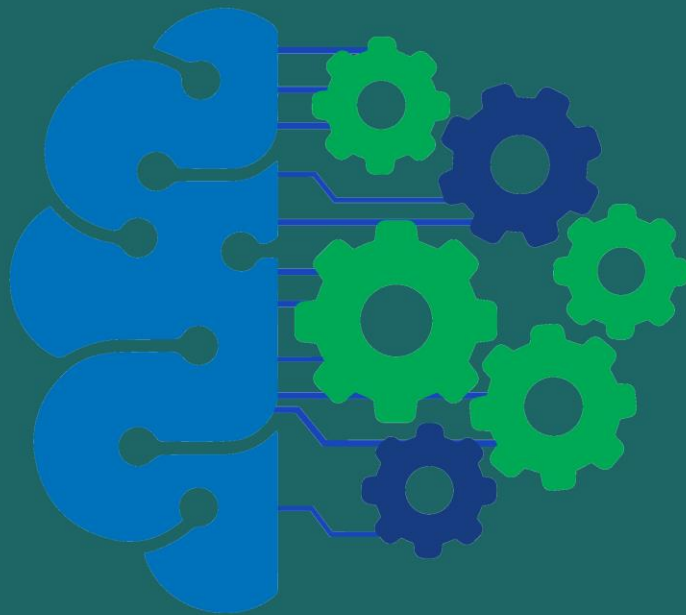


Neural networks for SAXS data analysis



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EMBL Hamburg

EMBL

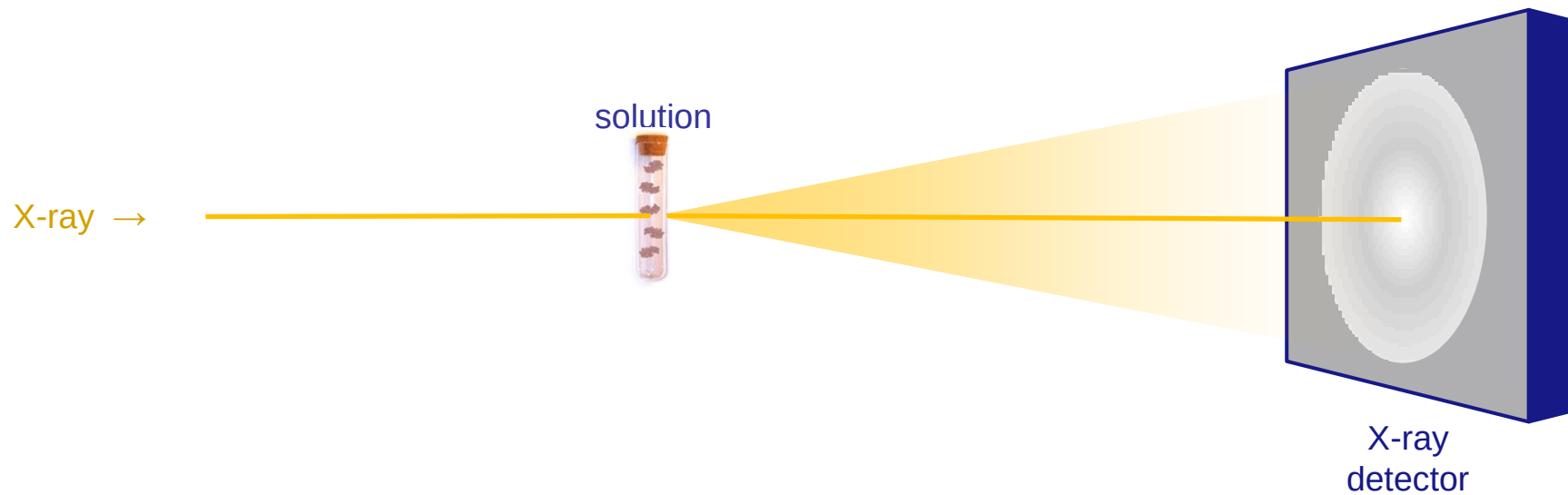


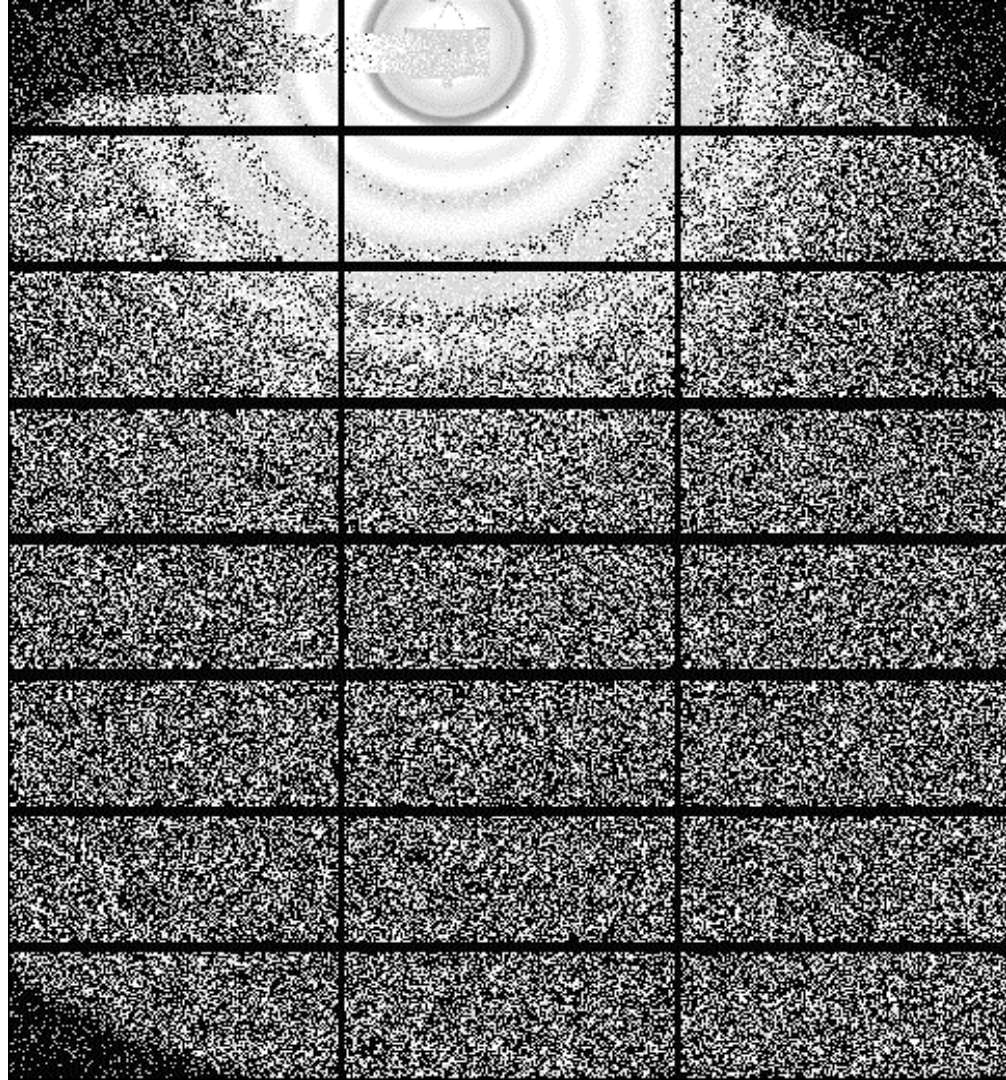
Outline

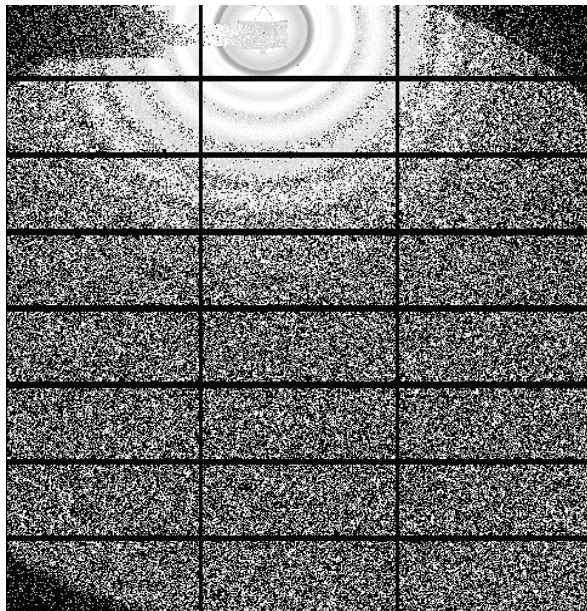
1. Small angle scattering
2. NN architecture and training set
3. Some results

