OMTON, Volume 32

Supplemental information

ADEVO: Proof-of-concept of adenovirus-directed

EVOlution by random peptide

display on the fiber knob

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Figure S1. High titer AdV production is achievable under conditions where the cross-packaging rate is low.

A: Experiment design. WT and ∆Fiber Ad5 genomes were transfected at a 1:1 ratio. Since only WT genomes can produce functional capsids, the ratio of ∆Fiber genomes in cells infected with progeny virions indicates the cross-packaging rate.

B: Progeny titer (blue) and cross-packaging rate (red) as a function of the AdV genome linearized plasmids transfection dose. The experiment was performed twice independently. With our AdV rescue protocols, doses of 5000 transfected plasmids per cell and below facilitated high titer AdV rescue while yielding less than 3% of cross-packaging.

Figure S2. Statistical modeling of library complexity.

A: Complexity estimates yielded by different relevant statistical models, namely Mh and Mth models of Chao (1), Chao et al. (2), Lee et al. (3), or the "Mth Chao (LB) estimate of R's Rcapture package (4), for the different libraries. Complexity estimates inferior to the observed complexity are written in orange. Models did not substantially differ in their relative variability displayed for estimates computed from different combinations of three aliquots from the same library when available. Chao's model was chosen because it came close to the average of all models and never yielded abherrantly low estimates. When four libraries aliquots had been sequenced, the registered complexity was the average of Chao's estimator for all four possible combinations of three aliquots, in order not to introduce bias with libraries for which only three aliquots had been sequenced.

B: Numbers of identified variants and overlap in the three or four sequenced aliquots (#A, #B, #C and/or #D) of round 0 libraries. The observed complexity is the total number of identified variants in all library aliquots combined.

Figure S3. Predicted furin cleavage sites in the selected fiber proteins.

Furin cleavage sites along the whole fiber protein of Ad5-WT or of the selected variants were predicted by the ProP online tool (5). The insert position is indicated in yellow.

Figure S4. Benzonase treatment eliminates virtually all non-encapsidated DNA without affecting encapsidated DNA.

A: Benzonase digests more than 99.9% of free DNA. Two million HEK293 cells were transfected with 500 billion copies of a plasmid carrying an ampicillin resistance gene. Four hours post transfection, cells and media were harvested, split in three equal groups and processed for benzonase treatment and DNA purification as NGS samples ("benzonase" condition), or without benzonase treatment ("untreated"), or without separating cells and media but instead conducting four freeze and thaw cycles of full lysate before benzonase treatment ("full lysate"). Undigested purified plasmids were numbered by qPCR and normalized on the "untreated" sample. $N=2$.

B: Benzonase treatment does not disrupt VPs. Two million HEK293 cells were transfected with 1 billion copies of linearized luciferase-expressing HAdV-C5 genomes using the jetOPTIMUS reagents. At 8 dpi, cells and media were harvested and frozen and thawed 4 times. Aliquots of 2% of the total lysate volume were taken and submitted or not to benzonase treatment with the same conditions as for NGS samples preparation except that cells and media were not separated. New HEK293 cells were subsequently infected with the lysate and submitted to luciferase assay in order to quantify infectious VPs whose genomes were left intact. N=2.

Table S1. Primers used in this study.

Blue sequences indicate either one on two random codons (Protocols 1, 2 and 3 random oligonucleotides), or the insert sequence of the primers used for variant recloning. "B" corresponds to C, T or G; "K" to T or G; and "V" to A, C or G.

Table S2. NGS analysis of genome libraries.

Aliquots of one Protocol 1 library and one Protocol 2 libraries were analysed by NGS after genome reassembly and prior to transfection. The dominant variant count is the number of reads corresponding to the most abundant variant of the library.

Table S3: NGS inserts counter (nucleotide)

The following python code takes as input the two .fastq files of paired reads returned by the sequencing company, and a list of parameters to be entered by the user. Reads are allocated to the corresponding library aliquot, undergo quality controls, then true unique inserts are identified. The output is one .csv table per library aliquot, containing the nucleotide sequence of each unique insert with their associated count, in order of decreasing abundance.

Table S4: NGS inserts counter (peptide)

The following python code takes as input an output file of Supplemental code 1, translates all insert nucleotide sequences into amino acid sequences, merges the synonymous inserts and return a new .csv table containing each unique insert peptide sequence with their associated count, in order of decreasing abundance.

