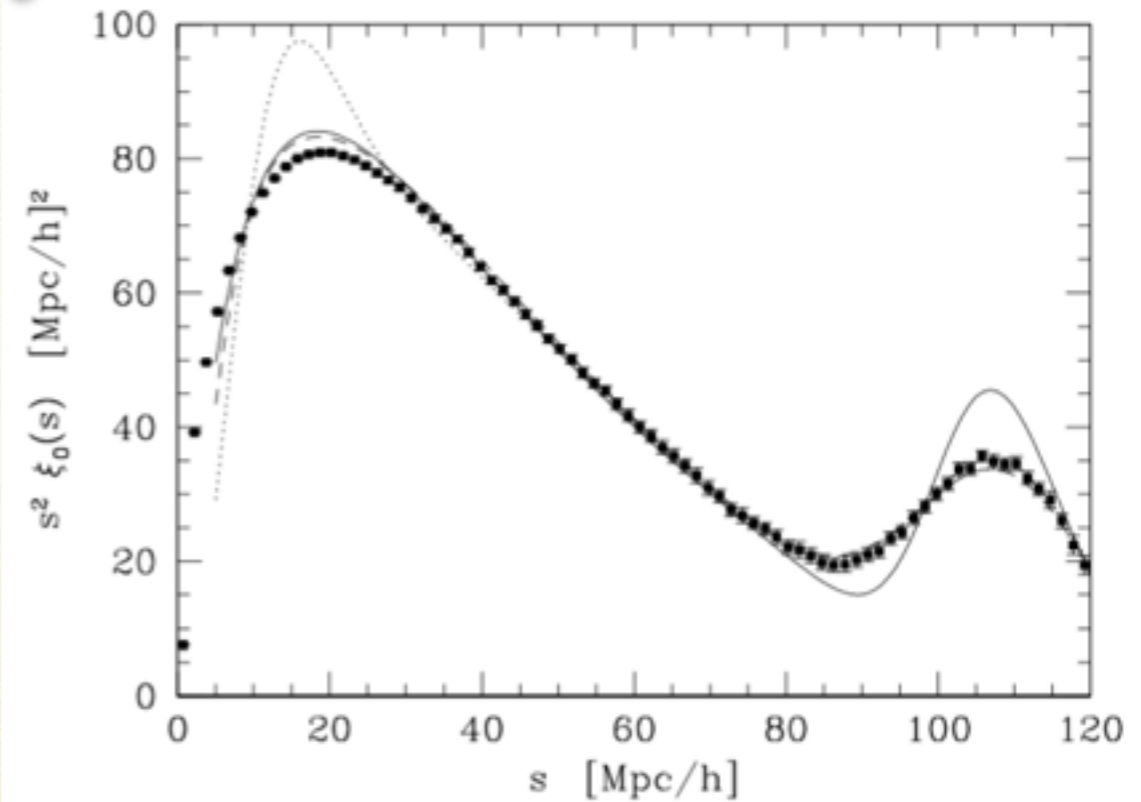
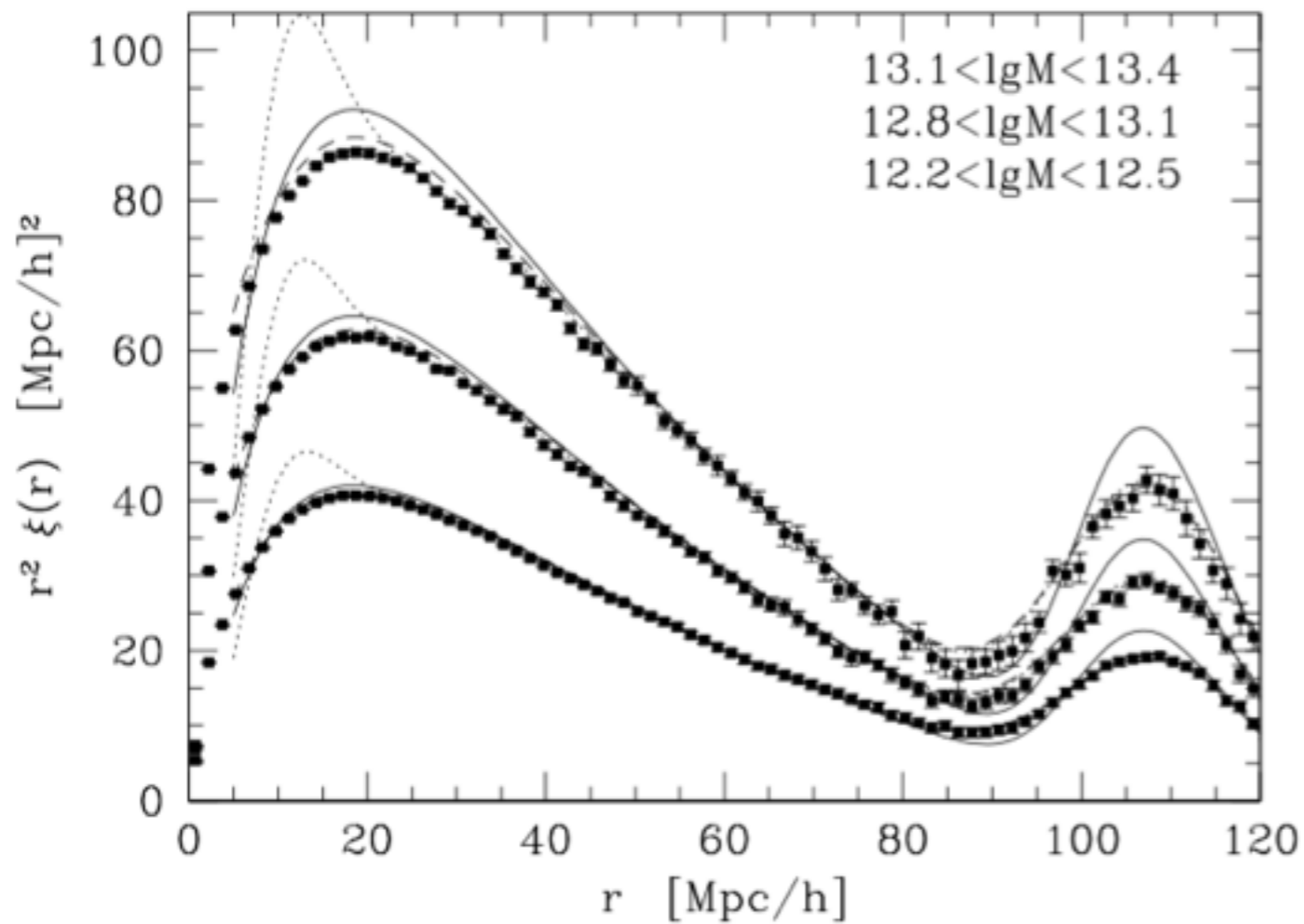
- Convolution Lagrangian Perturbation Theory (CLPT)