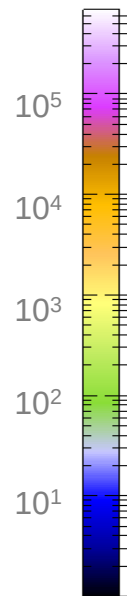
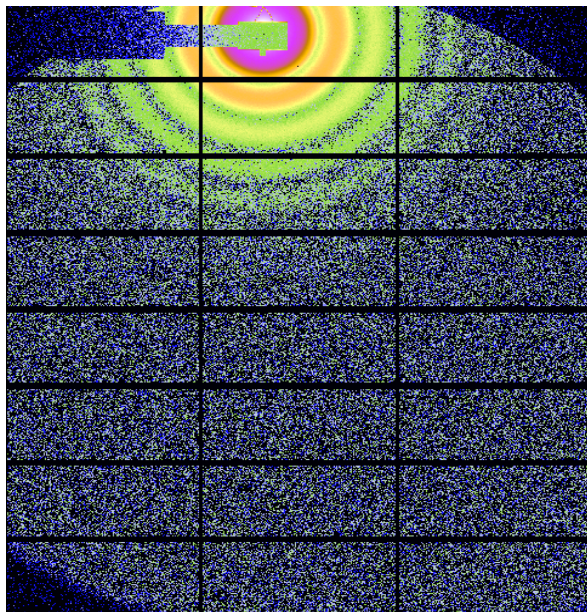
1. Small angle scattering

SAXS experiment

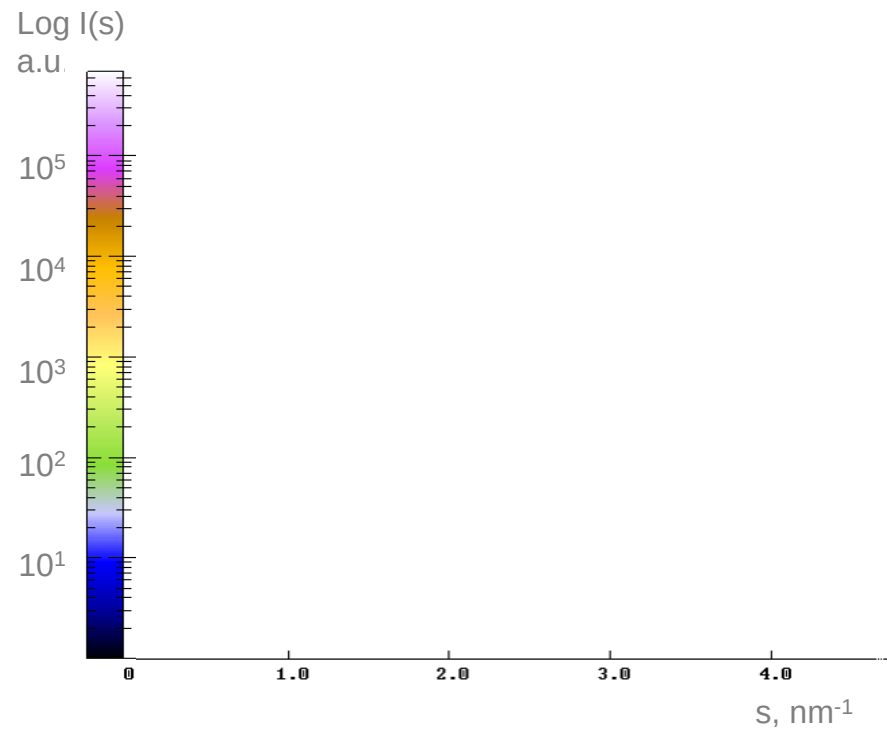
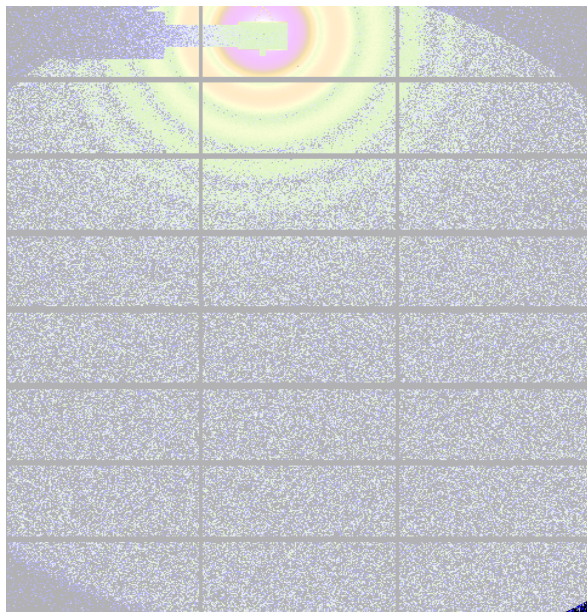