'''this file takes as input a .csv file of unique NGS inserts and their counts, as produced by Supplemental Code 1, and gives as output a .csv file of unique corresponding peptides and counts'''

inputfile='/Example_path_to_the_input_file.csv' output='/Example_path_to_the_output_file .csv'

from Bio.Seq import Seq import pandas as pd

```
rawseq=pd.read_csv(inputfile,sep=',')
peptides=[]
countDNA=[]
errorinsert=0
for k in range(rawseq.shape[0]):
   nt=rawseq.iloc[k,1]
```

```
if len(nt)\%3!=0:
     print('the sequence \#'+str(rawseq.iloc[k,0])+' does not have the correct length')
      print(nt)
      errorinsert+=1
   else:
      pep=str(Seq(nt).translate())
      peptides.append(pep)
      countDNA.append(rawseq.iloc[k,2])
   if 'X' in pep:
      print(pep,k,nt)
print('translation finished')
df=pd.DataFrame({})
df['sequence']=peptides
df['count']=countDNA
df=df.sort_values(by=['sequence'])
print('sorting finished')
uniquePeptides=[]
uniqueCounts=[]
memory=''#the last insert at each step of the recurrence, to know if the insert is unique or not
for k in range(df.shape[0]):
   peptide=df.iloc[k,0]
   number=df.iloc[k,1]
   if peptide==memory:
      uniqueCounts[-1]=uniqueCounts[-1]+number
   else:
      memory=peptide
      uniquePeptides.append(peptide)
      uniqueCounts.append(number)
print('unique peptides identified and counted')
print('number of unique peptides: ',len(uniquePeptides))
result=pd.DataFrame({})
result['sequence']=uniquePeptides
result['count']=uniqueCounts
result=result.sort_values(by=['count'], ascending=False)
print('count of the most abundant peptide: ',result.iloc[0,1])
result.to_csv(output)
```
Table S5: Library complexity estimator

The following python code takes as input the peptide sequences and counts tables returned by Supplemental code 2 for the library aliquots considered. It is determined in how many aliquots each insert is present. Then, the number of variants present in exactly 1, 2 or 3 aliquots (in case 3 library aliquots were considered) is counted. Finally, total library complexity is estimated using Anne Chao's simple capture-recapture estimator.

import pandas as pd from random import randint from statistics import mean,stdev

def WennDiagram(aliquots):

'''

 aliquots is a list of paths towards peptide or nt inserts counts files from the same library. the function identifies in which aliquots each insert was identified so that a Wenn Diagram of aliquot overlap can be manually built,

```
 and library size can be estimated in the functions below
 '''
   nbaliquots=len(aliquots)
   overlaps=[0 for k in range(2**nbaliquots)]
   # this list indicates the overlaps between aliquots following a basis 2 indexation.
  # for example, if there are 4 aliquots and an insert is present in aliquots
  \# 0 and 2 but not 1 and 3 (python numbering so there is an aliquot \# 0), the
  # insert will be counted at the position 2^{**}0 + 2^{**}2 = 5 (also python numbering)
   insertnotfused=[] # contains the list of insert in each aliquot
   nextinserts=[]
   nbinserts=[]
   index=[]
   stopsign=[]
   for aliquot in aliquots:
      insertlist=list(pd.read_csv(aliquot)['sequence'])
      insertlist.sort()
     nextinserts.append(insertlist[0])
     nbinserts.append(len(insertlist))
     insertlist.append('z')
      # this is the stop mechanism. In the next loop, the end of the list is
     # detected when the studied "insert" is 'z'. All true inserts will come
     # before this one in the alphabetic order
      insertnotfused.append(insertlist)
      index.append(0)
     stopsign.append('z')
  while nextinserts!=stopsign:
     studiedinsert=min(nextinserts)
     score=0
      for k in range(nbaliquots):
        if nextinserts[k]==studiedinsert:
          score\pm=2**k # now let's update nextinserts
          index[k]=1 nextinserts[k]=insertnotfused[k][index[k]]
      overlaps[score]+=1
```

```
 return overlaps
```

```
result=WennDiagram(['/path_to_aliquot_1_inserts_peptide_table.csv','/path_to_aliquot_2_inserts
peptide table.csv','/path_to_aliquot_3_inserts_peptide_table.csv.csv'])
```

```
def overlaps3toSum(ov):
```
 '''transforms a list of overlaps for 3 aliquots to a list which counts how many inserts were found in respectively 1, 2 or 3 libraries''' return $\frac{1}{|ov[1]+ov[2]+ov[4]}$, $ov[3]+ov[5]+ov[6]$, $ov[7]$]

execute the following lines only if the "result" list was built using the WennDiagram function from exactly three library aliquots

OverlapsCounts=overlaps3toSum(results) library_complexity=OverlapsCounts[0]+OverlapsCounts[1]+OverlapsCounts[2]+OverlapsCount $s[0]^{**}\overline{2}/(2*{\rm OverlapsCounts}[1])$ print(library_complexity)

Supplemental references

- 1. Chao, A. (1987) Estimating the Population Size for Capture-Recapture Data with Unequal Catchability. *Biometrics*, **43**.
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- 3. Lee, S.-M. and Chao, A. (1994) Estimating Population Size Via Sample Coverage for Closed Capture-Recapture Models. *Biometrics*, **50**.
- 4. Baillargeon, S. and Rivest, L.-P. (2007) TheRcapturePackage: Loglinear Models for Capture-Recapture inR. *Journal of Statistical Software*, **19**.
- 5. Duckert, P., Brunak, S. and Blom, N. (2004) Prediction of proprotein convertase cleavage sites. *Protein Eng Des Sel*, **17**, 107-112.