Carlson, Reid & White (2013);  
Vlah, Seljuk & Baldauf (2014)

- CLPT is an extension of the earlier version of iPT (LRT: Lagrangian Resummation Theory)
- CLPT extension is also straightforwardly applied for iPT



# CLPT



Carlson, Reid & White (2013)

# Impacts of biasing schemes

arXiv:1604.06579, with V. Desjacques

# Renormalized bias functions

- The “renormalized bias functions” is an essential piece in the IPT
  - Series of functions to characterize (nonlocal) biasing

$$\left\langle \frac{\delta^n \delta_X^L(\mathbf{k})}{\delta\delta_L(\mathbf{k}_1) \cdots \delta\delta_L(\mathbf{k}_n)} \right\rangle = (2\pi)^{3-3n} \delta_D^3(\mathbf{k}_{1\dots n} - \mathbf{k}) c_n^L(\mathbf{k}_1, \dots, \mathbf{k}_n). \quad (A)$$

- Counterpart of multi-point propagator for Lagrangian biasing

# Renormalized bias functions in “Halo model”

- Renormalized bias functions from Press-Schechter approach

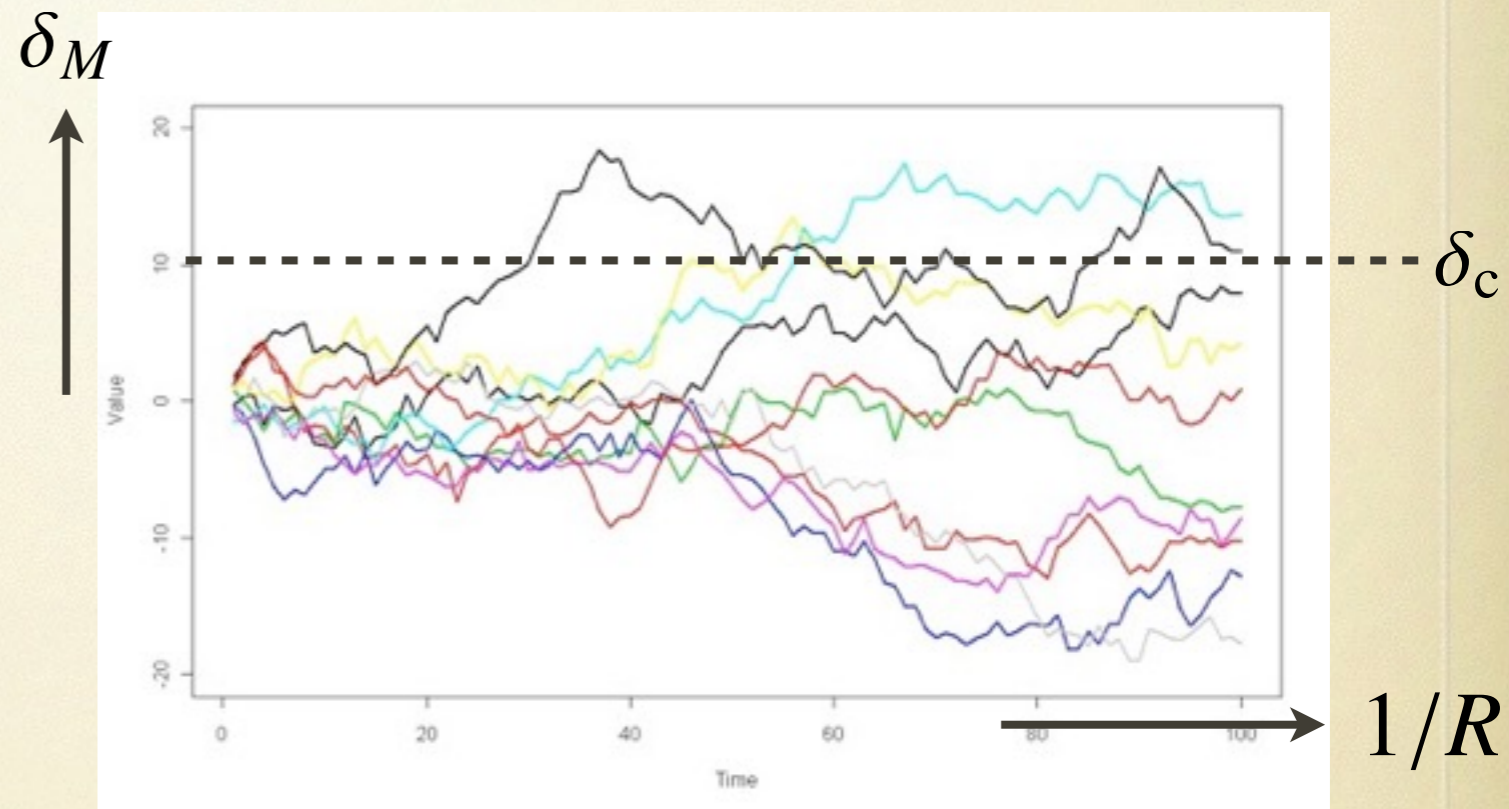
$$n(\mathbf{x}, M) = -\frac{2\bar{\rho}_0}{M} \frac{\partial}{\partial M} \Theta [\delta_M(\mathbf{x}) - \delta_c],$$

**“Localized”  
differential mass function**



$$c_1^L(k) = b_1^L W(kR) + \frac{1}{\delta_c} \frac{\partial W(kR)}{\partial \ln \sigma_M}$$

$$c_2^L(\mathbf{k}_1, \mathbf{k}_2) = b_2^L W(k_1 R) W(k_2 R) + \frac{\delta_c b_1^L + 1}{\delta_c^2} \frac{\partial}{\partial \ln \sigma_M} [W(k_1 R) W(k_2 R)]$$