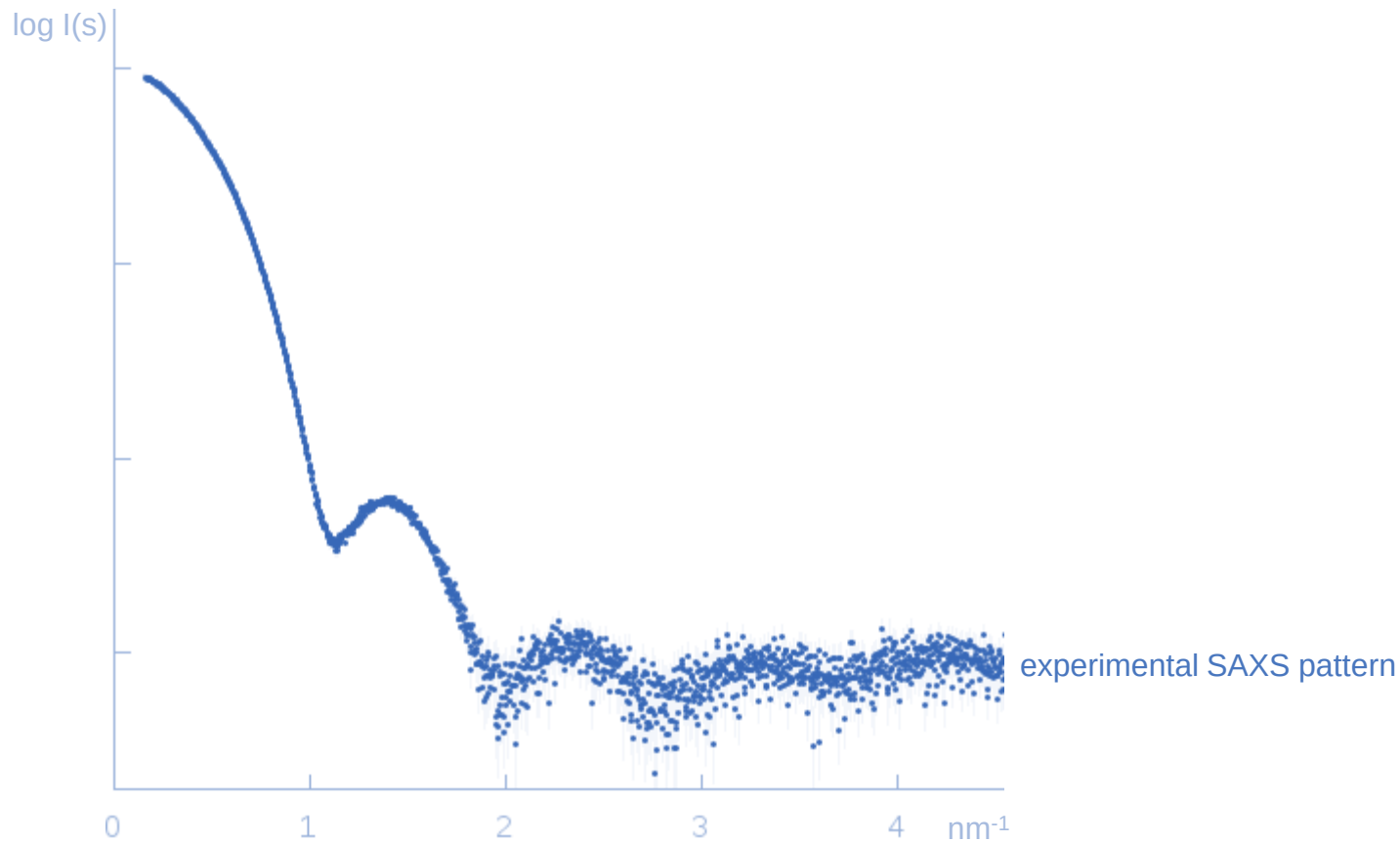




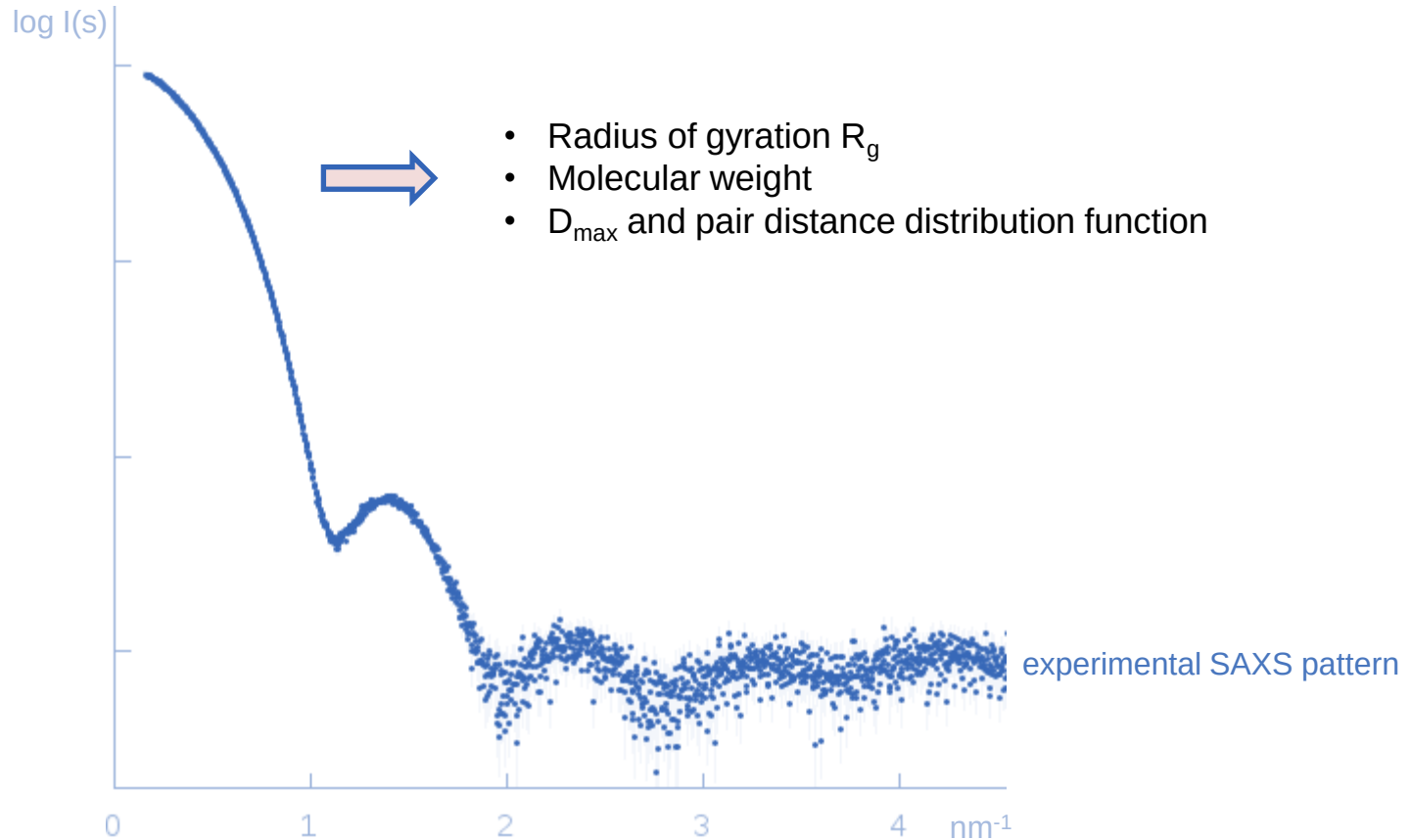
2D $\rightarrow$ 1D



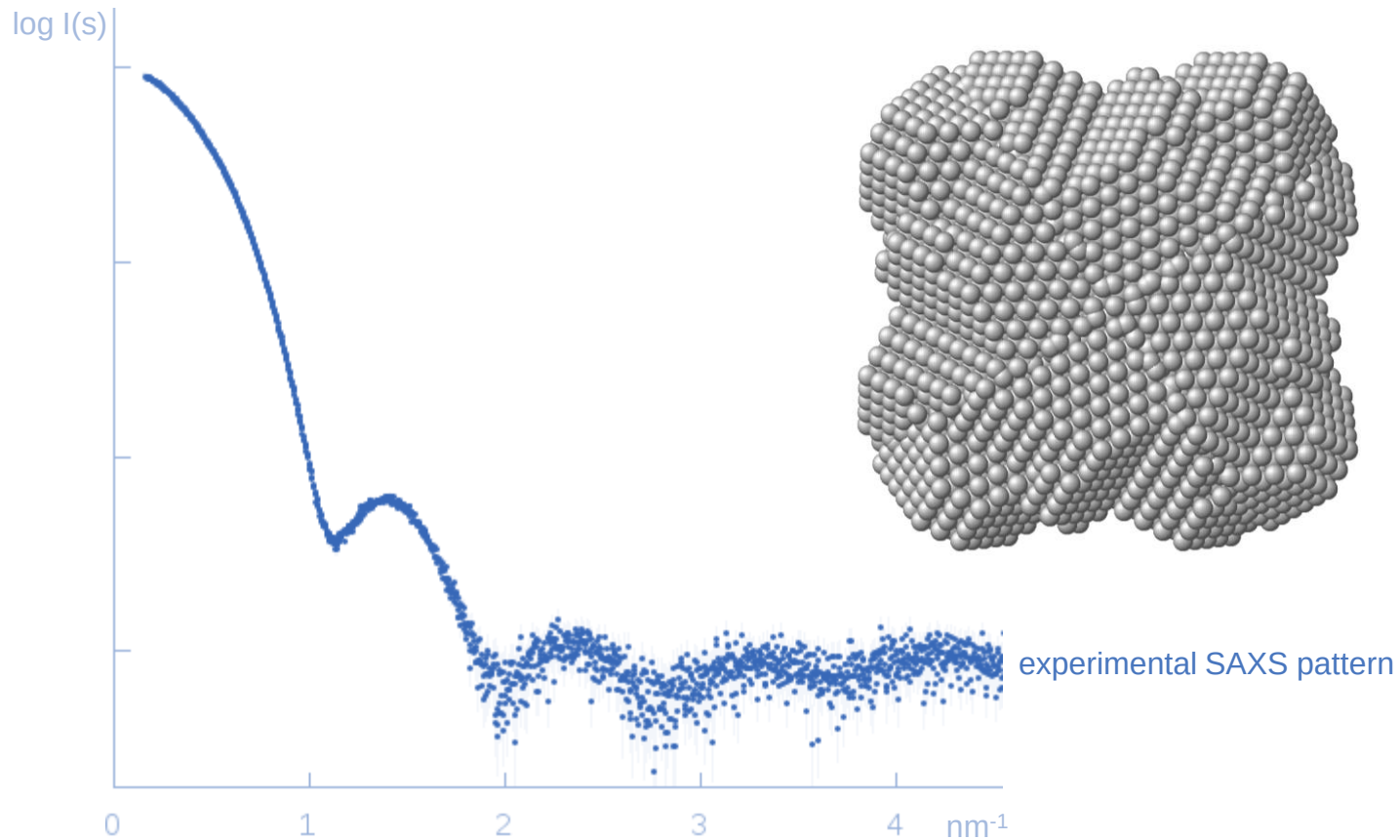
SA(X)S data



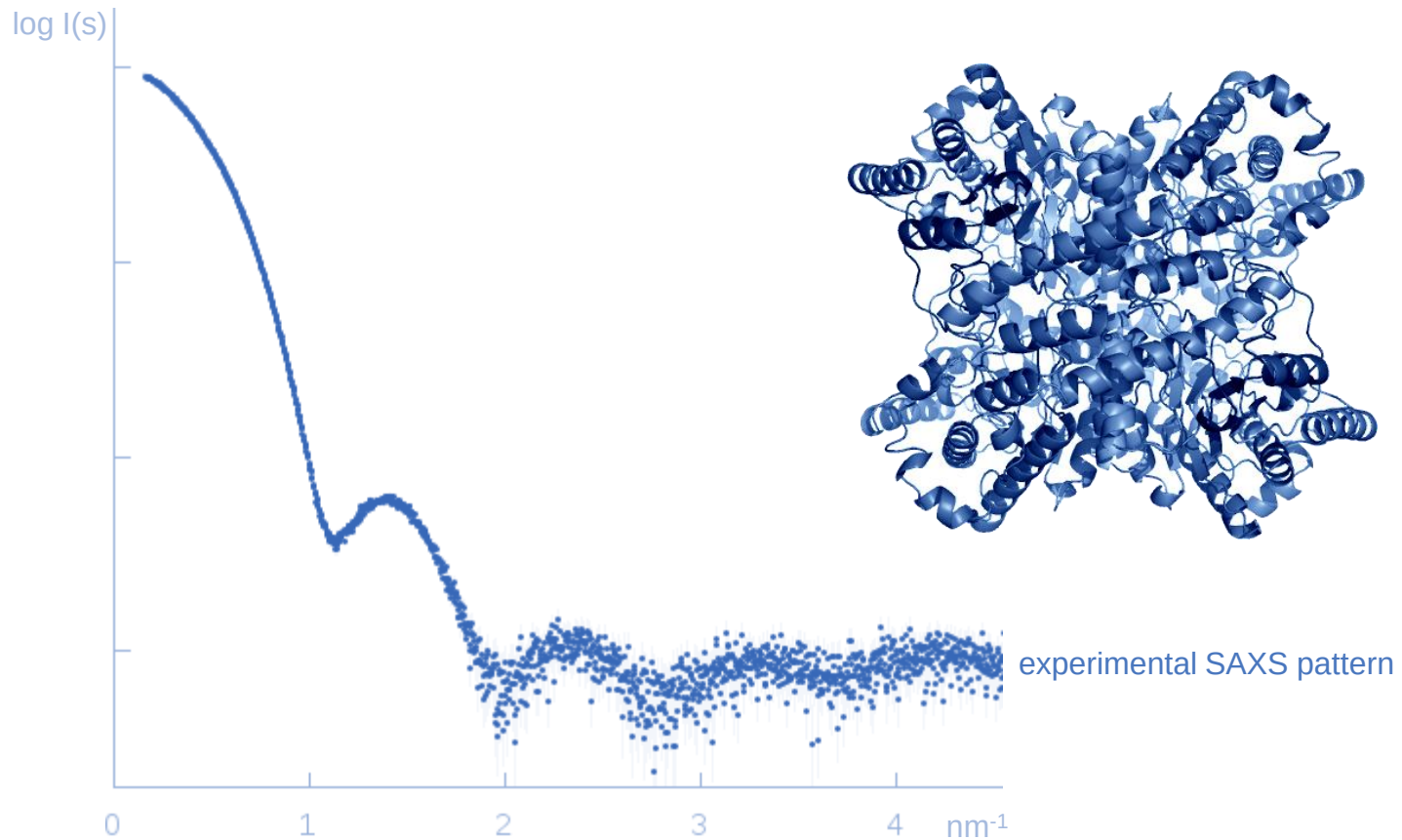
SA(X)S data



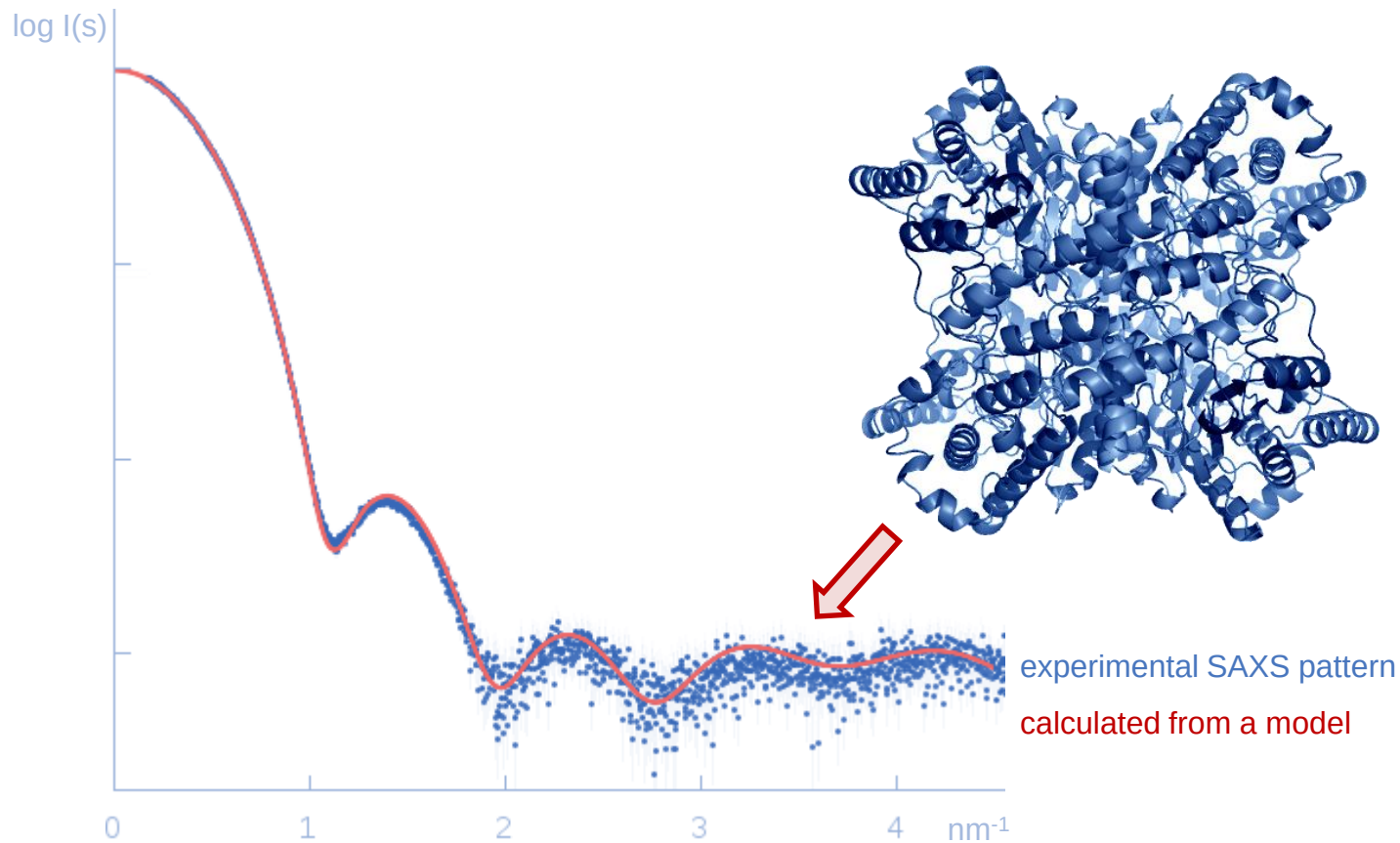
SA(X)S data



SA(X)S data

