

# CLASS-OneLoop

## Numerical aspects and outlook

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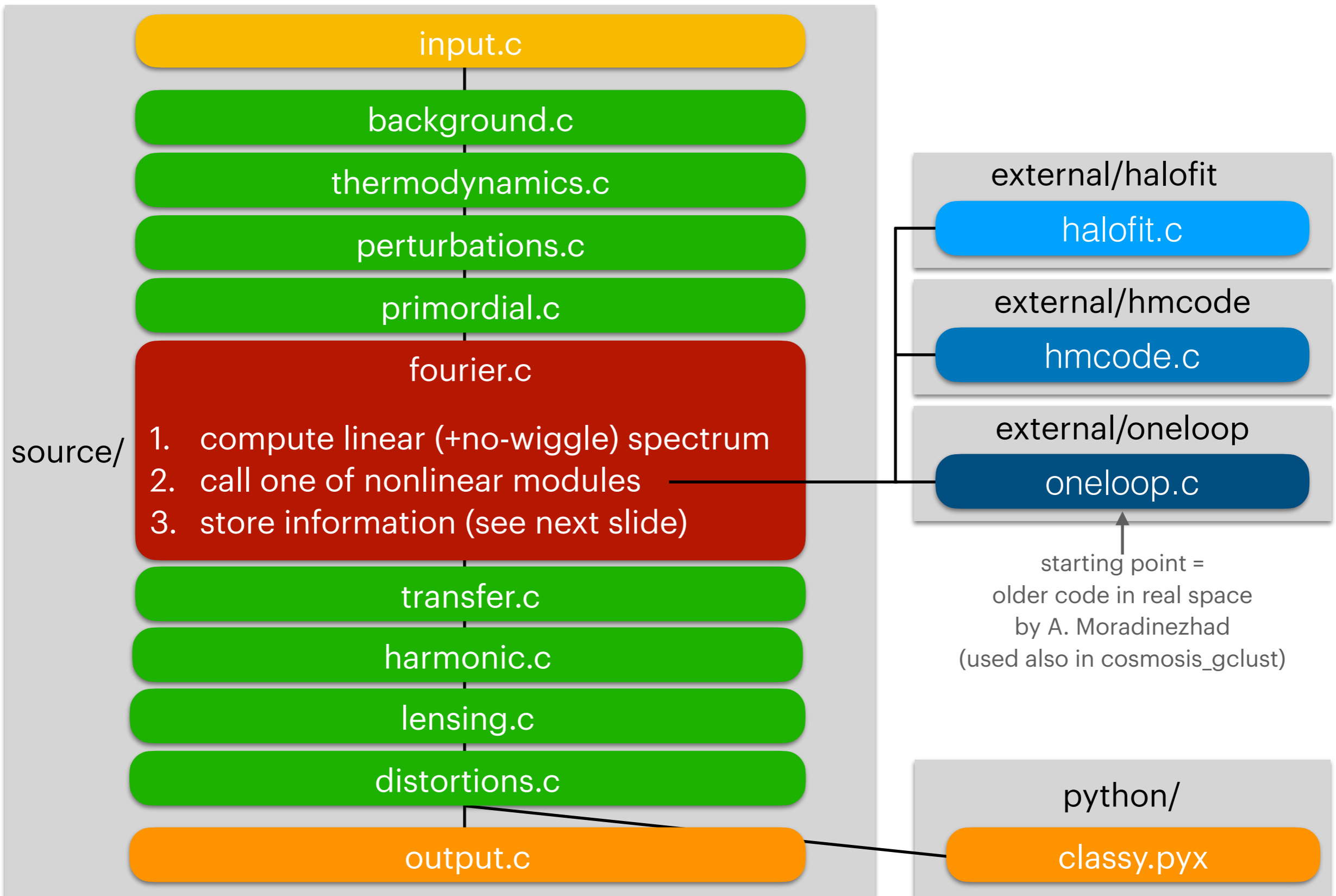


collab. with Dennis Linde, Azadeh Moradinezhad, Christian Radermacher, Santiago Casas,  
arXiv 2402.09778, accepted in JCAP

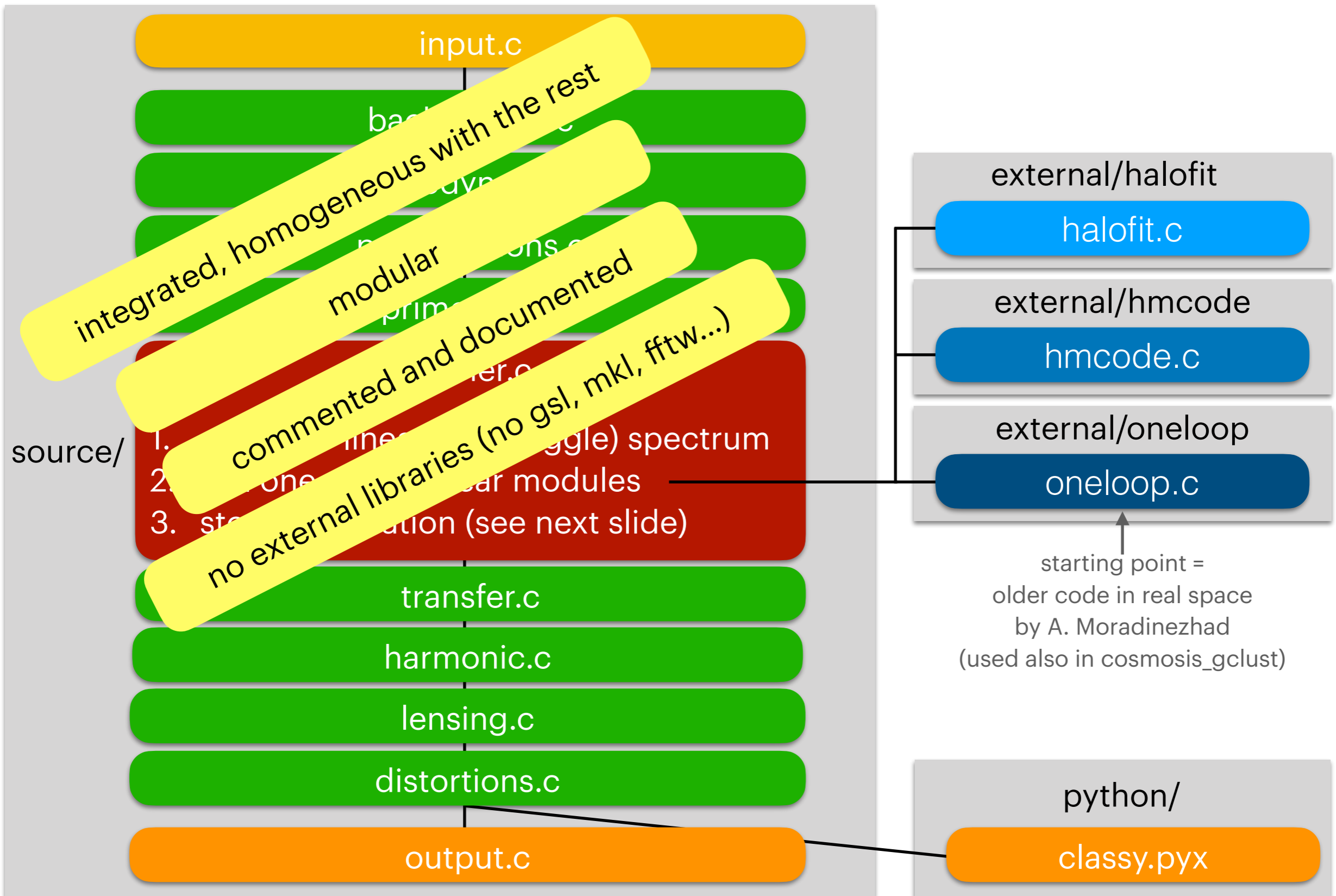
In preparation: release of the code in CLASS v3.4 + release paper with tests and documentation

Final version still under development, input/suggestions/requests welcome!