# Renormalized bias functions in “Peaks model”

$$n_{\text{pk}} = \frac{3^{3/2}}{R_*^3} \delta_{\text{D}}(\nu - \nu_c) \delta_{\text{D}}^3(\boldsymbol{\eta}) \Theta(\lambda_3) |\det \boldsymbol{\zeta}|,$$

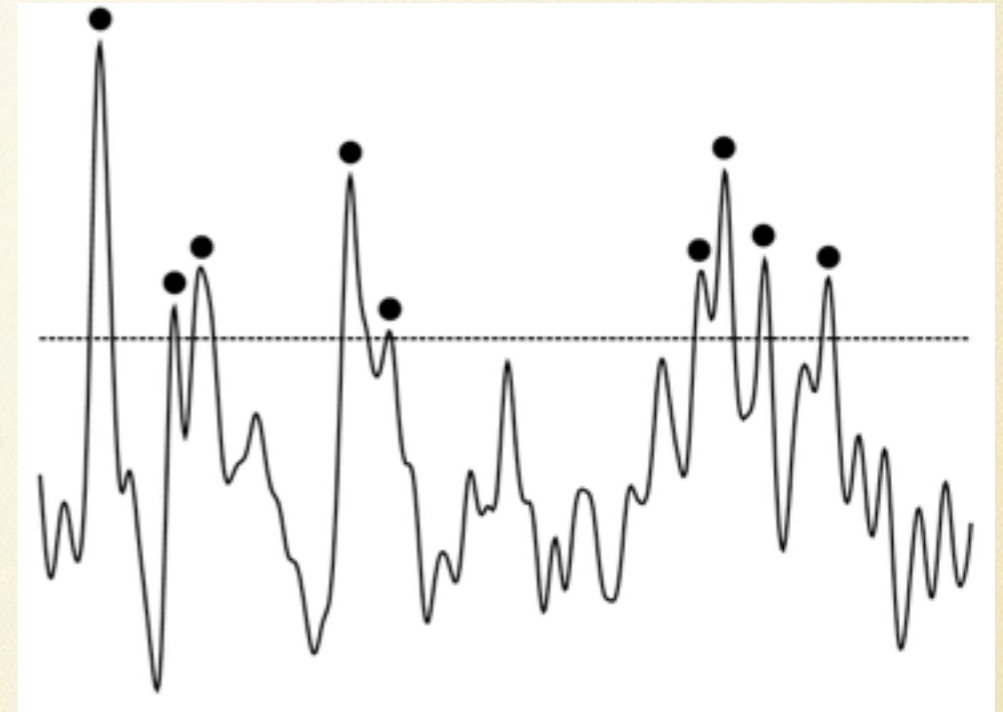
**differential  
number density of peaks**

**BBKS 1986**



$$c_1^{\text{L}}(k) = W(kR) [b_{10} + b_{11}k^2]$$

$$c_2^{\text{L}}(\mathbf{k}_1, \mathbf{k}_2) = W(k_1R)W(k_2R) \left\{ b_{20} + b_{11}(k_1^2 + k_2^2) + b_{02}k_1^2k_2^2 - 2\chi_1(\mathbf{k}_1 \cdot \mathbf{k}_2) + \omega_{10} [3(\mathbf{k}_1 \cdot \mathbf{k}_2)^2 - k_1^2k_2^2] \right\}$$



$$b_{ij} \equiv \frac{1}{\sigma_0^i \sigma_2^j \bar{n}_{\text{pk}}} \int d^{10}y n_{\text{pk}} H_{ij}(\nu, J_1) \mathcal{P},$$

$$\chi_k \equiv \frac{(-1)^k}{\sigma_1^{2k} \bar{n}_{\text{pk}}} \int d^{10}y n_{\text{pk}} L_k^{(1/2)} \left( \frac{3}{2} \eta^2 \right) \mathcal{P},$$

$$\omega_{10} \equiv \frac{(-1)^l}{\sigma_2^{2l} \bar{n}_{\text{pk}}} \int d^{10}y n_{\text{pk}} L_l^{(3/2)} \left( \frac{5}{2} J_2 \right) \mathcal{P}.$$

**Desjacques+ 2013,  
Lazeyras+ 2015**



# Renormalized bias functions in “ESP model”

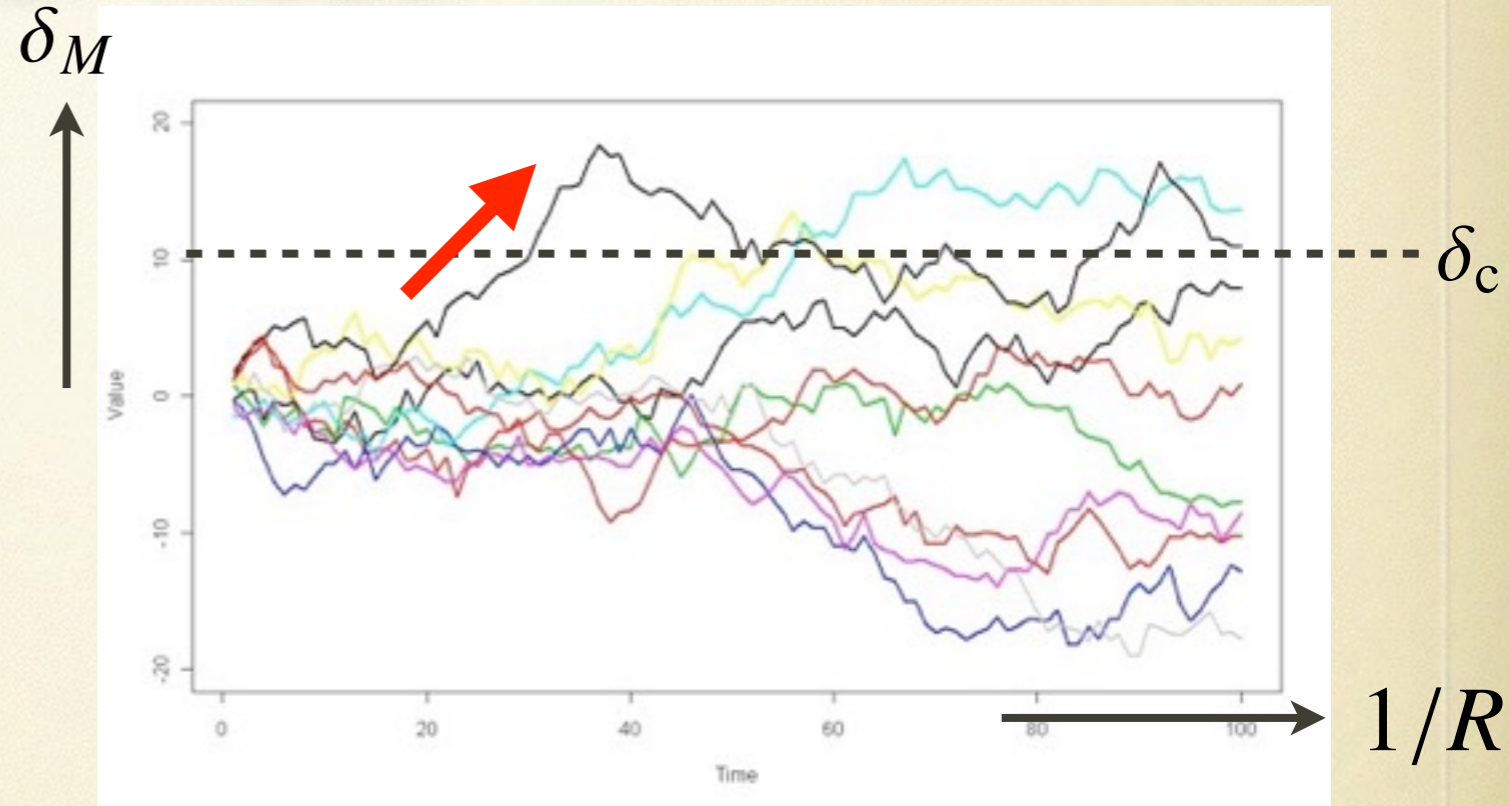
Excursion set theory  
+ Peak constraints  
+ Upcrossing constraint

