

Homology modeling with Modeller

Bioinformatics II Presentation

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Outline

- Goal: Predict the structure of 1yje:A using homology modeling
- Introduction:
 - What is homology modeling?
 - What is Modeller?
- Methods:
 - Initial search
 - Alignment
 - Structure prediction
- Results: Predicted structures and conclusions

Introduction: Homology modeling

- Homologous proteins have similar structures
- Many proteins are homologous to *something* with a known structure
- Structure is *much* more conserved than sequence

```
>P1:1yje_A
structureX:1YJE.pdb: 362 :A:+226 :A:: 2.40: 0.25
-----NLLTSLIRAHLDSPGNTAKLDYSKFQELVLPFRGKEDAGDVQQFYDL
LSGSLDVIRKWAEEKIPGFIELSPGDQDLLLESFALELFILRLAYRSKPGEGLIFCSGLVLHRLQCARGFGDWID
NILAFSRLHSLGVDVPAFACLSALVLTDRHGLQDPRRVEELQNRIASCLKEHMAAVS-----CLSRLLGKL
PELRTLCTQGLQRIFFCLKLEDLVPPPIVDKIFMDT----*
```

```
>P1:1yje_A
sequence:: : : : :-1.00:-1.00
GSSHHHHHHSSGLVPRGSHMQPPDASPTNLLTSLIRAHLDSPGNTAKLDYSKFQELVLPFRGKEDAGDVQQFYDL
LSGSLDVIRKWAEEKIPGFIELSPGDQDLLLESFALELFILRLAYRSKPGEGLIFCSGLVLHRLQCARGFGDWID
NILAFSRLHSLGVDVPAFACLSALVLTDRHGLQDPRRVEELQNRIASCLKEHMAAVAGDPQPASCLSRLLGKL
PELRTLCTQGLQRIFFCLKLEDLVPPPIVDKIFMDLSF*
```



Introduction: Modeller

- Free Python package due to Sali et al. (1994)
- No GUI—knowledge of Python required
- Predicts structures by “satisfaction of spatial restraints”
 - Spatial features common to template structures are globbed
 - Energy or probability functions are assigned to each spatial feature, imposing restraints
 - The model that best satisfies these restrains is selected
 - Examples: molecular PDF, GA341, DOPE (Shen and Sali 2006)

Methods: Initial search

- Target (1yje:A) was CSBlasted against the non-redundant PDB
 - CSBLAST looks for matches by considering the *context* around amino acids (Biegert et al. 2009)
- Hits with $e < 0.01$ were selected as potential templates
 - The first subsequent hit was *much* higher than 0.01, so a clear jump was visible
- An initial MSA was generated using MUSCLE (Edgar 2004)

Methods: Alignment

- Modeller has its own alignment method
 - Required inputs: sequence (in PIR format), .pdb files
 - Gap penalties are lessened outside of secondary structures
- High matches were globbed and placed in a multiple sequence alignment

```
# align the unknown sequence with all templates
seqname = (os.path.basename(seqfile).split('.')[0])
seqname = seqname[:-1]+seqname[-1].upper()
templates = read_templates('../templates/most_relevant.dat')[n_templates]
templates = filter(lambda t: t['id'] != '1YJE', templates)
n_templates = len(templates)

# check whether the alignment exists already
if not os.path.isfile(seqname+'-multiple_n_'+str(n_templates)+'_ali'):
    aln = alignment(env)
    for template in templates:
        id = template['id']
        chain = template['chain']
        tplname = get_tplname(template)

        mdl = model(env, file=id, model_segment=('FIRST:'+chain, 'LAST:'+chain))
        aln.append_model(mdl, align_codes=tplname, atom_files=id+'.pdb')

# add the unknown sequence to the alignment.
aln.append(file=seqfile, align_codes=seqname)

# Align sequence-structures
# Note: align2d is obsolete! --> salin
# Parameters
# - gap_function makes the gaps in the MSA dependent on the structural
#   context
aln.salign(gap_function=True)

# store the alignment
aln.write(file=seqname+'-multiple_n_'+str(n_templates)+'_ali', alignment_format='PIR')
```