# Overall structure of CLASS v3.4



# Overall structure of CLASS v3.4



# What happens in the oneloop module ?

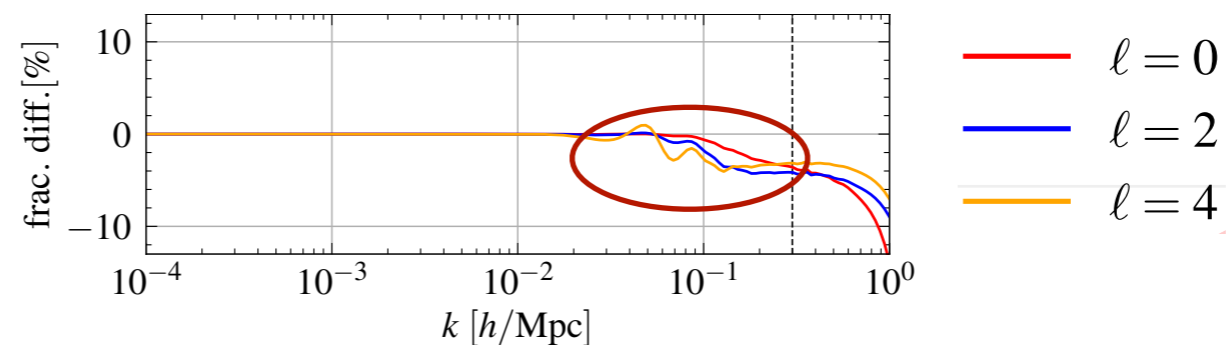
- ▶ log-Fourier approach with `oneloop_integration = log_fourier` (also: `direct_integration`)
  - ▶ if (cosmology-independent) kernels  $K_{ij}^n$  not found in binary files or cached in memory, compute them and write them, otherwise read them
  - ▶ log-Fourier transform linear spectrum  $P_{\text{lin}}$  into coefficients  $c_i$  at selected  $z_k$
  - ▶ compute 42 loops  $L^{(n)} = c_i K_{ij}^{(n)} c_j$  at these  $z_k$
- ▶ `oneloop_strategy = store_spectra` [stick to old CLASS logic]
  - ▶ uses input {biases} {counter-terms} {stoch. terms} to store  $P_X^{\text{real}}(k, z)$ ,  $P_X^{\text{rsd}}(k, z, \mu)$ ,  $P_{X, \ell=0,2,4}^{\text{rsd}}(k, z)$  for  $X \in \{\text{matter, tracer (e.g. galaxies), cross}\}$  at tabulated values
  - ▶ observables can be retrieved at any  $z$  within `output.c` or `classy.pyx` using interpolation
- ▶ `oneloop_strategy = store_loops` [fast/slow parameters in MCMC]
  - ▶ only store  $\mu$ -independent  $\{L^{(n)}\}$  at each  $z_k$
  - ▶ in `classy.pyx`, fast functions can build observables on demand for requested {bias} {counter-terms} {stoch. terms} {z} { $\mu$ }

can be just zero  $z = 0$ ,  
and then scaling with  $D(z, k)$ ,  $f(z, k)$ ,  
or several values

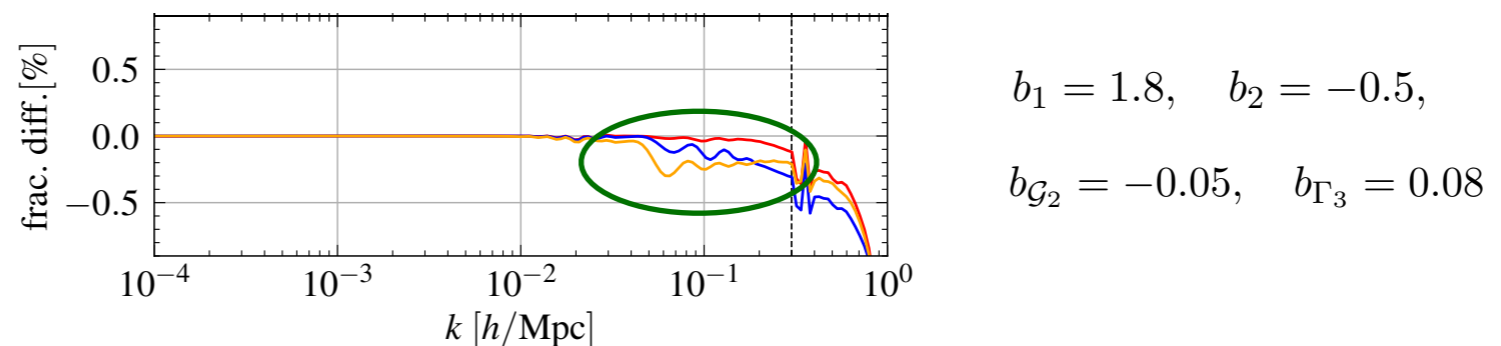


# What is the numerical accuracy of the oneloop module?

- ▶ Comparisons in  $k \in [0.01, 0.3] h/\text{Mpc}$  for  $N_{\text{FFT}} = 256$
- ▶ **Direct integration** vs. **FFTlog** (see 2402.09778 Fig 1):
  - ▶ Difference in each individual loop contribution  $< 0.1\%$
- ▶ **CLASS-OneLoop** vs. **CLASS-PT** [Chudaykin, Ivanov, Philcox, Simonovic 20] (see 2402.09778 Fig. 2,3,4):
  - ▶ Different de-wiggling algorithms (Gaussian filter vs. spectral decomposition):  
 $< 0.5\%$  in real space,  $< 5\%$  for multipoles; could be absorbed by counter-terms...