Appel&Jones 1990, Desjacques 2013,  
Paranjape&Sheth 2013, Biagetti+ 2014,...

$$n_{\text{ESP}} = - \left( \frac{d\sigma_{s0}}{dR_s} \right)^{-1} \Delta_{s0} \frac{\mu_s}{v_s} \Theta(\mu_s) n_{\text{pk}},$$



$$\mu_s = - \frac{1}{\Delta_{s0}} \frac{\partial \delta_s}{\partial R_s}, \quad \Delta_{s0} = \left\langle \left( \frac{\partial \delta_s}{\partial R_s} \right)^2 \right\rangle^{1/2}.$$



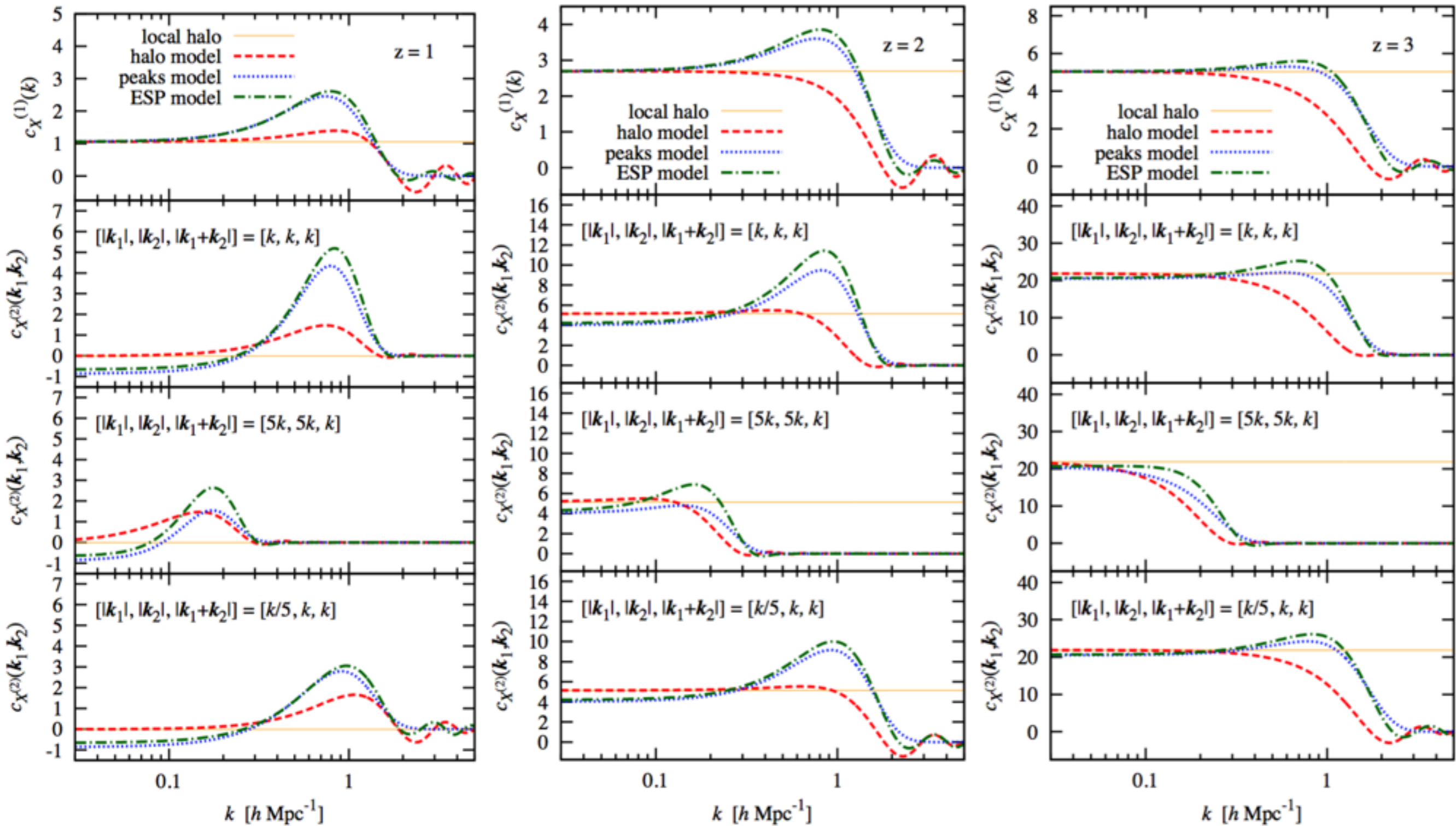
$$c_X^{(1)}(k) = b_{100} W(kR) + b_{010} k^2 \bar{W}(k\bar{R}) - b_{001} k W'(kR),$$

$$c_X^{(2)}(\mathbf{k}_1, \mathbf{k}_2) = b_{200} W(k_1 R) W(k_2 R) + b_{110} [k_2^2 W(k_1 R) \bar{W}(k_2 \bar{R}) + (1 \leftrightarrow 2)]$$

