Enhancing Hi-C contact map resolution with neural network

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1 Introduction

Recently, the high-throughput chromoseome comformation capture (Hi-C) technique has become a powerful tool for studying the three-dimensional structure of chromosomes. Hi-C data is usually expressed as a $n\times n$ 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. Which 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 make a 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 $R = \frac{S_h}{S_{\ell}}$. And P be the number of overlapping pixels between adjacent sub-maps. K be the size of low-resolution sub-maps.

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% Divide matrices M_\ell and M_h To M_\ell :
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for i = 1, 1+K-P, 1+2×(K-P), ...
for j = 1, 1+K-P, 1+2×(K-P), ...
IF i+K > M_{\ell} \mid \mid j+K > M_{\ell}, BREAK
ELSE extract K × K sub-maps whose left-top coordinate
is (i,j) from M_{\ell}.
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Do the same process to M_h :

for
$$i$$
 = 1, 1+($K-P$) \times R , 1+2×($K-P$) \times R , ... for j = 1, 1+($K-P$) \times R , 1+2×($K-P$) \times R , ... 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 .

Let C_{ℓ} and C_h be collections of the resulting low-resolution and high-resolution sub-maps. 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.

$$MSE[C_{\ell}, C_{h}] = \frac{1}{K \times K} \sum_{i=1}^{K} \sum_{j=1}^{K} (C_{\ell_{i,j}} - C_{h_{i,j}})^{2}$$

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.

Layer1(Pattern extraction) Base on every $K \times K$ sub-matrix, using $f \times f(13 \times 13)$ in HiCPlus) filters to extract patterns of each sub-contact-map. Which can represented by following formula:

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

Where * represent the convolution process. w represents $n \times f \times f$ filters. b is the bias.

Layer2(Low-res mapping to high-res)

Layer3(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. $Cell\ 147\ (2011),\ 728-741.$