- ▶ Same de-wiggling algorithms (Gaussian filter):  $< 0.2\%$  in real space,  $< 0.3\%$  for multipoles



- ▶ Also consistent with **cosmosis\_gclust** [Moradinezhad et al.], **velocilaptors** [Chen et al. 20], **PyBird** [D'Amico et al. 20]

# How fast is the oneloop module?

- ▶ time flow [PRELIMINARY] on MacBookPro Intel i9 2.3GHz 16 cores:
  - ▶  $\Lambda$ CDM +  $m_\nu$ , no CMB, single  $\{z_k\} = 0$
  - ▶  $P_{lin}$  computed till  $k_{max} = 50 h/\text{Mpc}$  and extrapolated till  $k_{max} = 10^3 h/\text{Mpc}$  ( $\times 4$  for tracers)
  - ▶ Request: spectrum  $P_{tracer}^{rsd}(k, \mu, z)$  for array of 3  $z$ , 137  $k$ ,

slower version

Times in [s] , $N_c = \#$ of cores	FFTlog $N_{FFT} = 256$
kernels $\mathbf{K}$ (once per MCMC)	$73/N_c = 4.5$
log-Fourier transform $P_{lin}$ into $\mathbf{c}$ (once per cosmology and $z_k$ )	$0.005/N_c \sim 10^{-4}$
individual loops ( $\mathbf{L} = \mathbf{c K c}$ ) (once per cosmology and $z_k$ )	$2.1/N_c = 0.13$
build spectrum from loops (scale with # of output $z$ , here 3)	$0.006/N_c \sim 10^{-4}$
rest of CLASS with $N_c = 16$	0.5
total CLASS with $N_c = 16$ (cached kernels)	0.6

→ scales like  $N_{FFT} \ln(N_{FFT})$

→ scales like  $N_{FFT}^2$

# Can we do better?

- ▶ Yes, at least by revisiting the log-Fourier transform...

- ▶ logFT of  $P_{\text{lin}}$ :  $N_{\text{FFT}}$  coefficients  $c_j = \frac{1}{T} \int_{\ln(k_{\text{min}})}^{\ln(k_{\text{max}})} d\ln k P_{\text{lin}}(k) \exp \left[ \left( \frac{2\pi i j}{T} - \nu \right) \ln(k) \right]$

- ▶ `fourier_mode = fourier_mode_fft` (FFTlog)

- ▶ discrete Fast Fourier Transform with  $N_{\text{FFT}}$  values  $k_i$ , divide-and-conquer algorithm
- ▶ (implemented in C from scratch by N. Schöneberg for [arXiv:1807.09540](https://arxiv.org/abs/1807.09540) in `tools/fft.c`)
- ▶ decrease number of coefficients  $N_{\text{FFT}} \Rightarrow$  decrease their accuracy

- ▶ `fourier_mode = fourier_mode_spline` (SFTlog)

- ▶ spline  $P_{\text{lin}}(k_i) \rightarrow$  piece-wise cubic polynomial, moments  $P_i'' \rightarrow c_j = \sum_i \alpha_{i,j} P_i''$
- ▶ (implemented in C from scratch by C. Radermacher for this work in `tools/array.c`)
- ▶ works with  $P_{\text{lin}}(k_i)$  sampled at **non-evenly-spaced  $\ln k_i$**  (e.g. taken from previous modules and dense for BAO, sparse elsewhere)
- ▶ decrease number of coefficients  $N_{\text{FFT}}$  with constant accuracy

# Can we do better?

- ▶ time flow [PRELIMINARY] on MacBookPro Intel i9 2.3GHz 16 cores:
  - ▶  $\Lambda$ CDM +  $m_\nu$ , no CMB, single  $\{z_k\} = 0$
  - ▶  $P_{lin}$  computed till  $k_{max} = 50 h/\text{Mpc}$  and extrapolated till  $k_{max} = 10^3 h/\text{Mpc}$  ( $\times 4$  for tracers)
  - ▶ Request: spectrum  $P_{tracer}^{rsd}(k, \mu, z)$  for array of 3  $z$ , 137  $k$ , 5  $\mu$ 

accuracy stable at  $10^{-4}$  level

Times in [s] , $N_c = \#$ of cores	FFTlog $N_{FFT} = 256$	SFTlog $N_{FFT} = 96$ $N_k = 301$
kernels $\mathbf{K}$ (once per MCMC)	$73/N_c = 4.5$	$10/N_c = 0.6$
log-Fourier transform $P_{lin}$ into $\mathbf{c}$ (once per cosmology and $z_k$ )	$0.005/N_c \sim 10^{-4}$	$0.010/N_c \sim 5 \cdot 10^{-4}$
individual loops ( $\mathbf{L} = \mathbf{c} \mathbf{K} \mathbf{c}$ ) (once per cosmology and $z_k$ )	$2.1/N_c = 0.13$	$0.3/N_c = 0.02$
build spectrum from loops (scale with # of output $z$ , here 3)	$0.006/N_c \sim 10^{-4}$	$0.006/N_c \sim 10^{-4}$
rest of CLASS with $N_c = 16$	0.5	0.5
total CLASS with $N_c = 16$ (cached kernels)	0.6	0.5

# Conclusions

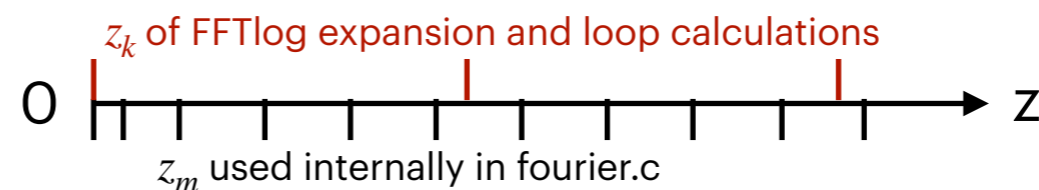
- ▶ `Class-OneLoop` already **fast enough** ( $\sim 20$  ms) unless  $P_{\text{lin}}$  calculation substituted by emulator...
- ▶ Before release of v3.4: need time to polish style (**user-friendliness**), set robust default precision parameters, provide **clear documentation** in release paper
- ▶ Possible developments -> panel discussion this evening



# Back-up slides

# Which redshifts are used in oneloop.c ?

- ▶ `oneloop_integration = log_fourier` (also: `direct_integration`, uses CUBA library)
  - ▶ if (cosmology-independent) kernels  $K_{ij}^n$  not found in binary files, compute them and write them, otherwise read them
  - ▶ log-Fourier transform linear spectrum coefficients  $c_i$  at selected  $z_k$
  - ▶ compute  $n = 1, \dots, 40$  loops  $L^n = c_i K_{ij}^n c_j$  at each  $z_k$



- ▶ `oneloop_redshift = single`

- ▶ only use  $z_k = 0$
- ▶ rescale to any other  $z$  using growth factor/rate of the model