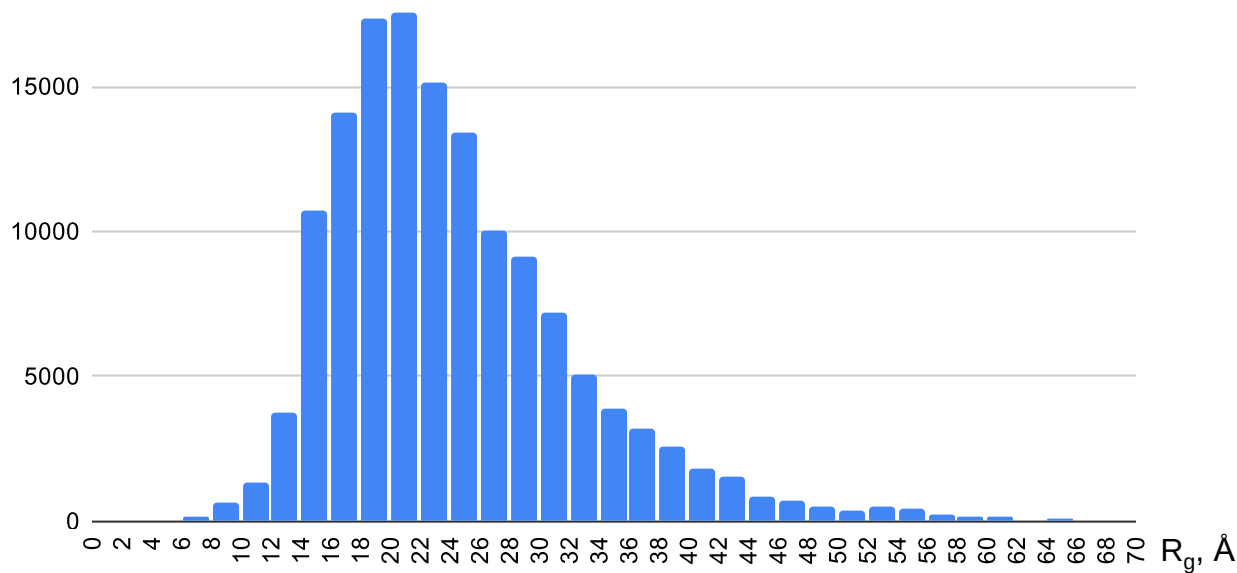


SA(X)S data

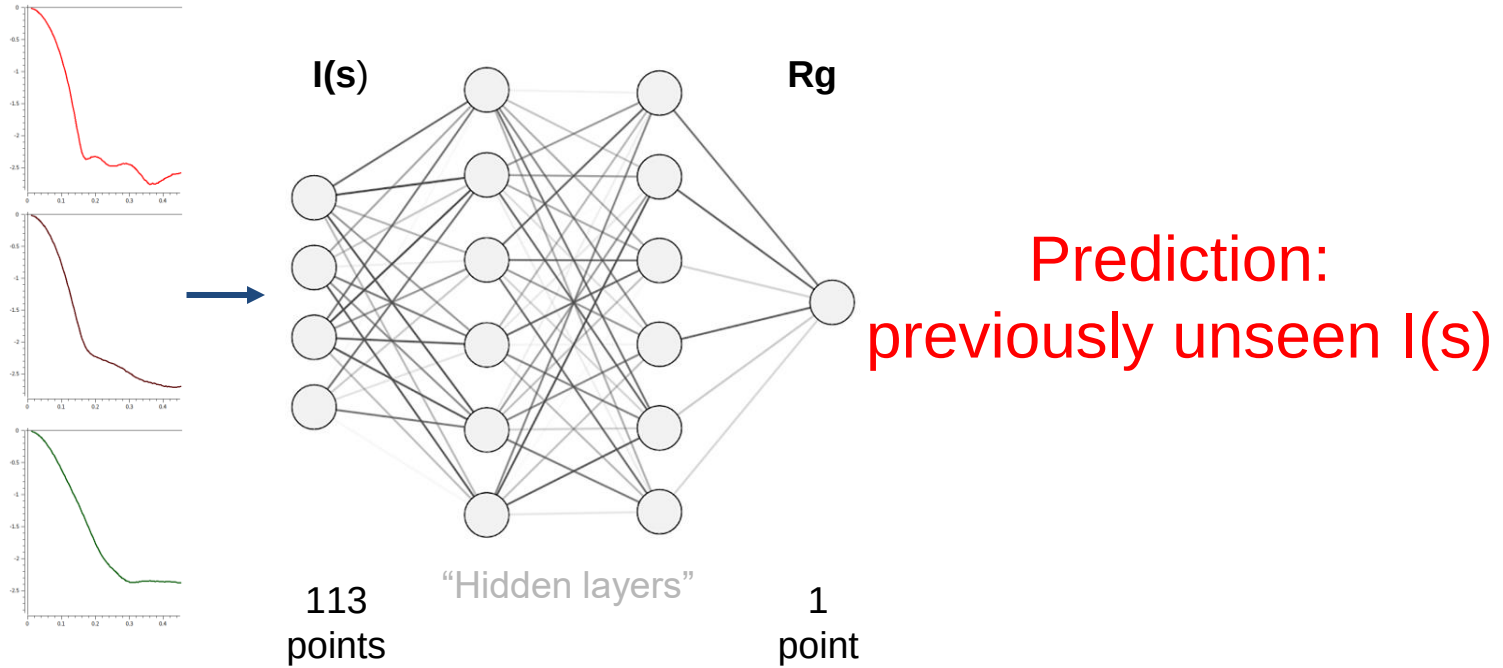


2. Training set and NN architecture

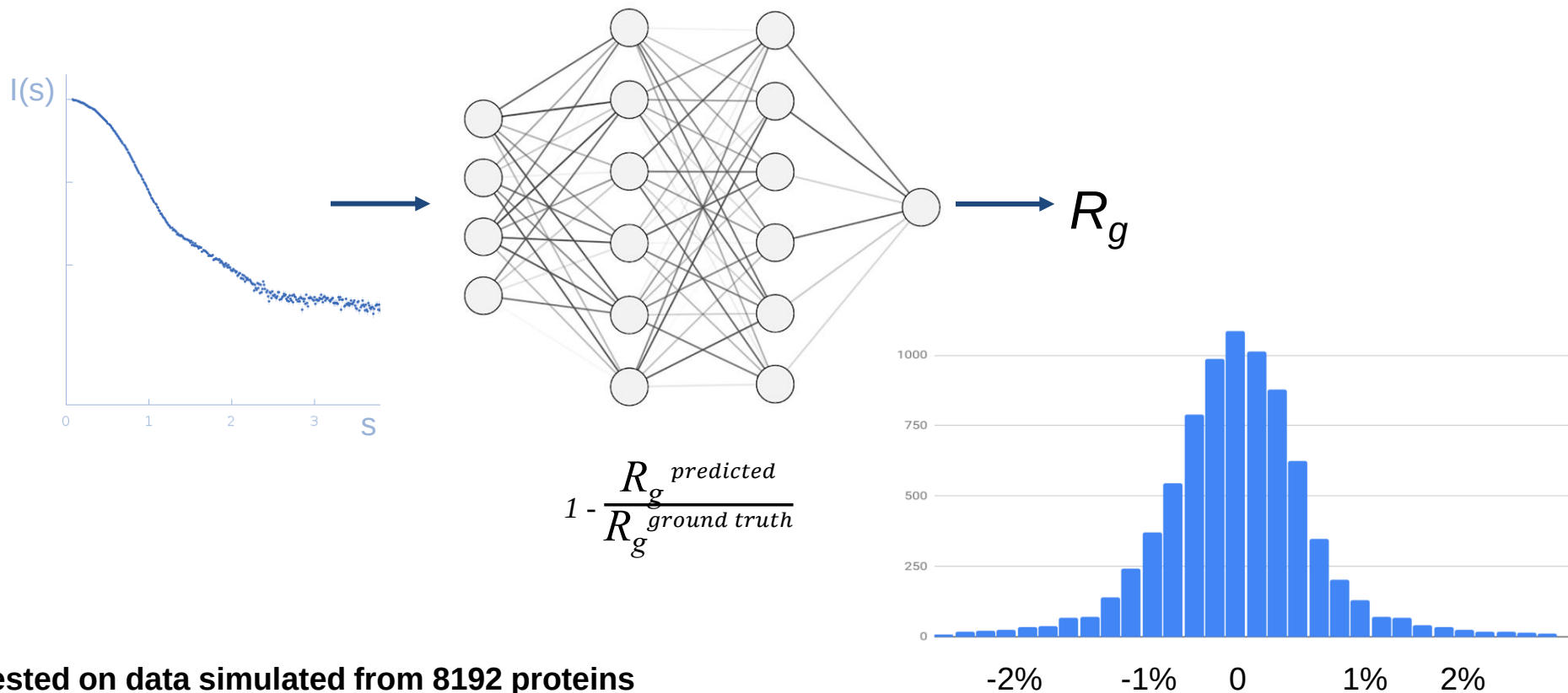
Training set: > 150 000 proteins



Supervised learning: thousands of $I(s)$ - R_g pairs