Methods: Structure prediction

- MSA was used to predict the structure of 1yje:A
 - The automodel class makes this easy
 - More powerful tools available
 - Select best of three models
 - Output: .pdb file

```
# automodel reads the alignment file and actually does the homology modeling for us.
# the output is a .pdb file, to be seen in a.outputs.
# the parameter assess_methods is used to check the quality of the model
# (see also Kryzhanovych and Fidelis, 2009)
a = automodel(env,
              alnfile = seqname+'-multiple_n_'+str(n_templates)+'_ali',
              knowns = map(get_tplname, templates),
              sequence = seqname,
              assess_methods=[getattr(assess, am) for am in assessment_models])

# index of the first/last model
# (determines how many models to calculate)
a.starting_model= 1
a.ending_model = 3

# do the actual homology modeling
a.make()
```

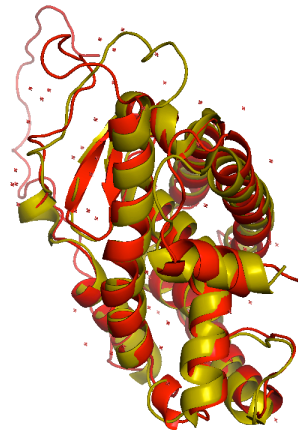
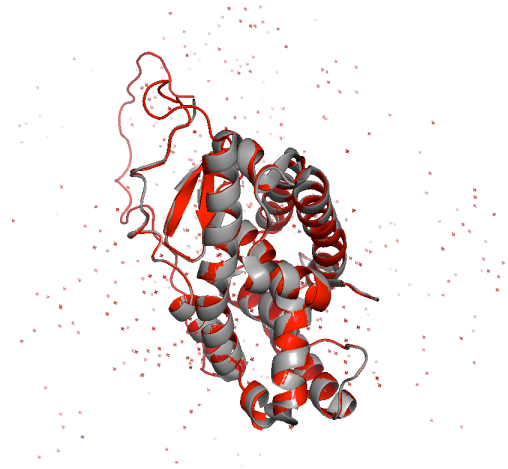
```
EXPDTA      THEORETICAL MODEL, MODELLER 9.10 2012/06/14 15:57:40
REMARK      6 MODELLER OBJECTIVE FUNCTION:      1825.7863
REMARK      6 MODELLER BEST TEMPLATE % SEQ ID:  27.064
ATOM        1  N   GLY      1   -50.586 -28.636 -32.626  1.00 14.91      N
ATOM        2  CA  GLY      1   -51.112 -28.454 -31.254  1.00 14.91      C
ATOM        3  C   GLY      1   -50.015 -28.632 -30.261  1.00 14.91      C
ATOM        4  O   GLY      1   -48.839 -28.659 -30.621  1.00 14.91      O
ATOM        5  N   SER      2   -50.381 -28.748 -28.971  1.00145.17    N
ATOM        6  CA  SER      2   -49.381 -28.916 -27.961  1.00145.17    C
ATOM        7  CB  SER      2   -49.843 -28.497 -26.554  1.00145.17    C
ATOM        8  OG  SER      2   -48.790 -28.684 -25.619  1.00145.17    O
ATOM        9  C   SER      2   -49.009 -30.362 -27.901  1.00145.17    C
.....
```

Evaluation of prediction quality

- ...can be performed using some quantitative measures
 - Assessments like Krystafovych et al. (2009)
 - molpdf, GA341 and DOPE (all built-in)
- ...can be done visually in PyMOL
 - With true structure (Flaig et al. 2005)
 - Without

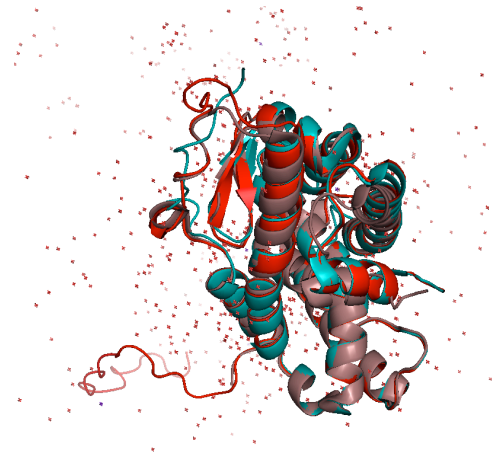
Results: Prediction checks

- Top: 1yje:A (red) fitted onto one template (1ovl:A)
- Bottom: Predicted structure (red) compared to real one
- Note poor fitting of loops, but overall good prediction of secondary structures
- DOPE score: -31212