- ▶ `oneloop_redshift = all`

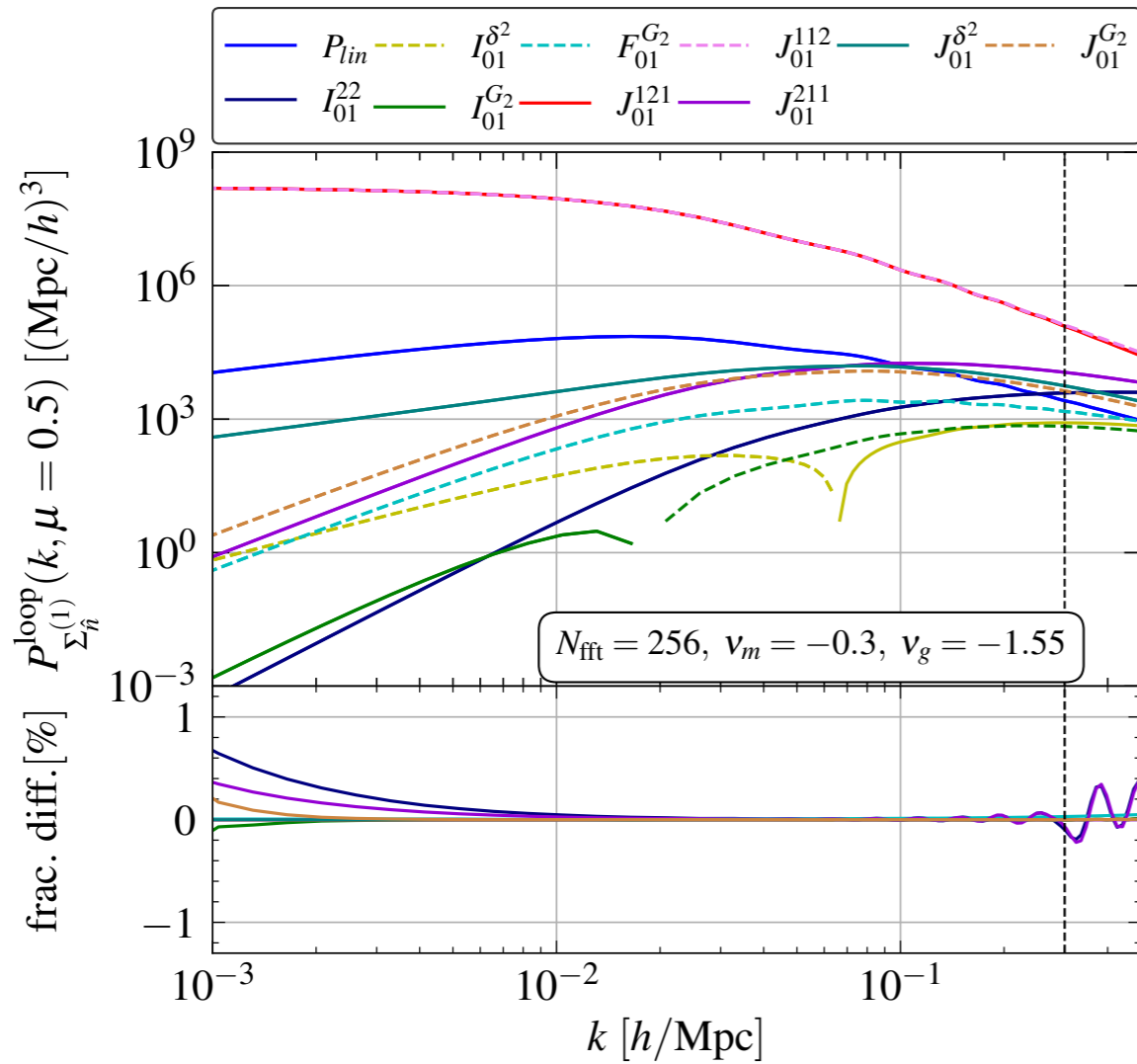
- ▶ expansion/loops at each  $z_m$  used internally by `fourier.c` (about a hundred)
- ▶ no rescaling, but still, kernels are  $z$ -independent...

- ▶ `oneloop_redshift = few`

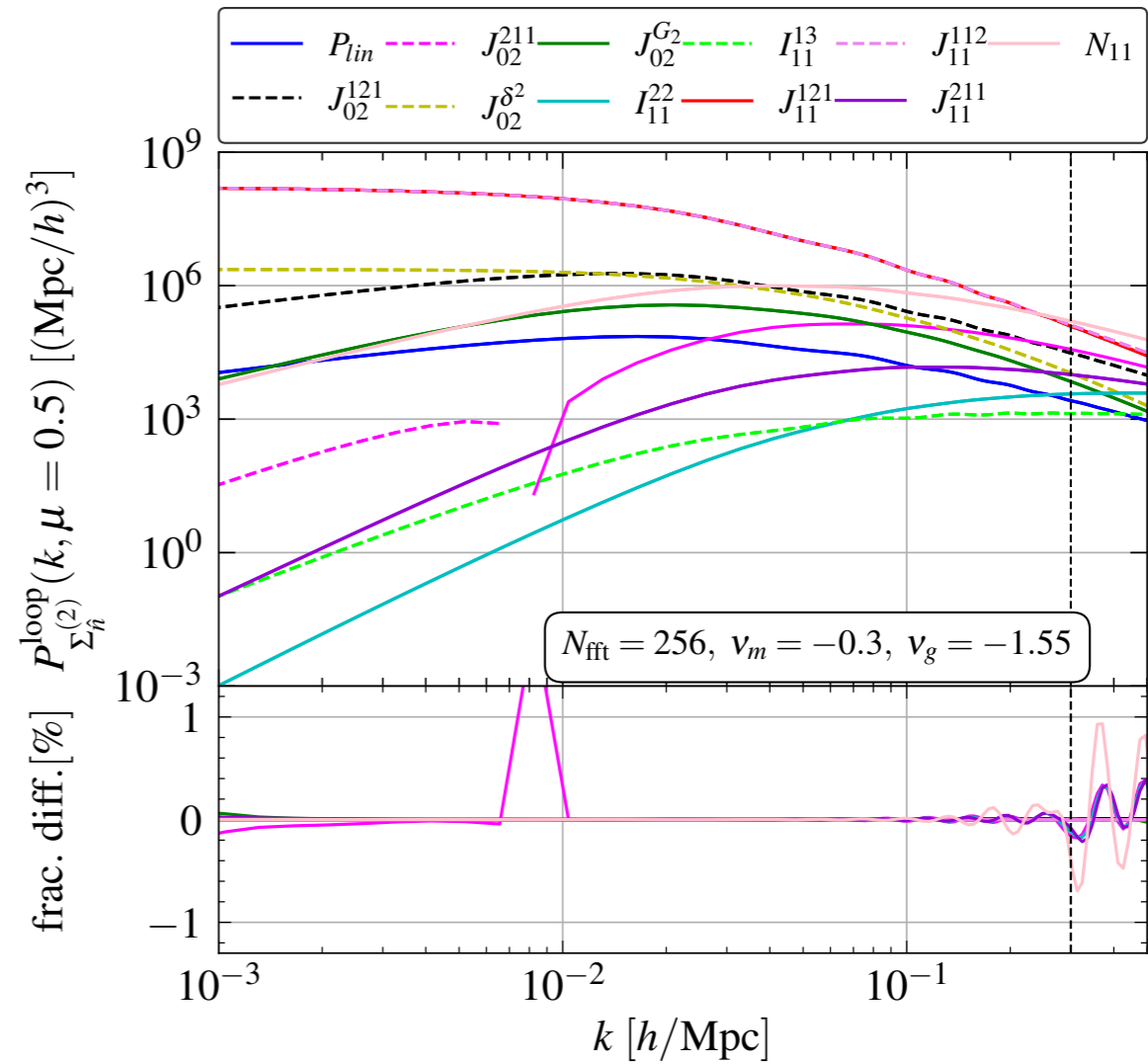
- ▶ expansion/loops at  $z_k$  passed by user with "`z_pk = .., , ..`"
- ▶ rescaling to closest  $z_k$  using growth factor/rate of the model