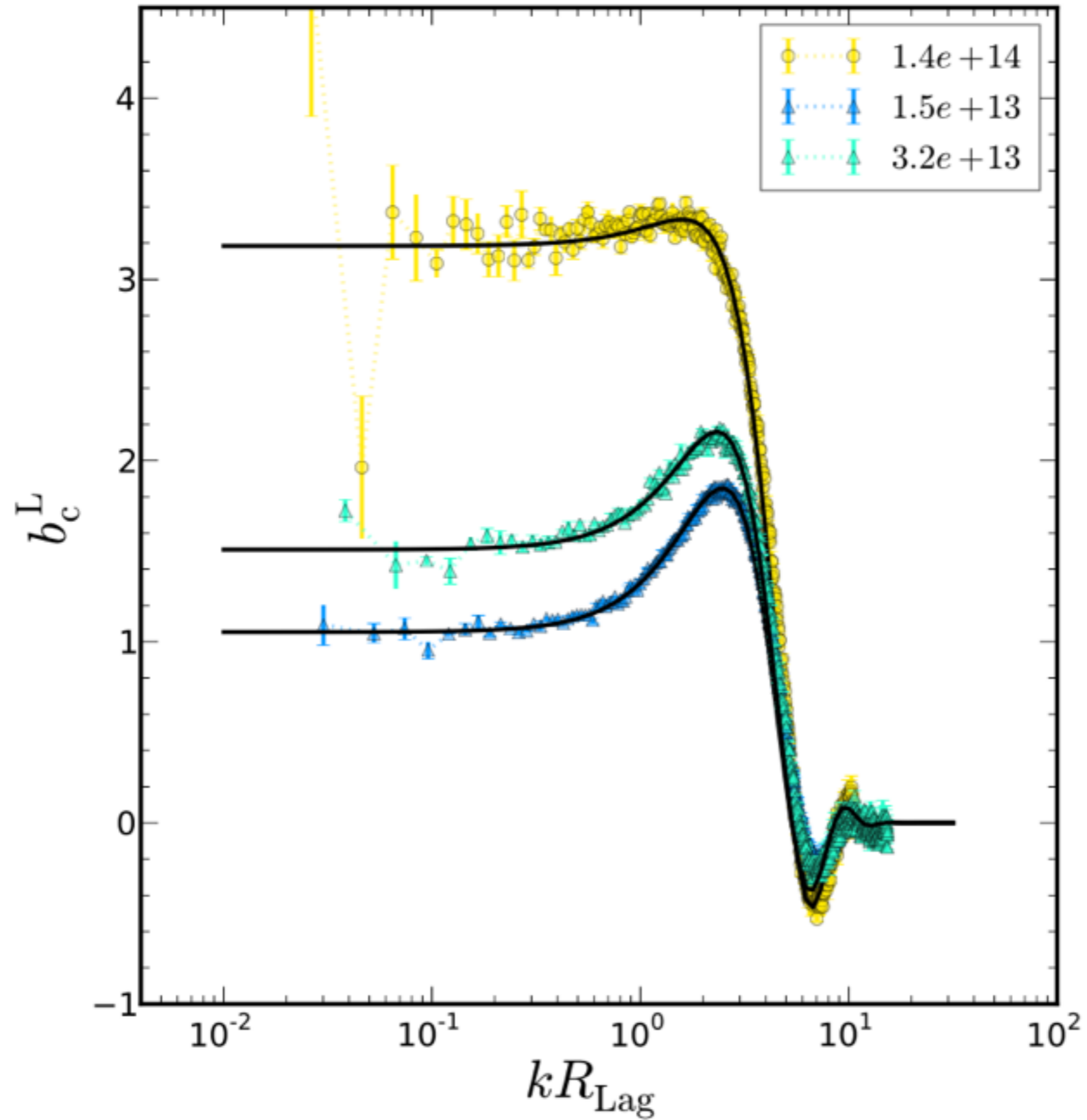
$$+ \{ b_{020} k_1^2 k_2^2 + \omega_{10} [3(\mathbf{k}_1 \cdot \mathbf{k}_2)^2 - k_1^2 k_2^2] - 2\chi_1(\mathbf{k}_1 \cdot \mathbf{k}_2) \} \bar{W}(k_1 \bar{R}) \bar{W}(k_2 \bar{R})$$

$$- b_{101} [k_1 W'(k_1 R) W(k_2 R) + (1 \leftrightarrow 2)] - b_{011} [k_1 k_2^2 W'(k_1 R) \bar{W}(k_2 \bar{R}) + (1 \leftrightarrow 2)] + b_{002} k_1 k_2 W'(k_1 R) W'(k_2 R),$$