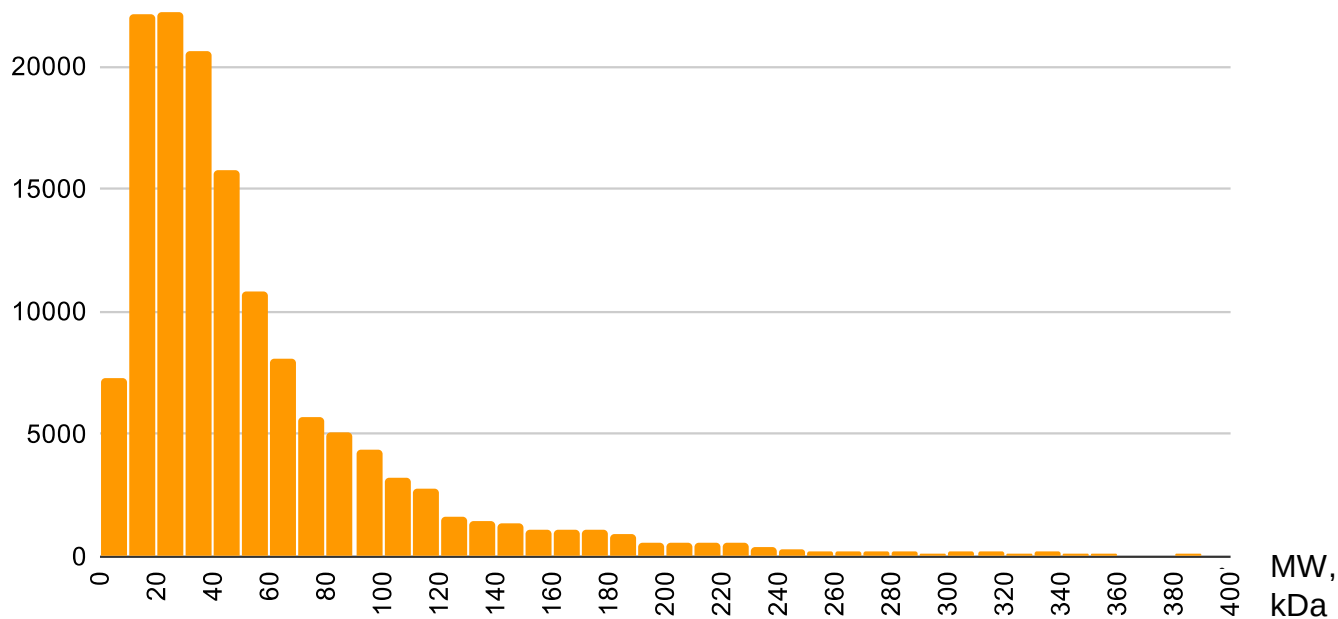


Determine R_g from SAXS data

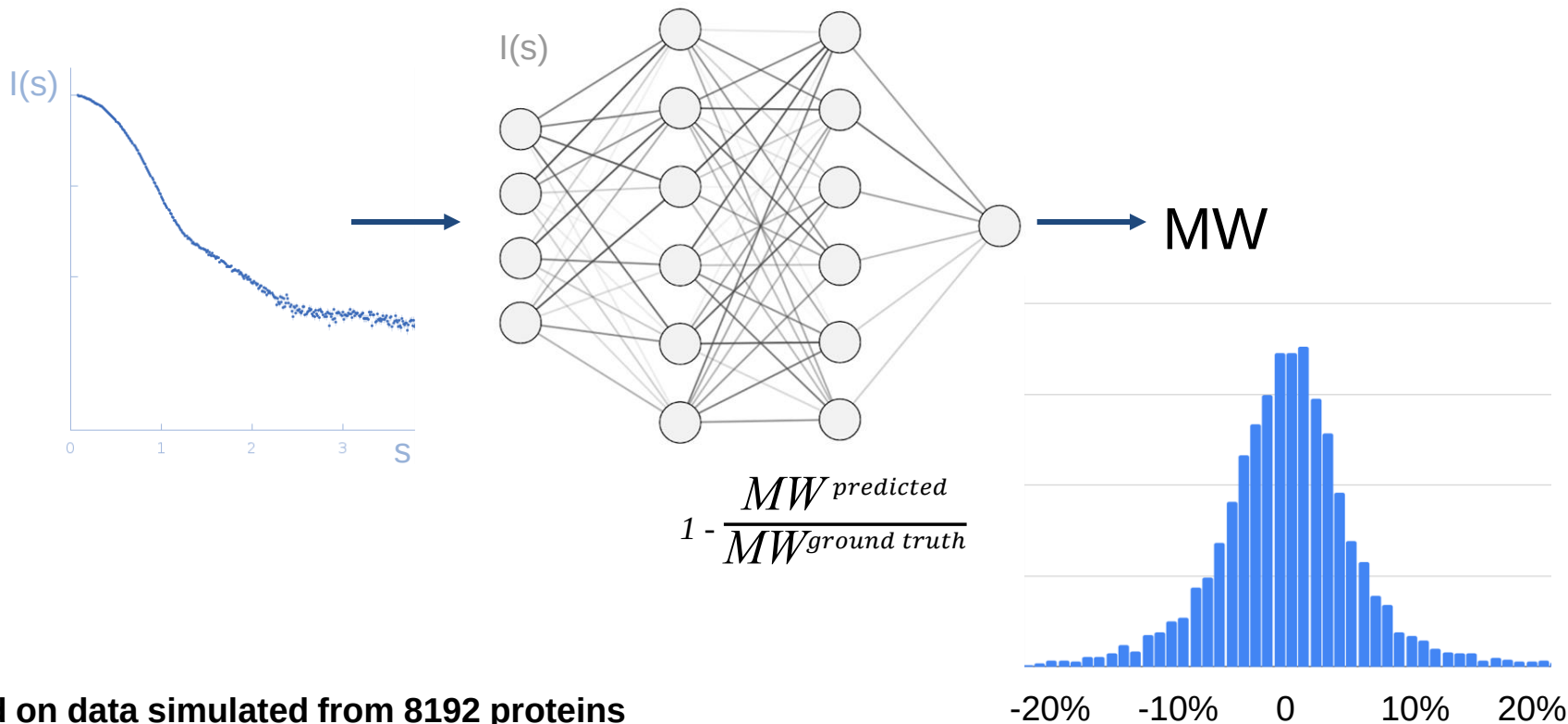


Tested on data simulated from 8192 proteins

MW distribution

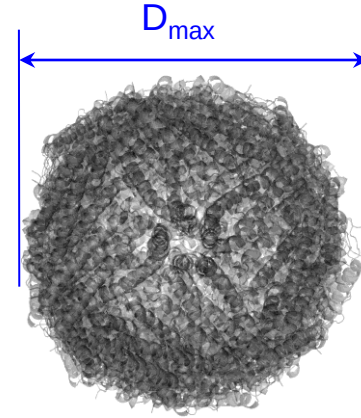
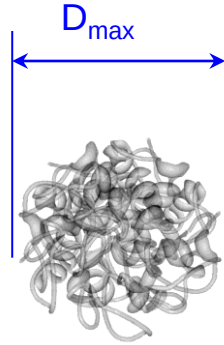


Determine *molecular weight* from SAXS data



