LECTURE 12: THE MEASUREMENT OF POWER SPECTRUM AND CORRELATION FUNCTION

GONG-BO ZHAO

1. The FKP weight

The FKP weight is the optimal weight for measurement of power spectrum multipoles [1].

(1)
$$P(k) = P(\mathbf{k}) \equiv \int d^3r \xi(\mathbf{r}) e^{i\mathbf{k}\cdot\mathbf{r}}$$

(2)
$$F(\mathbf{r}) \equiv \frac{w(\mathbf{r}) \left[n_g(\mathbf{r}) - \alpha n_s(\mathbf{r}) \right]}{\left[\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r}) \right]^{1/2}}$$

$$\langle |F(\mathbf{k})|^2 \rangle = \frac{\int d^3r \int d^3r' w(\mathbf{r}) w(\mathbf{r}') \langle [n_g(\mathbf{r}) - \alpha n_s(\mathbf{r})] [n_g(\mathbf{r}') - \alpha n_s(\mathbf{r}')] \rangle e^{i\mathbf{k}\cdot(\mathbf{r}-\mathbf{r}')}}{\int d^3r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})}$$

(3)
$$\langle n_g(\mathbf{r}) n_g(\mathbf{r}') \rangle = \overline{n}(\mathbf{r}) \overline{n} (\mathbf{r}') \left[1 + \xi (\mathbf{r} - \mathbf{r}') \right] + \overline{n}(\mathbf{r}) \delta (\mathbf{r} - \mathbf{r}')$$

$$\langle n_s(\mathbf{r}) n_s(\mathbf{r}') \rangle = \alpha^{-2} \overline{n}(\mathbf{r}) \overline{n} (\mathbf{r}') + \alpha^{-1} \overline{n}(\mathbf{r}) \delta (\mathbf{r} - \mathbf{r}')$$

$$\langle n_g(\mathbf{r}) n_s(\mathbf{r}') \rangle = \alpha^{-1} \overline{n}(\mathbf{r}) \overline{n} (\mathbf{r}')$$

(4)
$$\langle |F(\mathbf{k})|^2 \rangle = \int \frac{d^3k'}{(2\pi)^3} P(\mathbf{k'}) \left| G(\mathbf{k} - \mathbf{k'}) \right|^2 + (1+\alpha) \frac{\int d^3r \overline{n}(\mathbf{r}) w^2(\mathbf{r})}{\int d^3r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})}$$

(5)
$$G(\mathbf{k}) \equiv \frac{\int d^3 r \overline{n}(\mathbf{r}) w(\mathbf{r}) e^{i\mathbf{k} \cdot \mathbf{r}}}{\left[\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})\right]^{1/2}}$$

(6)
$$\langle |F(\mathbf{k})|^2 \rangle \simeq P(\mathbf{k}) + P_{\text{shot}}$$

(7)
$$P_{\text{shot}} \equiv \frac{(1+\alpha) \int d^3 r \overline{n}(\mathbf{r}) w^2(\mathbf{r})}{\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})}$$

(8)
$$\hat{P}(\mathbf{k}) = |F(\mathbf{k})|^2 - P_{\text{shot}}$$

Date: April 19, 2019.

(9)
$$\hat{P}(k) \equiv \frac{1}{V_k} \int_{V_k} d^3 k' \hat{P}\left(\mathbf{k'}\right)$$

(10)
$$\sigma_P^2(k) \simeq \frac{1}{V_L} \int d^3k' \left| P(k)Q\left(\mathbf{k'}\right) + S\left(\mathbf{k'}\right) \right|^2$$

where

(11)
$$Q(\mathbf{k}) \equiv \frac{\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r}) e^{i\mathbf{k}\cdot\mathbf{r}}}{\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})}$$

and

(12)
$$S(\mathbf{k}) \equiv \frac{(1+\alpha) \int d^3 r \overline{n}(\mathbf{r}) w^2(\mathbf{r}) e^{i\mathbf{k}\cdot\mathbf{r}}}{\int d^3 r \overline{n}^2(\mathbf{r}) w^2(\mathbf{r})}$$