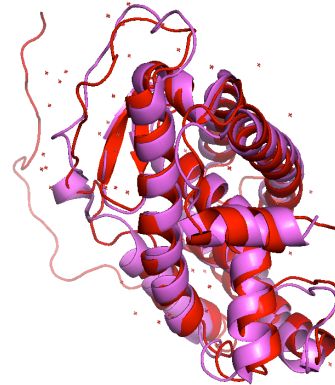
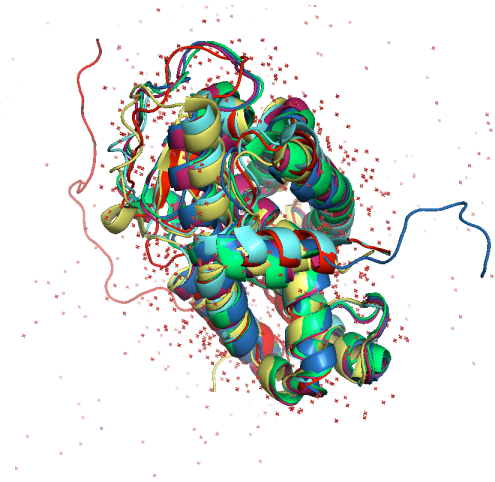
Results: Prediction checks

- Top: 1yje:A (red) fitted onto two templates (1ovl:A and 1pdu:A)
- Bottom: Predicted structure (red) compared to real one
- Loop fitting appears to have improved slightly
- DOPE score: -31385



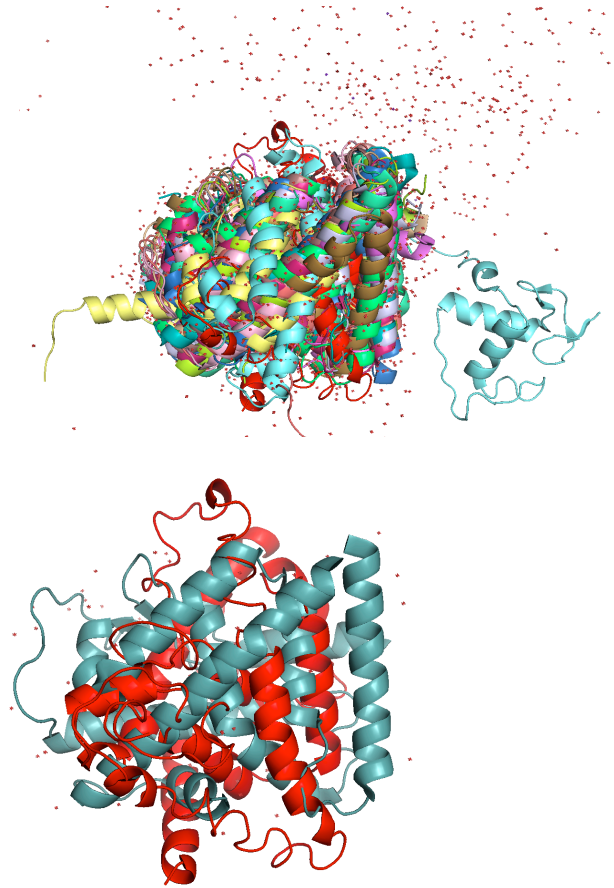
Results: Prediction checks

- Top: 1yje:A (red) fitted onto five templates
- Bottom: Predicted structure (red) compared to real one
- Loop fitting is now pretty good, overall
- DOPE score: -30780



Results: Prediction checks

- Top: 1yje:A (red) fitted onto fifteen templates
- Bottom: Predicted structure (red) compared to real one
- There appear to be diminishing returns
- DOPE score: -21209
- Notice the jump!



Conclusions

- Modeller can accurately predict secondary structures and even loop regions
 - Performs well when it doesn't get confused
 - Some level of skill is needed in selecting matches
- Current PDB may be enough to solve structure prediction problem (Zhang 2005)
 - Homology modeling may be the way of the future; alignment and selection are limiting
 - Just a bit too ambitious

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References

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