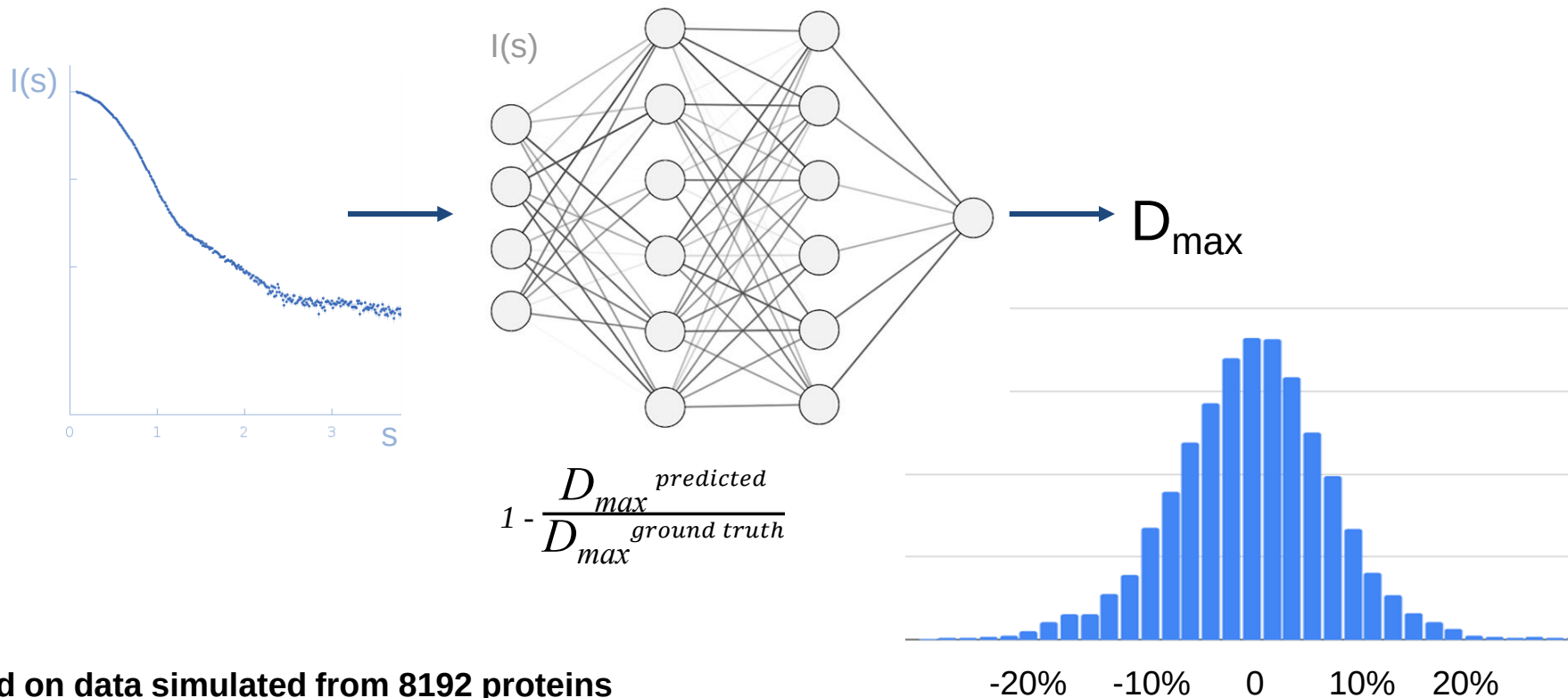
Tested on data simulated from 8192 proteins

Maximum intra-particle distance of the structure ($D_{\max}$)

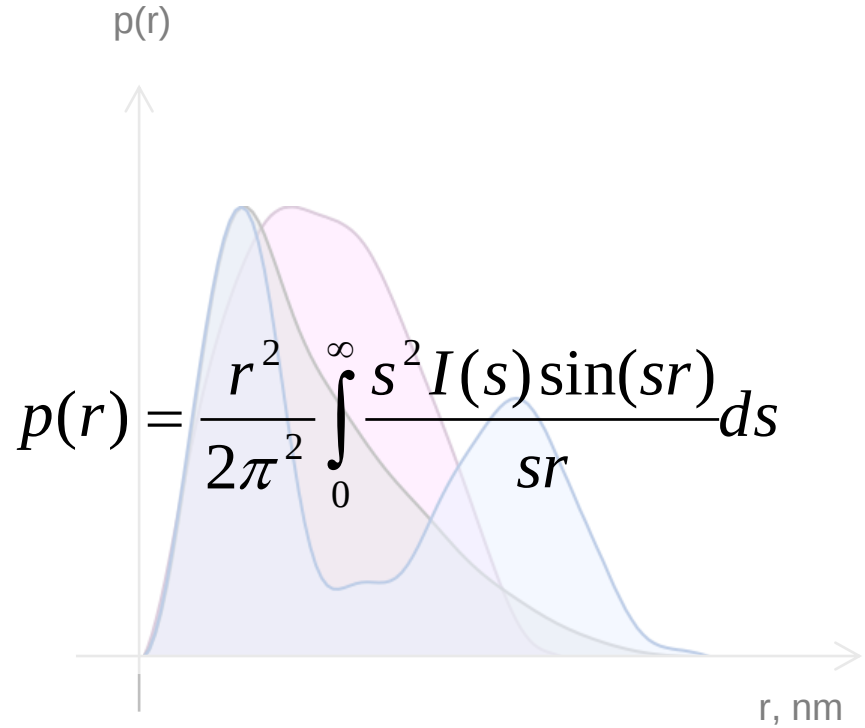
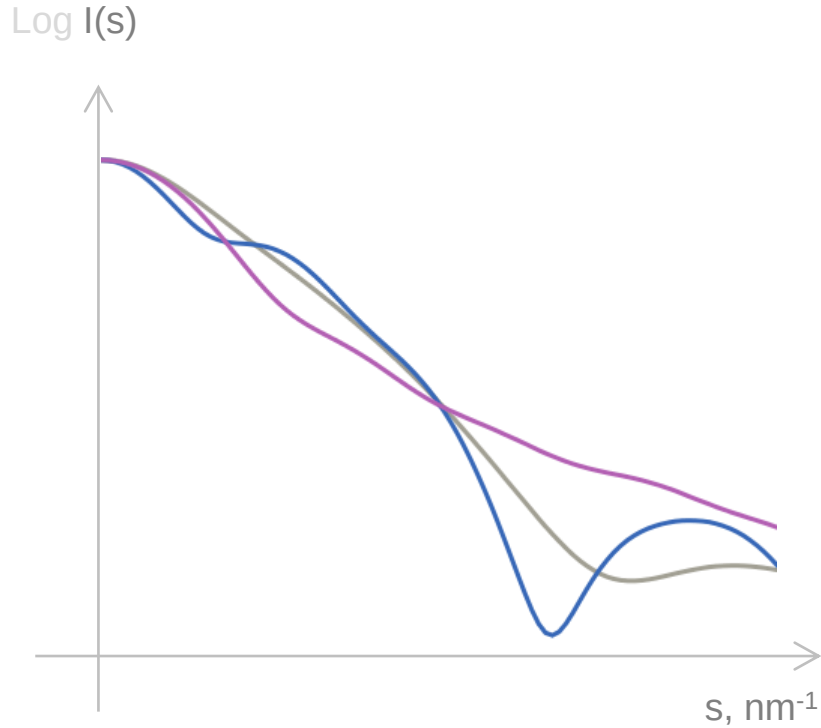


No direct mathematical way to obtain!