(13)
$$\frac{\sigma_P^2(k)}{P^2(k)} = \frac{(2\pi)^3 \int d^3 r \overline{n}^4 w^4 [1 + 1/\overline{n}P(k)]^2}{V_k \left[\int d^3 r \overline{n}^2 w^2 \right]^2}$$

The KFP weight taking the following form minimises Eq (13):

(14)
$$w_0(\mathbf{r}) = \frac{1}{1 + \overline{n}(\mathbf{r})P(k)}$$

2. The measurement of the anisotropic correlation function

Pair counts in 'data' are compared to pair counts in 'random' samples that follow the geometry of the survey. We assume a catalogue of n_d objects in the data sample and n_r in the random sample and then calculate three sets of numbers of pairs as a function of the binned comoving separation s,

• Data-data pairs dd(s) that normalized by the total number of data pairs:

(15)
$$DD(s) = \frac{dd(s)}{n_d (n_d - 1)/2};$$

• Random-Random pairs rr(s) that normalized by the total number of random pairs:

(16)
$$RR(s) = \frac{rr(s)}{n_r(n_r - 1)/2};$$

• Data-Random pairs dr(s) that normalized by the total number of cross pairs:

(17)
$$DR(s) = \frac{dr(s)}{n_r n_d}.$$

The most commonly used Landy & Szalay estimator [4]:

(18)
$$\hat{\xi}(s) = \frac{DD - 2DR + RR}{RR}.$$

The task now is to compute the pair counts quickly for a large sample, which is challenging. For example, for BOSS DR12 sample, there are $\sim 1 \mathrm{M}$ data samples, and $\sim 100 \mathrm{M}$

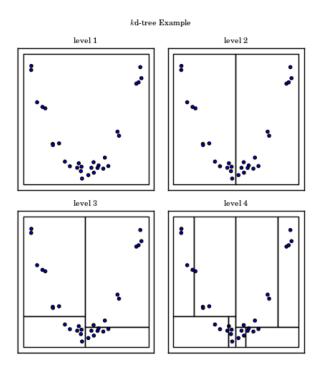


FIGURE 1. An example of kdtree.

random samples! So this is mission impossible for a brutal force algorithm with $O(N^2)$ complexity.

The way out is to use the kd-tree algorithm ¹.

The kdtree code for the measurement: https://github.com/wll745881210/KDTPCF

¹ http://www.linuxclustersinstitute.org/conferences/archive/2008/PDF/Dolence_98279.pdf

If node1 and node2 have already been compared in reverse

function dualtree(node1, node2) {

return

```
Let dmin = minimum distance between node1 and node2
If dmin is greater than maximum distance of interest
  return
Let dmax = maximum distance between node1 and node2
Let binmin = distance bin for distance dmin
Let binmax = distance bin for distance dmax
If binmin = binmax
  Add node1.cnt x node2.cnt to bin count
Else if node1 and node2 are leaf nodes
  For all points i owned by node1
    Let smin = minimum distance between point i and node2
    Let smax = maximum distance between point i and node2
     Let binmin = distance bin for smin
     Let binmax = distance bin for distance smax
      If binmin = binmax
        Add node2.cnt to bin count
     Else
        Find distances to all points in node2 from point i
        Find distance bins and update the bin counts
 Else
    If node1.cnt > node2.cnt
      dualtree(node1.leftChild, node2)
      dualtree(node1.rightChild, node2)
      dualtree(node1, node2.leftChild)
      dualtree(node1, node2.rightChild)
}
```

FIGURE 2. The pseudo-code of kdtree.

References

- [1] H. A. Feldman, N. Kaiser and J. A. Peacock, Astrophys. J. 426, 23 (1994) doi:10.1086/174036 [astroph/9304022].
- [2] D. Bianchi, H. Gil-Marn, R. Ruggeri and W. J. Percival, Mon. Not. Roy. Astron. Soc. 453, no. 1, L11 (2015) doi:10.1093/mnrasl/slv090 [arXiv:1505.05341 [astro-ph.CO]].
- [3] N. Hand, Y. Li, Z. Slepian and U. Seljak, JCAP 1707, no. 07, 002 (2017) doi:10.1088/1475-7516/2017/07/002 [arXiv:1704.02357 [astro-ph.CO]].
- [4] S. D. Landy and A. S. Szalay, Astrophys. J. **412**, 64 (1993). doi:10.1086/172900