# Numerical accuracy of FFTlog/integration scheme?

- Direct integration versus FFTlog for loops with first / second velocity momenta



$0.01 < k [\text{Mpc}^{-1}h] < 0.3$

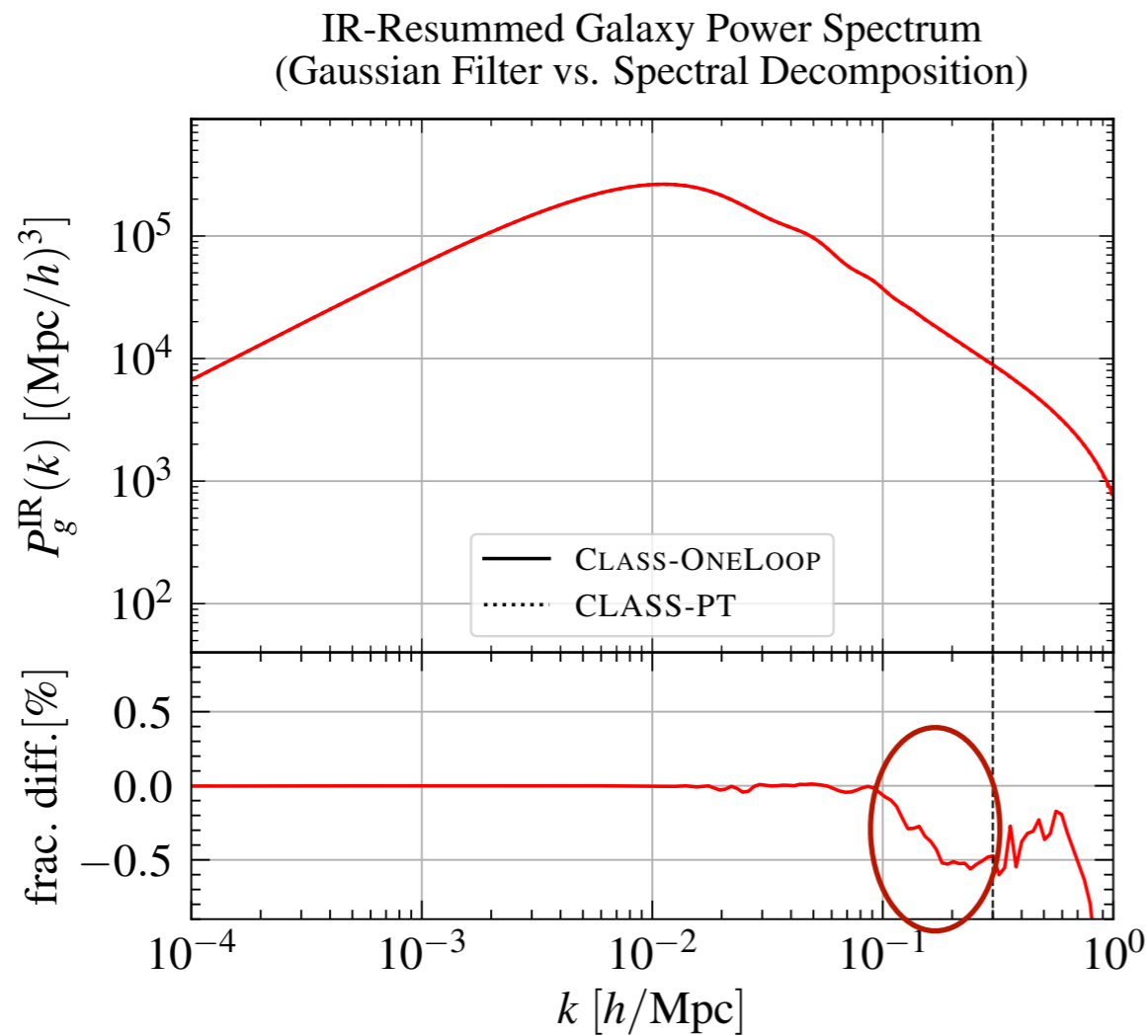


$0.01 < k [\text{Mpc}^{-1}h] < 0.3$

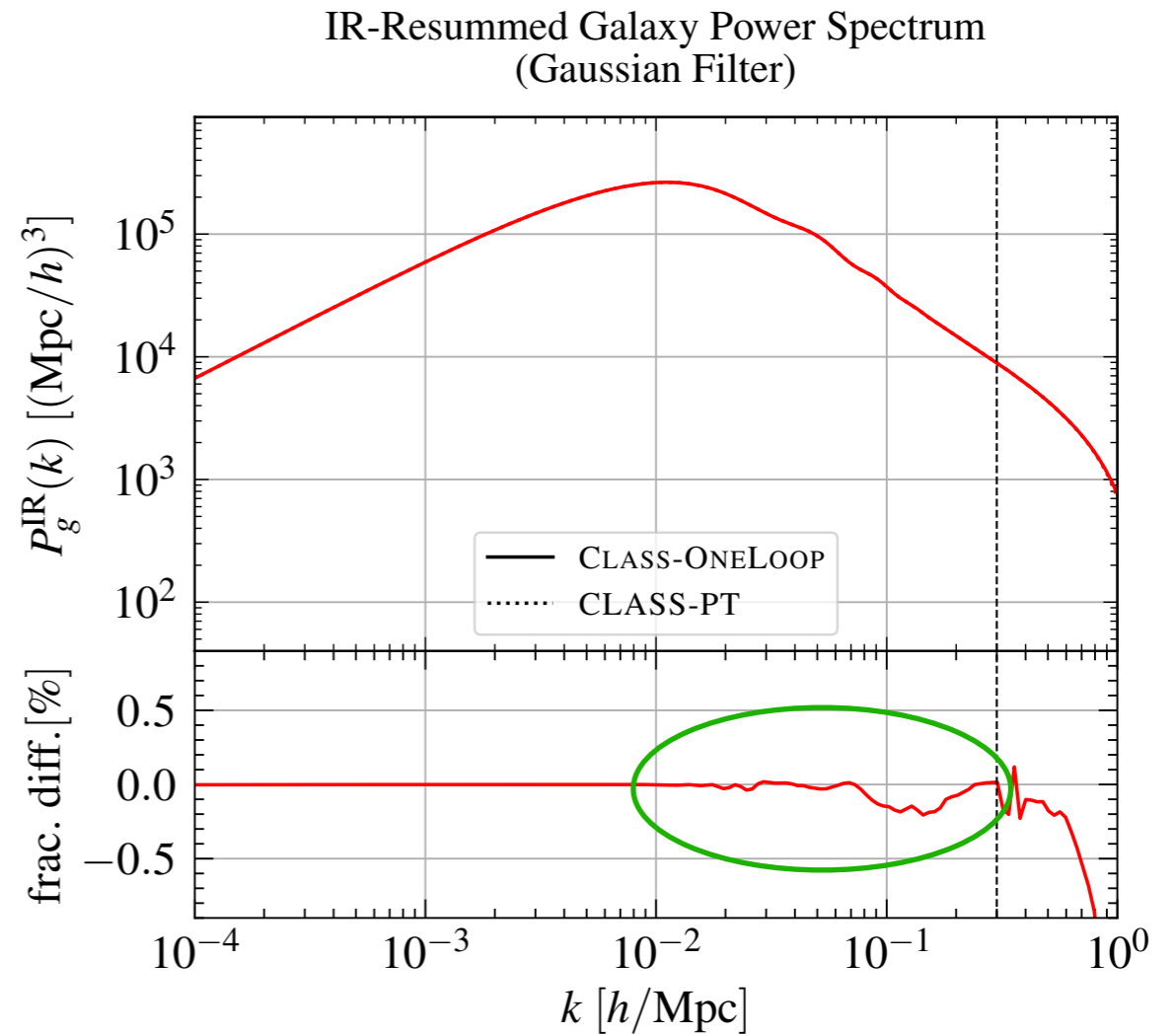
- Always < 0.1% difference in targeted k range for  $N_{\text{FFT}} = 256$

