

Enhancing Hi-C contact map resolution with neural network

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1 Introduction

Recently, the high-throughput chromosome conformation capture (Hi-C) technique has become a powerful tool for studying the three-dimensional structure of chromosomes. Hi-C data is usually expressed as an $n \times n$ matrix. The resolution of Hi-C data is defined as the bin size of each cell of the matrix. Hi-C data at kilobase level are requisite for future genome 3D structure studies. Rao et al.(2014) generated Hi-C data with 1 kilobase resolution. However, millions of sequenced reads are required to archive this resolution with a huge amount of money and time consumption.

Zhang et al. presented a approach to enhance the resolution of Hi-C data called HiCPlus. HiPlus generated low-resolution data by down-sampling the number of sequenced reads and then a neural network was used to create the mapping between high-resolution contact map and low-resolution contact map.

[1]

2 Methods

Let D be a set of paired ends reads of a Hi-C experiment. We generate low and high contact maps from D , denoted by M_ℓ and M_h . Let S_ℓ and S_h be the size of M_ℓ and M_h . Let R be the ratio of S_ℓ to S_h , which can be represented by $R = \frac{S_\ell}{S_h}$. And P be the number of overlapping pixels between adjacent sub-maps. Let K be the size of low-resolution sub-maps.

% Divide matrices M_ℓ and M_h

To M_ℓ :

for $i = 1, 1+K - P, 1+2 \times (K - P), \dots$

 for $j = 1, 1+K - P, 1+2 \times (K - P), \dots$

 IF $i + K > M_\ell$ || $j + K > M_\ell$, BREAK

 ELSE extract $K \times K$ sub-maps whose left-top coordinate is (i, j) from M_ℓ .

Do the same process to M_h :

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for  $i = 1, 1+(K-P) \times R, 1+2 \times (K-P) \times R, \dots$ 
  for  $j = 1, 1+(K-P) \times R, 1+2 \times (K-P) \times R, \dots$ 
    IF  $i + K \times R > M_h$  ||  $j + K \times R > M_h$ , BREAK
    ELSE extract  $(K \times R) \times (K \times R)$  sub-maps whose left-top
      coordinate is  $(i, j)$  from  $M_h$ .

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Let C_ℓ and C_h be collections of the resulting low-resolution and high-resolution sub-maps. Let \hat{C}_h be the collection of the high-resolution sub-matrices generated by neural network. We train a neural network using C_ℓ and C_h . The mean square error(MSE) is used as loss function in the training process.

$$MSE[C_\ell, C_h] = \frac{1}{K \times K} \sum_{i=1}^{K \times R} \sum_{j=1}^{K \times R} (\hat{C}_{h_{i,j}} - C_{h_{i,j}})^2$$

Where $C_{\ell_{i,j}}$ and $C_{h_{i,j}}$ represent left-top coordinate (i, j) in C_ℓ and C_h respectively. We can use (f, n) to represent the parameters of each layer. Parameter f means the size of the filter and n means the number of filter.

1st Layer (Pattern extraction) Base on every $K \times K$ sub-matrix, using $f \times f$ (13×13 in HiCPlus) filters to extract patterns of each sub-contact-map. Which can be represented by following formula:

$$F_1(X) = ReLU(w_1 * X + b_1)$$

where $*$ represent the convolution process. w_1 represents $n \times f \times f$ filters. b_1 is the bias.

2nd Layer (Low-res mapping to high-res)

3rd Layer (Predicted contact maps generation)

Use other chromosome.

Do the same dividing process like M_ℓ and M_h

Calculate the Pearson's correlation between the output and M_h .

Step 1 Data preparation and processing

Since this experiment is to validate the algorithm for mapping low-resolution data to high-resolution data, high-resolution data are required.

In order to compare to some state-of-the-art approaches (HiCPlus and HiCNN), we use data sets (such as GM12878 from GSE63525) which are also used in other approaches. We start from generating a 10kb resolution contact map using Hi-C Pro. Then we perform down-sampling on high-resolution data. We use BAM files to generate low-resolution contact maps by changing the bin size bigger. We generate three contact maps with bin sizes are 20kb, 30kb and 40kb, respectively. We use chromosome 1-8 as training sets, and chromosome 17 as test set.

Step 2 Learning by Neural network

We separate the low-resolution contact map into many $K \times K$ sub-matrices. Those sub-matrices are used as inputs.

2.1 Layer Structure

We consider the

References

- [1] MIZUSHIMA, N., AND KOMATSU, M. Autophagy: renovation of cells and tissues. *Cell* 147 (2011), 728–741.