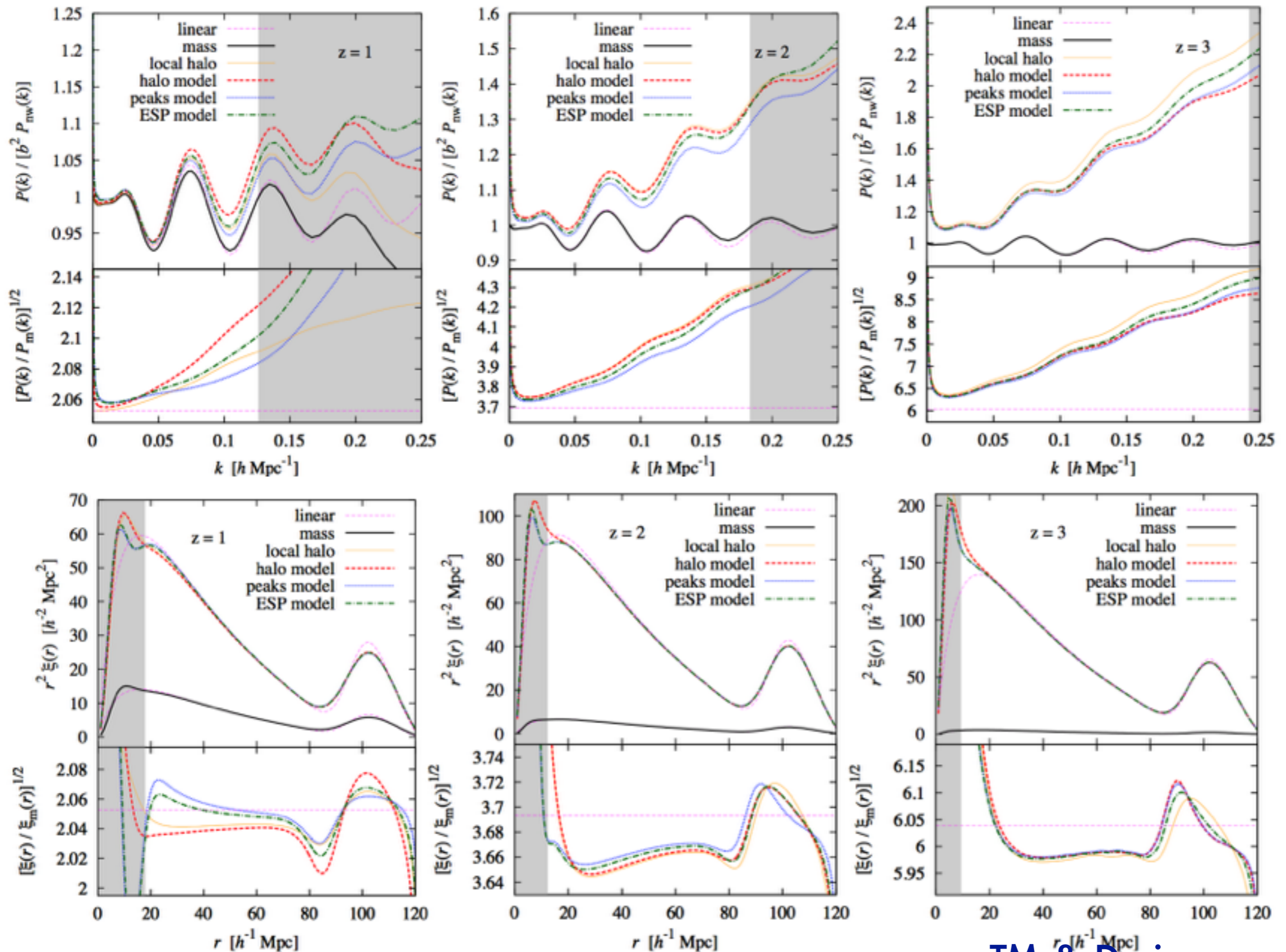
# Renormalized bias functions



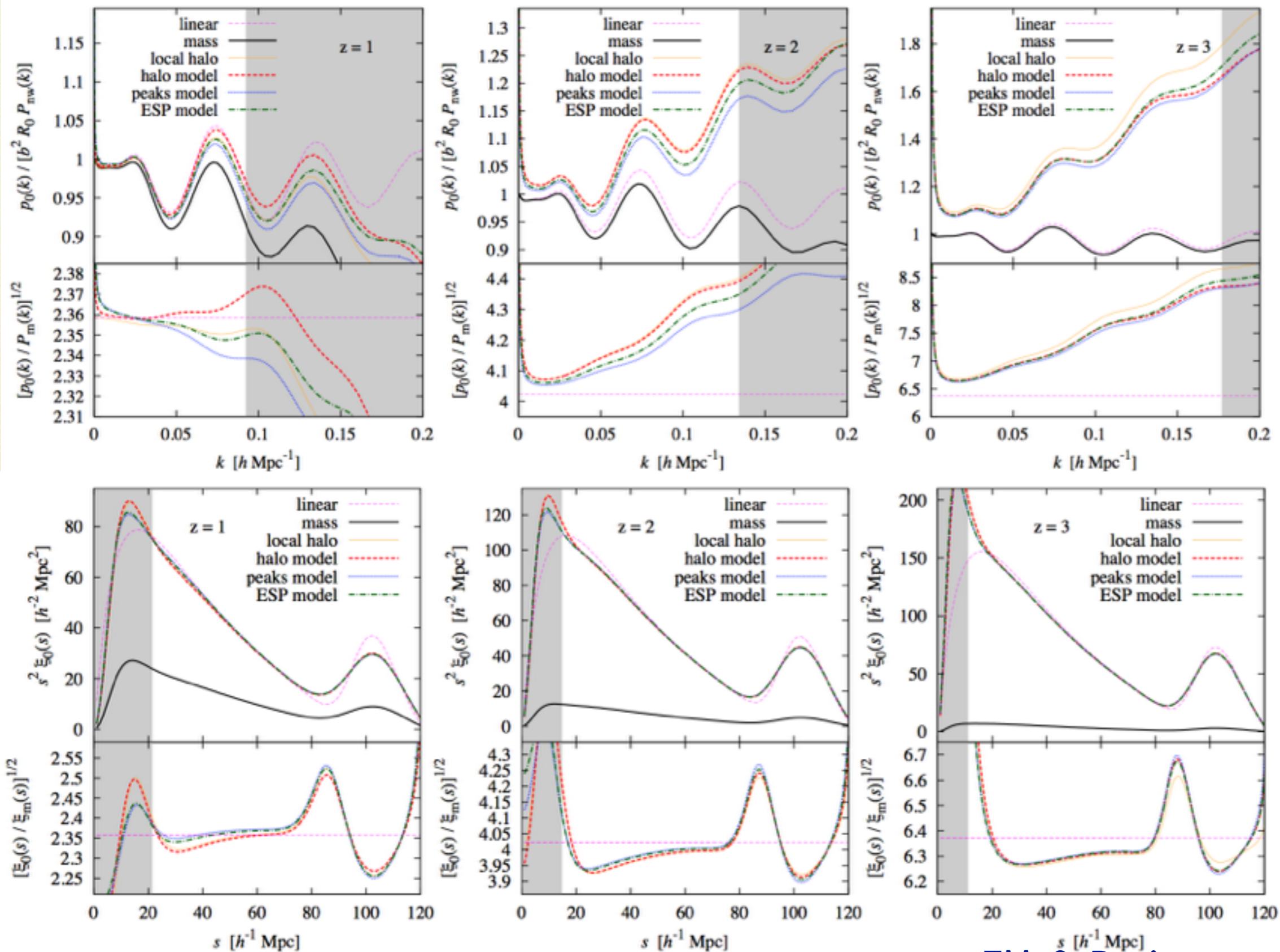
$z = 0.97$



# Power spectra & correlation functions


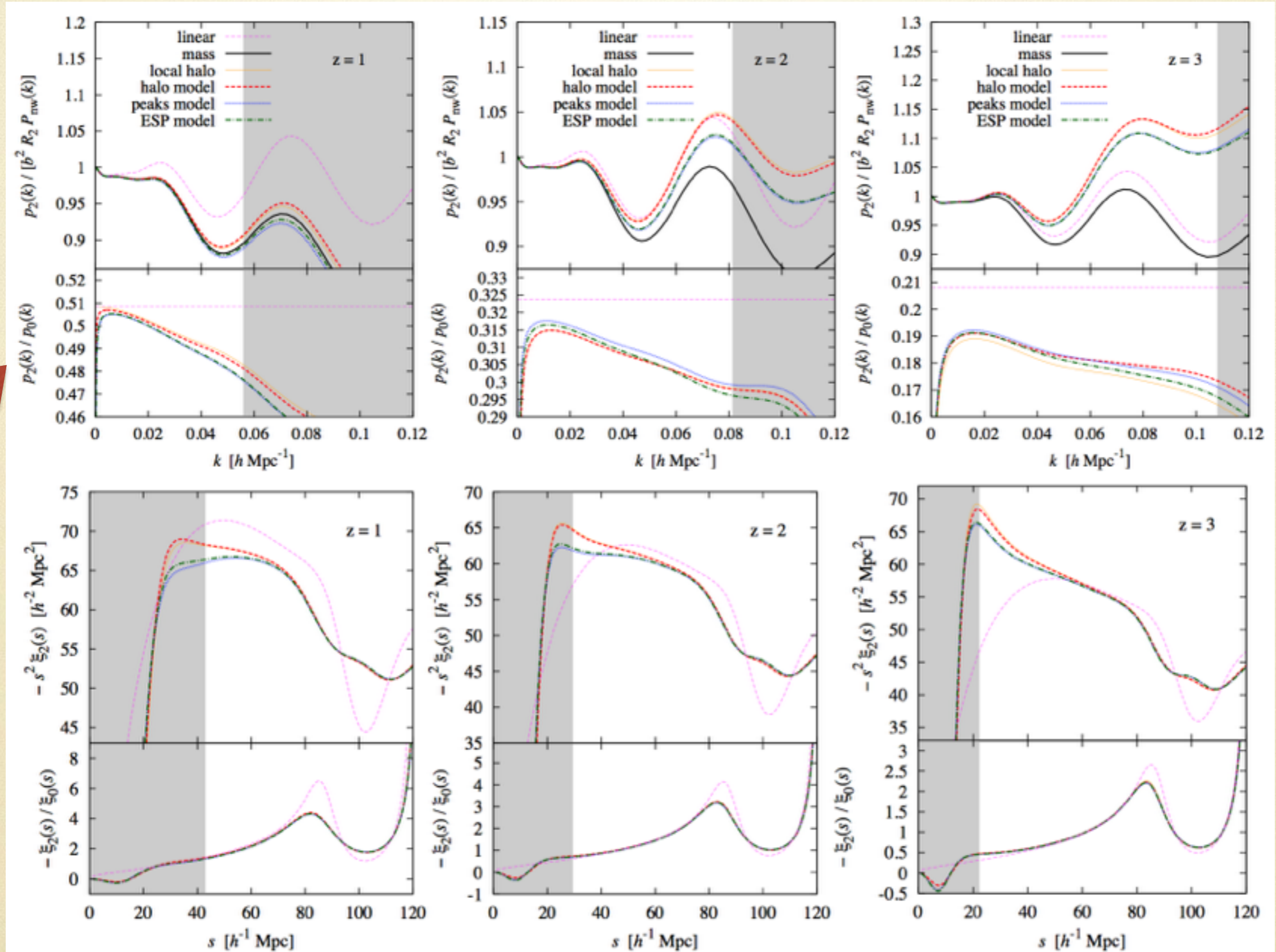