# Comparison with CLASS-PT?

- ▶ Comparison in real space at  $z = 0$ ,  $N_{\text{FFT}} = 256$ ,  $b_1 = 1.8$ ,  $b_2 = -0.5$ ,  $b_{g_2} = -0.05$ ,  $b_{\Gamma_3} = 0.08$
- ▶ Difference dominated by dewiggling method for IR resummation; absorbed by counter-terms?



←→  
 $0.01 < k [\text{Mpc}^{-1}h] < 0.3$

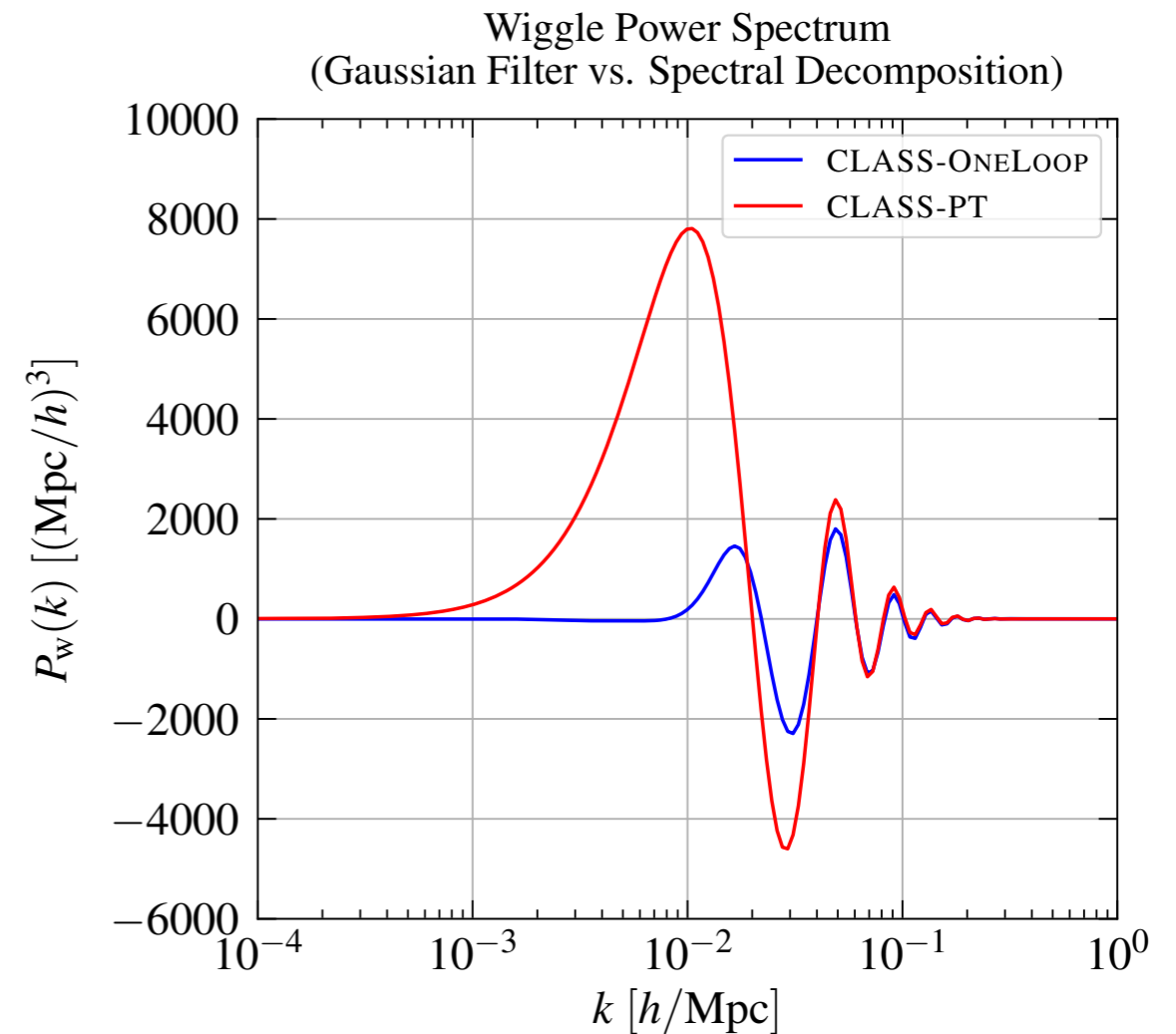
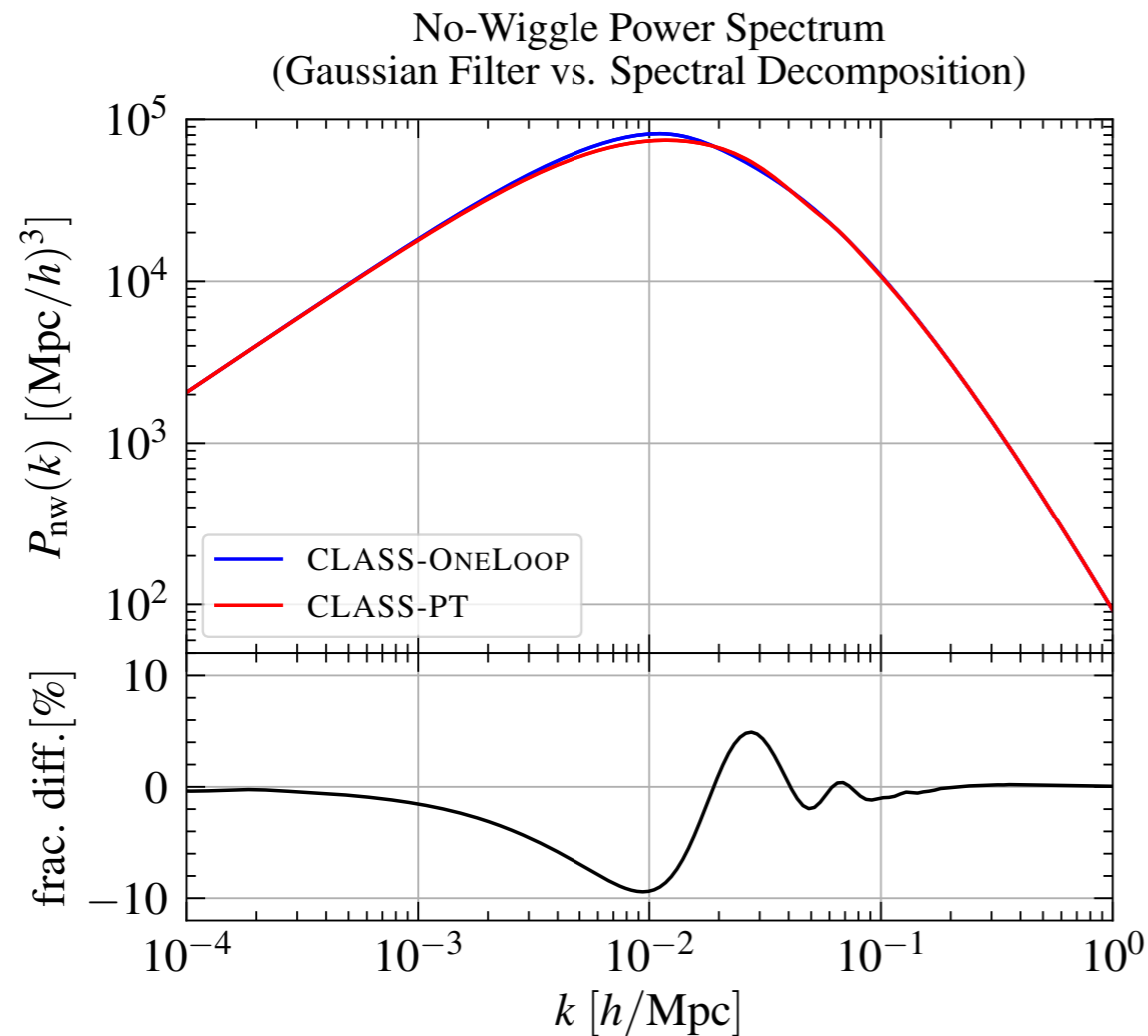