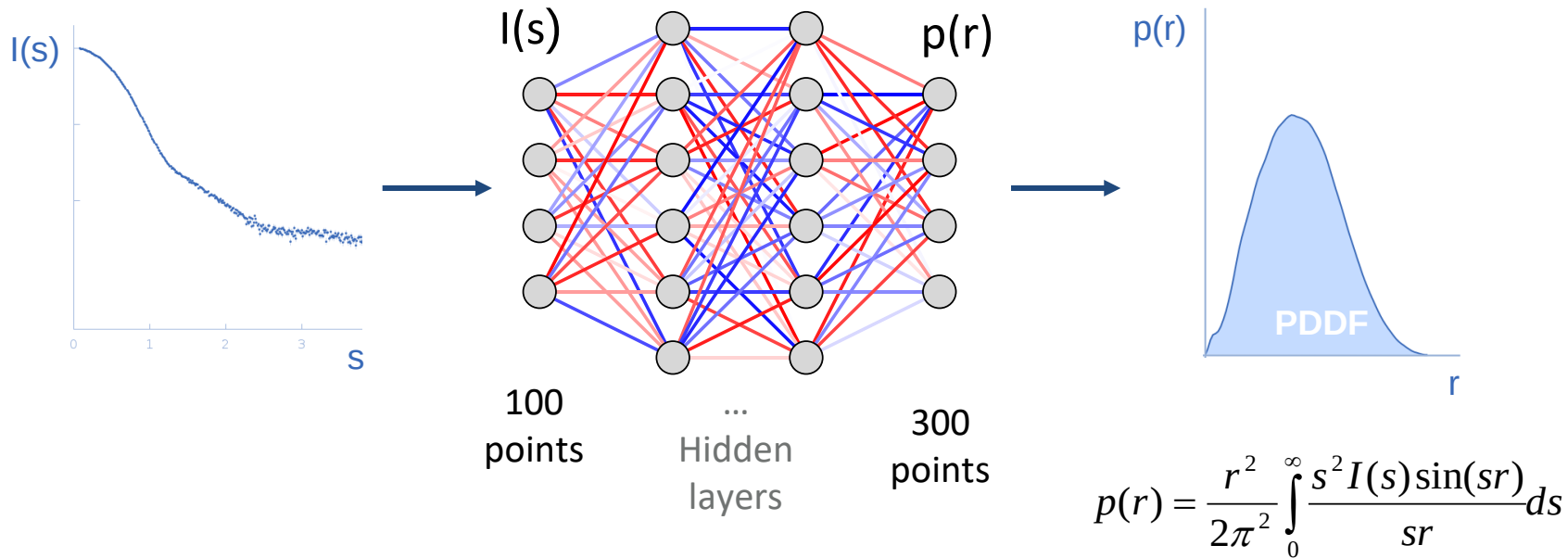
Determine D_{max} from SAXS data



Pair-distance distribution function (PDDF)



Determine **PDDF** from SAXS data



3. Where are we now?



Biological
Small Angle Scattering



DARA

[Home](#) > [Web services](#) > [DARA](#) > [gNNom](#)

GNNOM

Fit a PDDF $p(r)$ to experimental data

Experimental data (*.dat)

Choose File

SASDF99.dat

Angular units (optional)

▼

$s = 4\pi\sin(\theta)/\lambda$

Macromolecule type

protein ▼

Submit

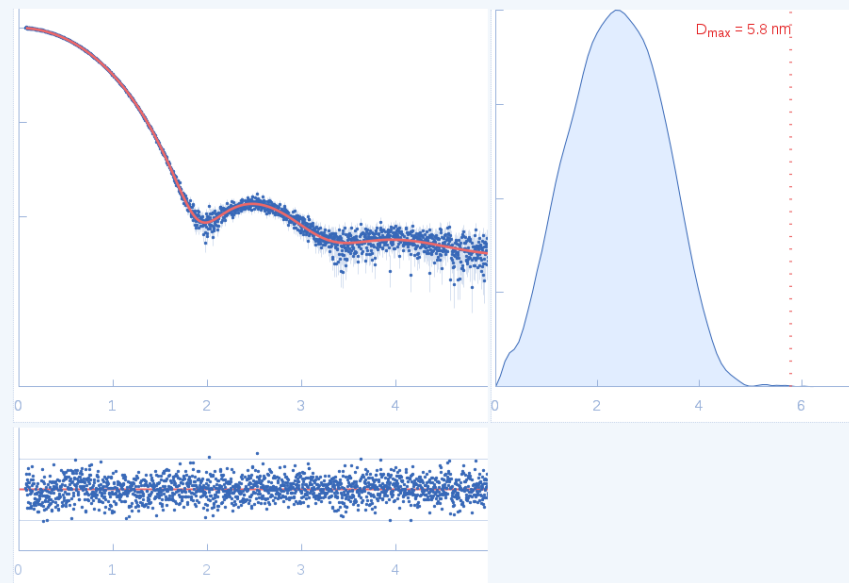
GNNOM

SASDFP8

Predicted R_g 1.8 nm
 $p(r)$ R_g 1.8 nm
 autorg R_g 1.8 nm

Predicted D_{max} 5.8 nm
 $p(r)$ D_{max} 6.2 nm
 Predicted MW 23.1 kDa
 χ^2 1.0

SASDFP8 – Carbonic anhydrase



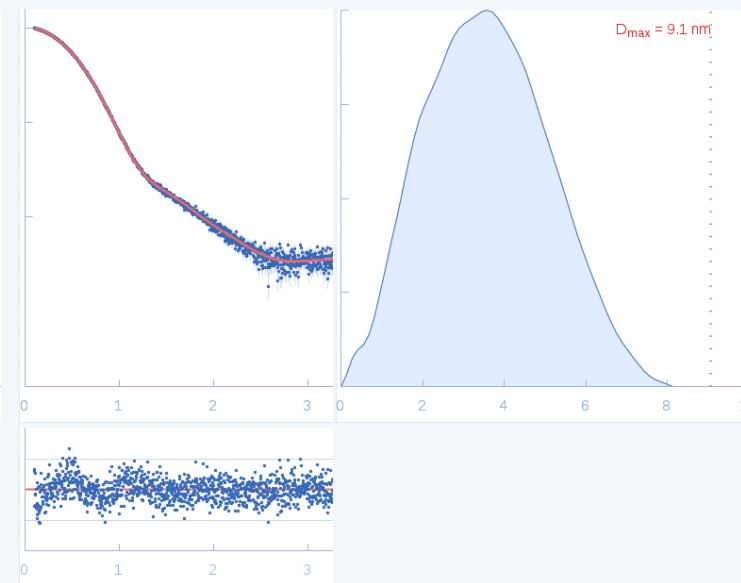
GNNOM

SASDFQ8

Predicted R_g 2.8 nm
 $p(r)$ R_g 2.8 nm
 autorg R_g 2.7 nm

Predicted D_{max} 9.1 nm
 $p(r)$ D_{max} 8.2 nm
 Predicted MW 56.7 kDa
 χ^2 1.1

SASDFQ8 – BSA standard



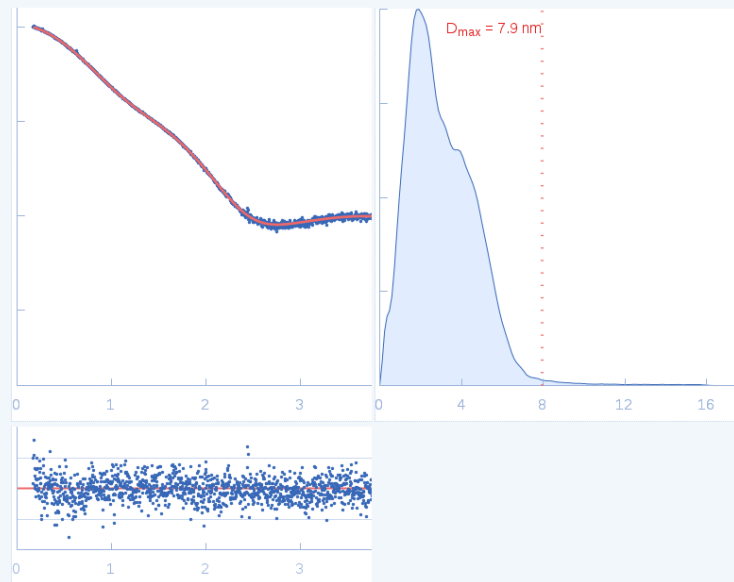
GNNOM

SEC-LIG-Jan16-clo

Predicted R_g 24.5 Å
 $p(r)$ R_g 24.9 Å
 autorg R_g 23.9 Å

Predicted D_{max} 79.3 Å
 $p(r)$ D_{max} 16.5 Å
 Predicted MW 19.3 kDa
 χ^2 3.5

PDZ domain (extended)