# Redshift space, monopole

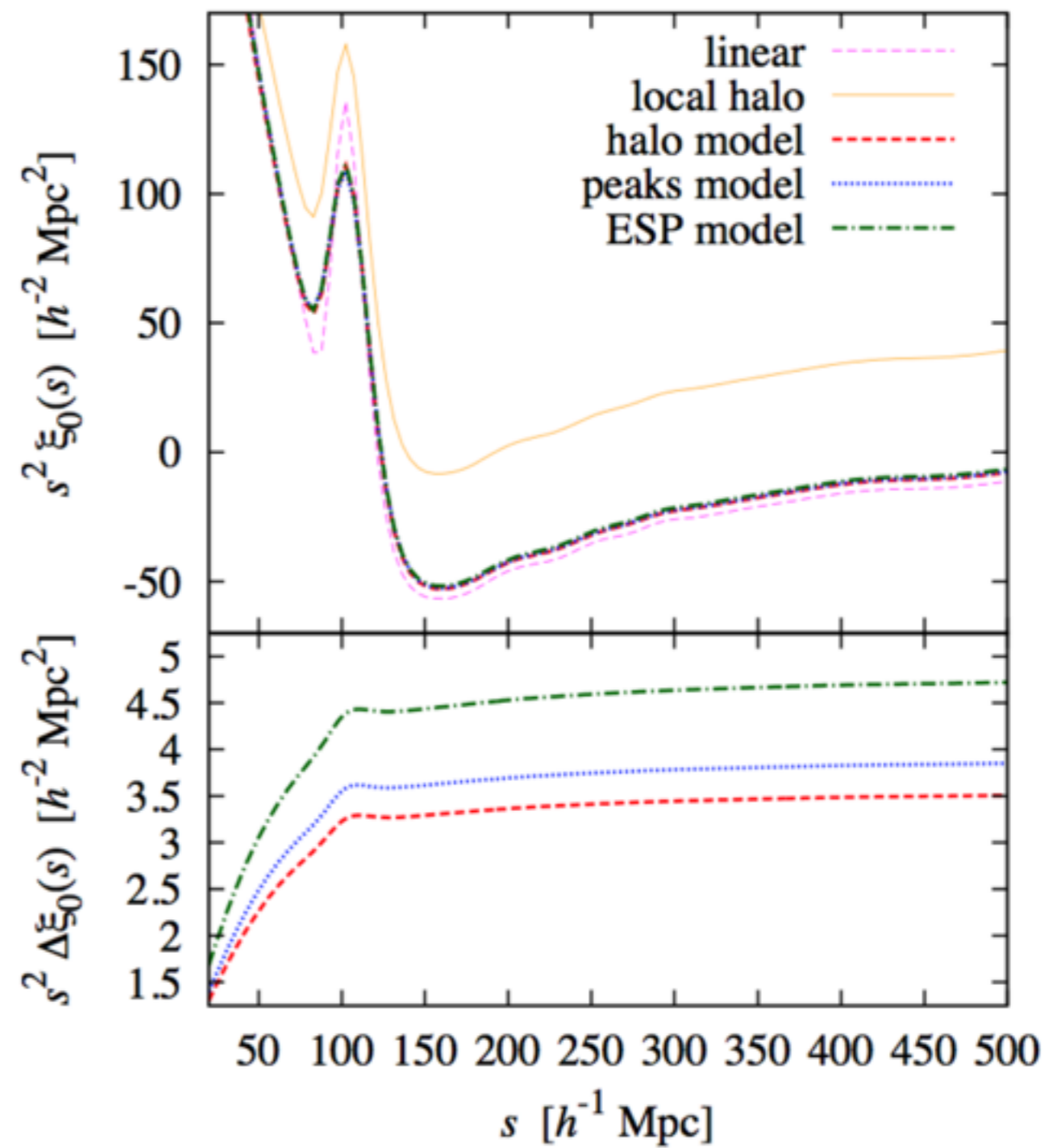
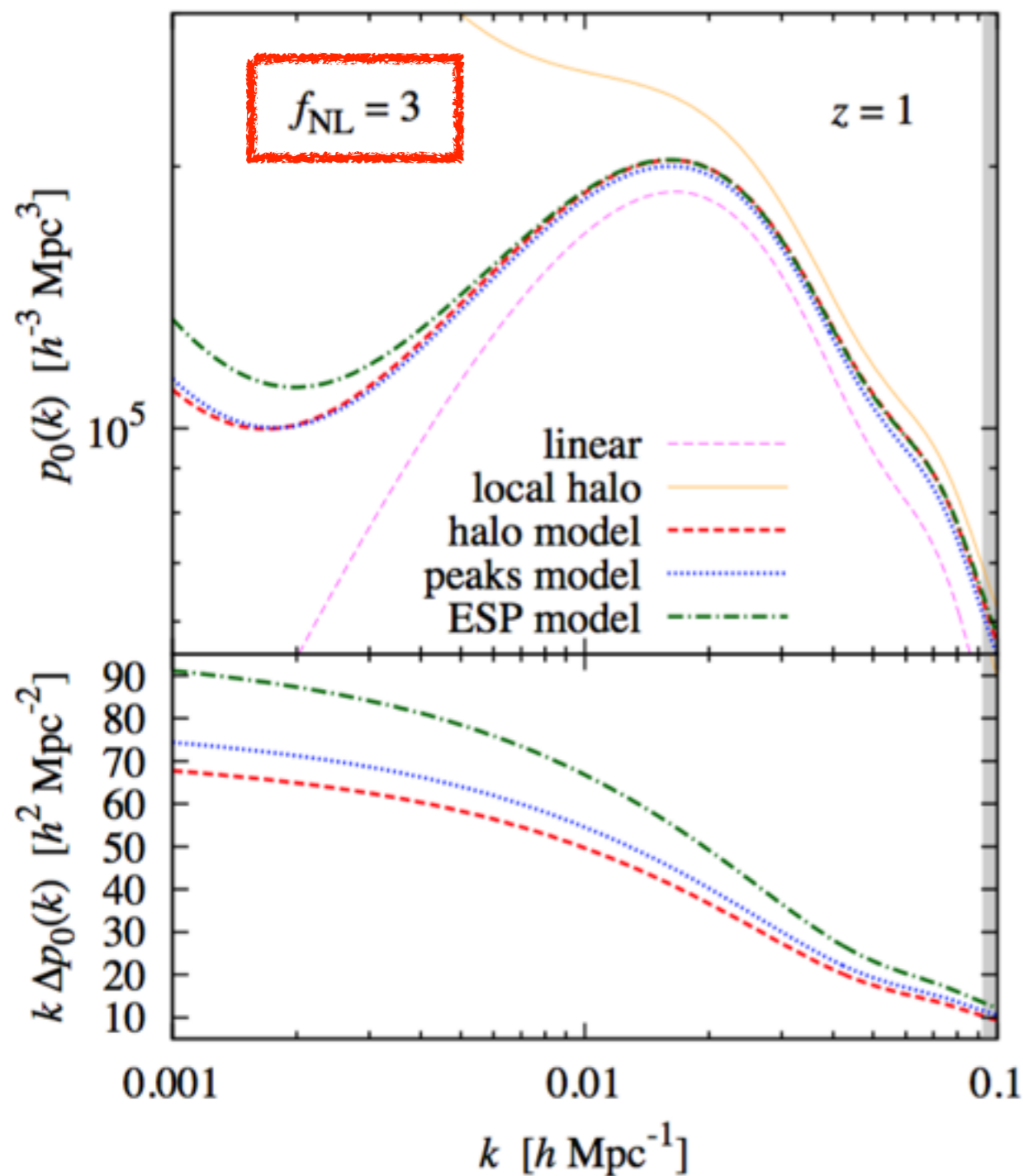


# Redshift-space distortions, quadrupole

$f\sigma_8$

# Primordial non-Gaussianity



# Summary

- Dependence on the bias models in weakly nonlinear regime
  - 2-4% for the power spectrum
  - $< 1\%$  for the correlation function
- Still important effects for the precision cosmology