←→  
 $0.01 < k [\text{Mpc}^{-1}h] < 0.3$

- ▶  $< 0.2\%$  when sticking to same Gaussian filter

# Comparison with CLASS-PT?

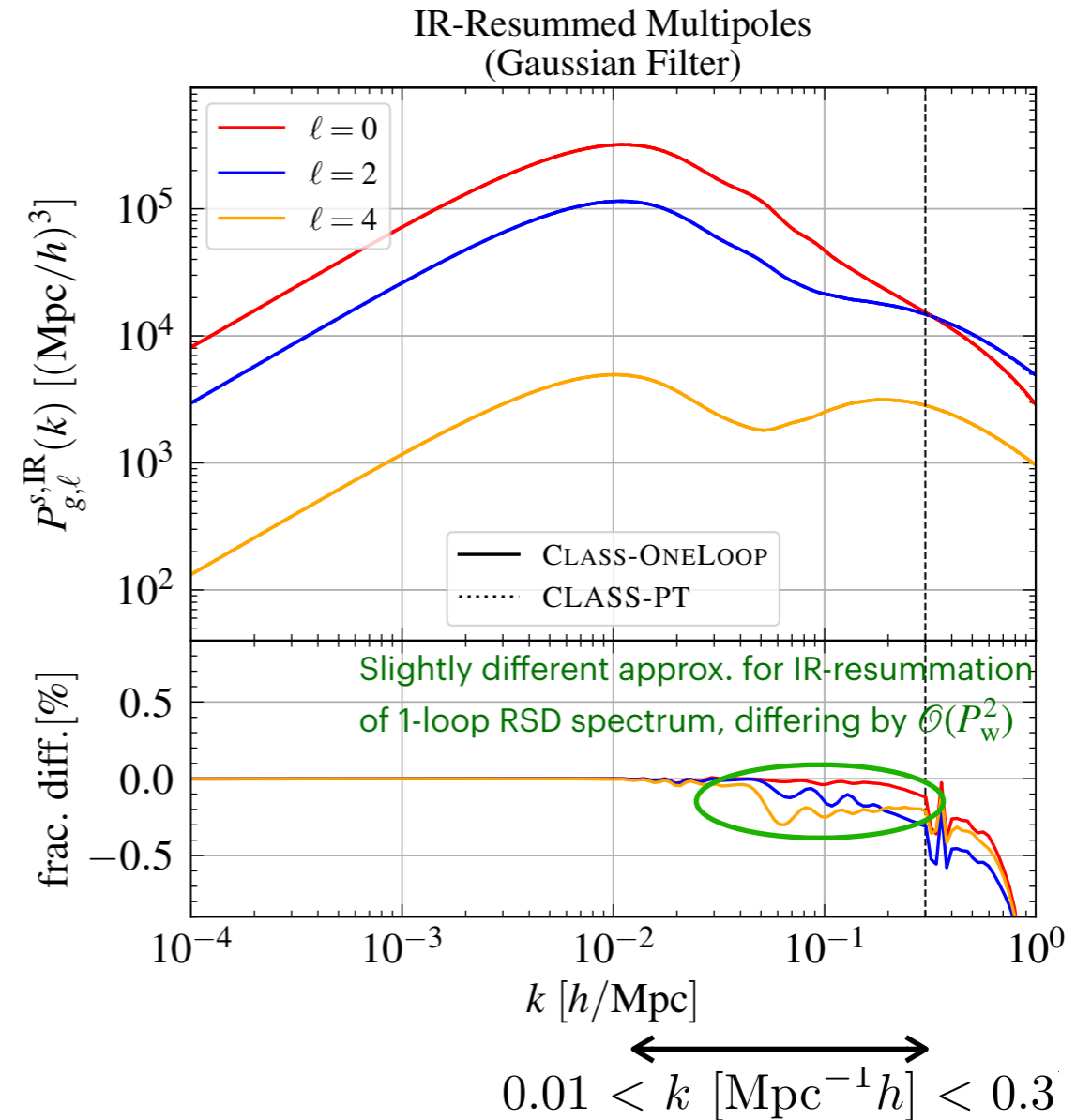
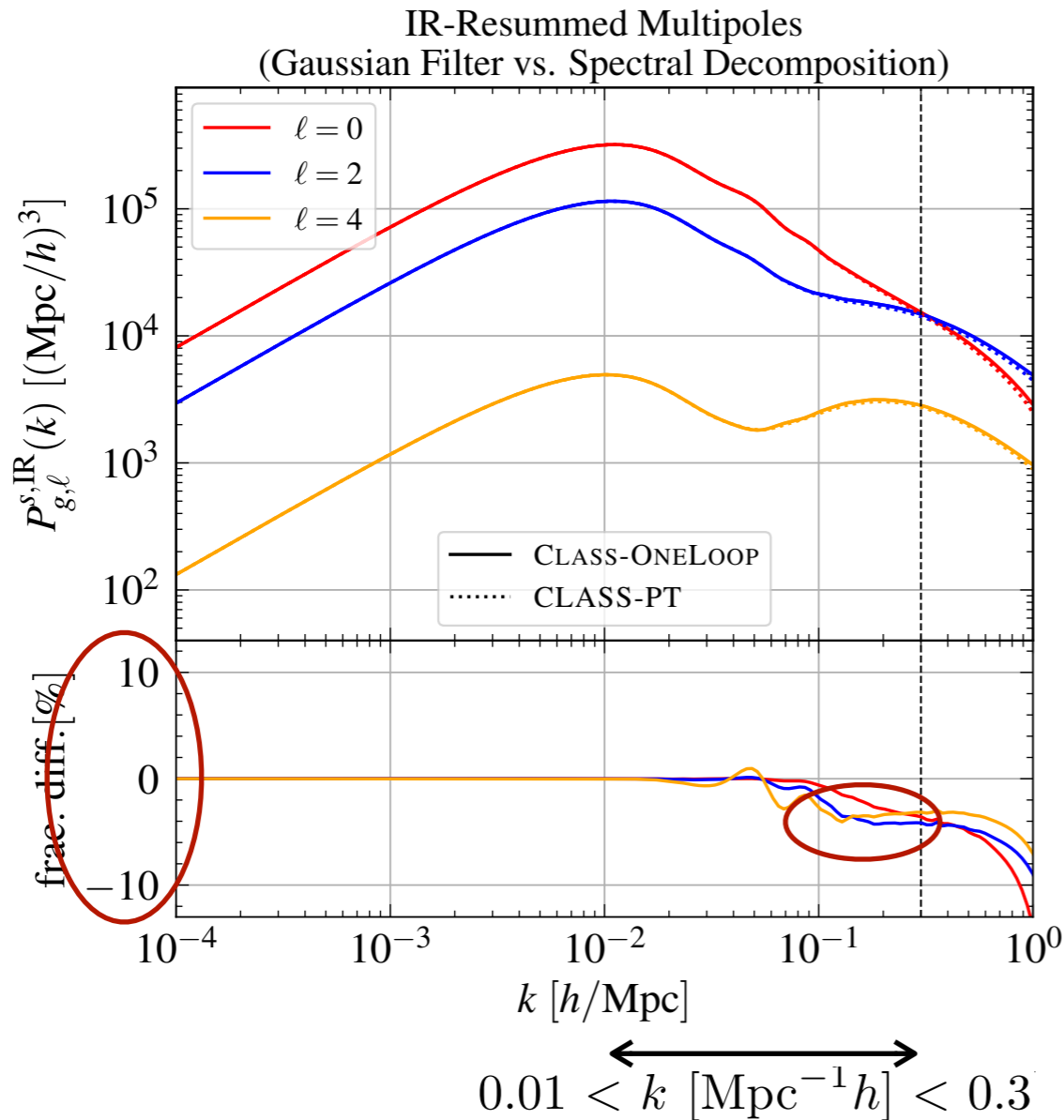
- ▶ Dewiggling method for IR resummation:
- ▶ Spectral decomposition (DST) [Hamann et al. 2010] vs. Gaussian filtering (of  $P_{\text{lin}}/P_{\text{HE}}$ )