Download [SEC-LIG-Jan16-clo.fit](#)
 Download [pdf-predicted.dat](#)

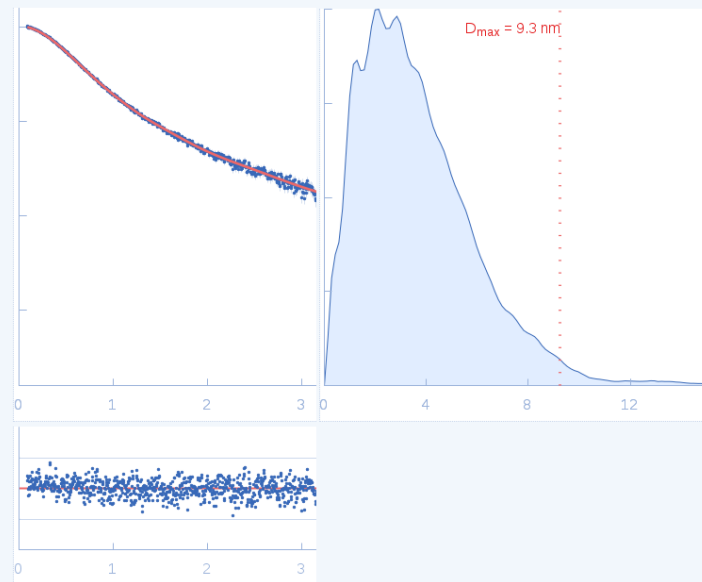
GNNOM

Colicin-N

Predicted R_g 3.0 nm
 $p(r)$ R_g 3.0 nm
 autorg R_g 2.8 nm

Predicted D_{max} 9.3 nm
 $p(r)$ D_{max} 15.8 nm
 Predicted MW 19.0 kDa
 χ^2 0.7

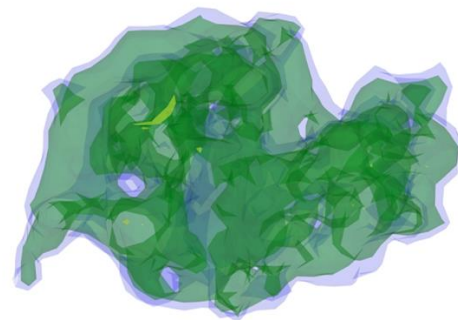
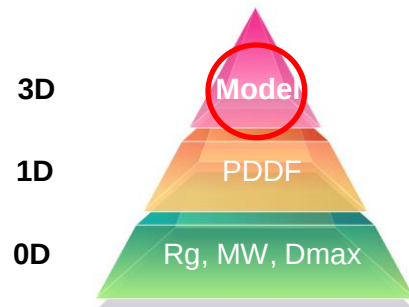
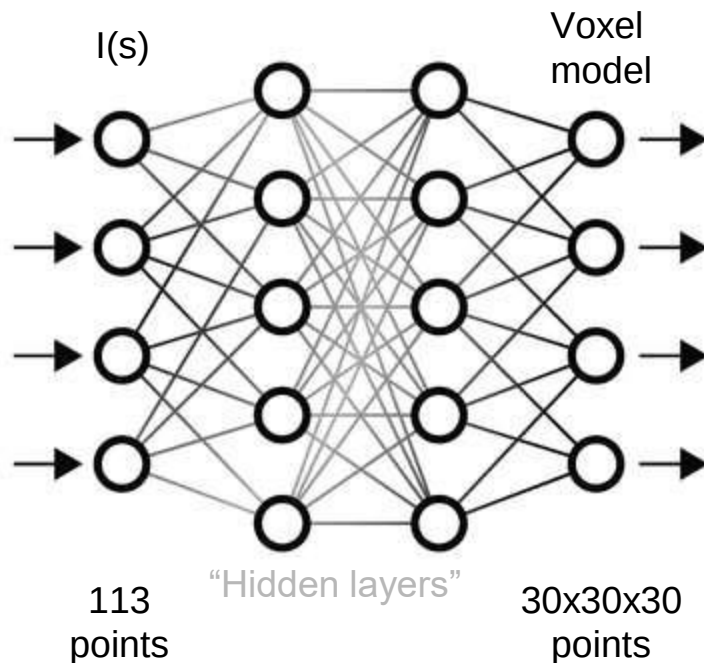
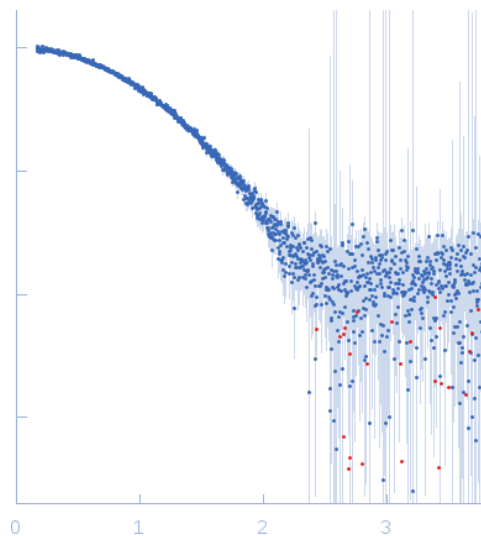
SASDC53 – Colicin N (IDP)



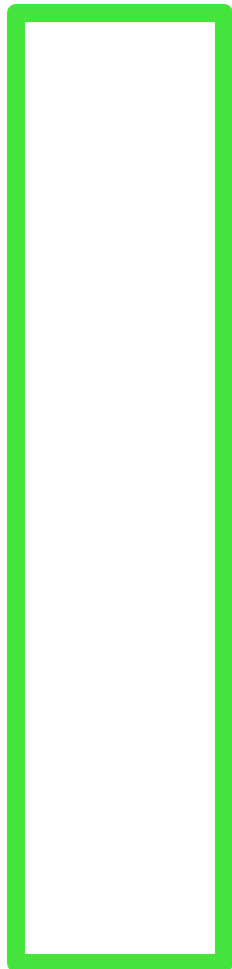
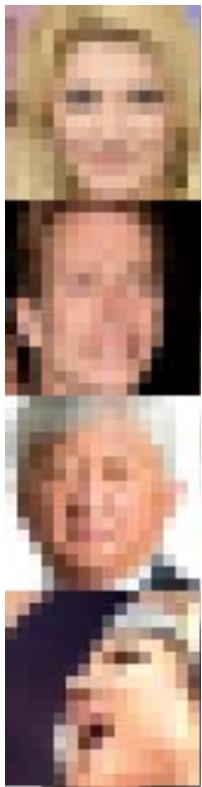
Download [Colicin-N.fit](#)
 Download [pdf-predicted.dat](#)

What's next?

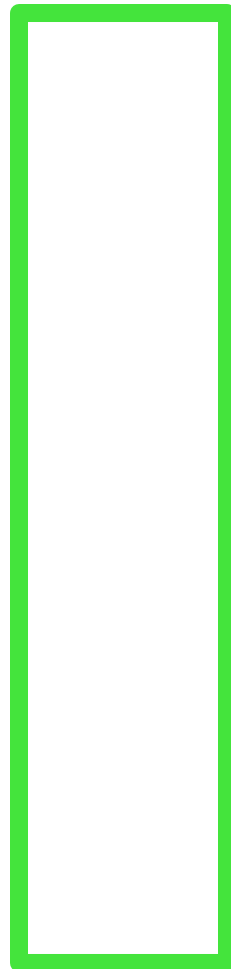
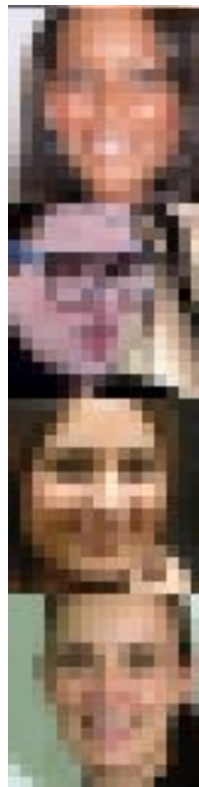
Modelling! (3D)



Input



Input



Group members

[Dima Molodenskiy](#)

[Al Kikhney](#)

[Dmitri Svergun](#)

[Clément Blanchet](#)

[Cy Jeffries](#)

[Daniel Franke](#)

[Melissa Gräwert](#)

[Haydyn Mertens](#)

[Clemente Borges](#)

[Stefano Da Vela](#)

[Tobias Gräwert](#)

[Andrey Gruzinov](#)

[Martin Schroer](#)

[Karen Manalastas](#)

[Taja Cheremnykh](#)



EMBL



Keras



TensorFlow



Thank you!

<https://dara.embl-hamburg.de/gnnom.php>

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