# Comparison with CLASS-PT?

- ▶ Comp. in redshift space at  $z = 0$ ,  $N_{\text{FFT}} = 256$ ,  $b_1 = 1.8$ ,  $b_2 = -0.5$ ,  $b_{G_2} = -0.05$ ,  $b_{\Gamma_3} = 0.08$
- ▶ Difference dominated by dewiggling method for IR resummation; absorbed by counter-terms?



- ▶  $< 0.3\%$  when sticking to same Gaussian filter

# Performance?

- ▶ FFTlog expansion domain:  $k_{\min} = 10^{-6} h/\text{Mpc}$ ,  $k_{\max} = 10^3 h/\text{Mpc}$  ( $\times 4$  for tracers) (avoid ringing)
- ▶ use  $P_{\text{lin}}(k)$  extrapolation for  $k \in [50, 10^3] h/\text{Mpc}$  without impact on  $P_{\text{oneloop}}(k < 1 h/\text{Mpc})$
- ▶ on RWTH Aachen cluster node: 2 Intel Xeon Platinum 8160 (24 cores each), 192GB RAM
- ▶ [preliminary from paper I]: already improved, can still gain more

- ▶ if kernels are not cached:

- ▶ computation of kernels  $K_{ij}^n$  + spectrum coefficients  $c_i$  + loops  $L^n = c_i K_{ij}^n c_j$  at one  $z$ :

	$N_{\text{FFT}} = 128$	$N_{\text{FFT}} = 256$	$N_{\text{FFT}} = 512$	Direct integration
in seconds:				
4 threads	$0.61 \pm 0.26$	$2.05 \pm 0.15$	$6.98 \pm 0.14$	$\sim 600$
8 threads	$0.40 \pm 0.09$	$1.38 \pm 0.12$	$3.51 \pm 0.23$	-
16 threads	$0.52 \pm 0.11$	$1.13 \pm 0.13$	$2.00 \pm 0.22$	-

- ▶ If kernels are cached:

- ▶ computation of spectrum coefficients  $c_i$  + loops  $L^n = c_i K_{ij}^n c_j$  at one  $z$ :

	$N_{\text{FFT}} = 128$	$N_{\text{FFT}} = 256$	$N_{\text{FFT}} = 512$
in seconds:			
4 threads	$0.101 \pm 0.008$	$0.400 \pm 0.003$	$1.467 \pm 0.085$
8 threads	$0.046 \pm 0.004$	$0.212 \pm 0.018$	$0.776 \pm 0.037$
16 threads	$0.028 \pm 0.003$	$0.105 \pm 0.003$	$0.382 \pm 0.0$

- ▶ slightly smaller than to rest of CLASS!

# Which log-Fourier Transform algorithms?

- ▶ No external libraries (only optionally CUBA if direct integration required)

- ▶ logFT of  $P_{\text{lin}}$  :  $c_n = \frac{1}{T} \int_{\ln(k_{\text{min}})}^{\ln(k_{\text{max}})} d \ln k \ l$

- ▶ `fourier_mode = fourier_mode_fft`

- ▶ discrete Fast Fourier Transform
- ▶ scales like  $N_{\text{FFT}} \ln(N_{\text{FFT}})$ , but t
- ▶ implemented in c from scratch

- ▶ `fourier_mode = fourier_mode_spline`

- ▶  $P_{\text{lin}}(k_i)$  always splined for inter
- ▶  $c_n$  = sum of piece-wise analytic
- ▶ works with  $P_{\text{lin}}(k_i)$  sampled at r

- ▶ scales like  $N_{\text{FFT}} \times N_k$ , but then loop calculation still scales like  $N_{\text{FFT}}^2$
- ▶ decorrelates  $N_{\text{FFT}}$  from  $N_k$  samples.  $N_{\text{FFT}}$  can be reduced without degrading  $c_n$  precision.
- ▶ implemented in c from scratch by C. Radermacher for this work in tools/array.c

