

Introduction

As DNA sequencing becomes increasingly routine in clinics for diagnosis and prescribing treatment of diseases, the problem of accurately characterizing the effects of genetic variants is becoming ever more important. Classification standards, such as the American College of Medical and Genomics standard, have been proposed to classify genetic variants ranging from “pathogenic” to “benign”¹. However, the effects for most variants are unknown and have thus been classified as “variants of unknown significance” (VUS). Even more poorly understood are the gene products that manifest from these variants and the mechanisms by which they carry out their effects.

In the visual system, genetic variation in the *RHO* gene which encodes for the visual pigment rhodopsin is the most common cause of retinitis pigmentosa (RP), a degenerative retinal disease. Over 150 different mutations in *RHO* have been identified² and more are being discovered every year. However, many of these variants are VUS because our ability to understand their effects is being outpaced by the rate that new variants are being discovered.

To keep pace with the increasing number of VUS requires a systematic approach that can be easily scaled and is able to classify variants accurately, which is difficult to achieve with a light activated protein. Recent advances in sequencing technologies, however, have accelerated the development of deep scanning mutagenesis, a promising solution to this dilemma that offers the potential for rapid classification of the effects for thousands of variants in a single experiment.

In this study, we aim to develop and apply a deep scanning mutagenesis approach that combines fluorescence-activated cell sorting and deep sequencing to help classify rhodopsin variants and to determine the mechanisms by which these variants confer their effects.

An engineered yeast system for measuring rhodopsin activation

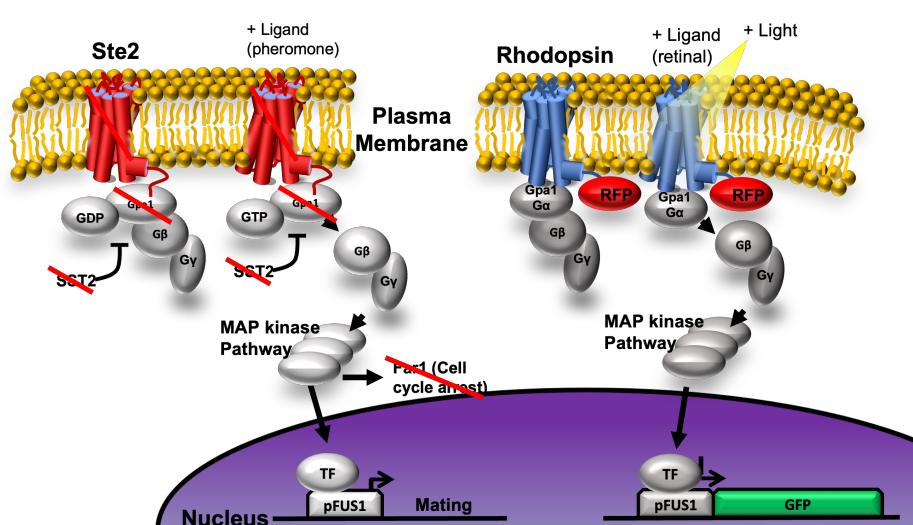


Figure 1. Heterologous expression of rhodopsin in an engineered yeast strain to enable light-dependent activation of the yeast mating pathway. (a) The yeast mating pathway. (b) Rhodopsin activates an engineered yeast mating pathway which activates a GFP fluorescent reporter. Rhodopsin is also fused to ymScarlet fluorescent protein to track its expression and localization in yeast cells.

A variant library for testing all possible missense variants within a functionally important region in rhodopsin

- The variant library comprises of 1,976 missense variants where each of the 104 sites in the target region was mutated to all 19 other amino acids
- Extracellular loop 2 (EL2) is a major hotspot where RP mutations are known to cluster
- Helix 5 (H5), Intracellular Loop 3 (ICL3), and Helix 6 (H6) undergo pronounced conformational shifts during receptor activation
- ICL3 are within contact distance with downstream $G\alpha$ G protein

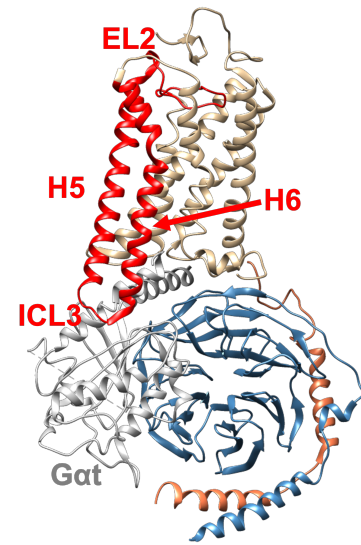


Figure 2. Structure of rhodopsin in complex with G proteins.

Rhodopsin activation and protein abundance varies widely among variants

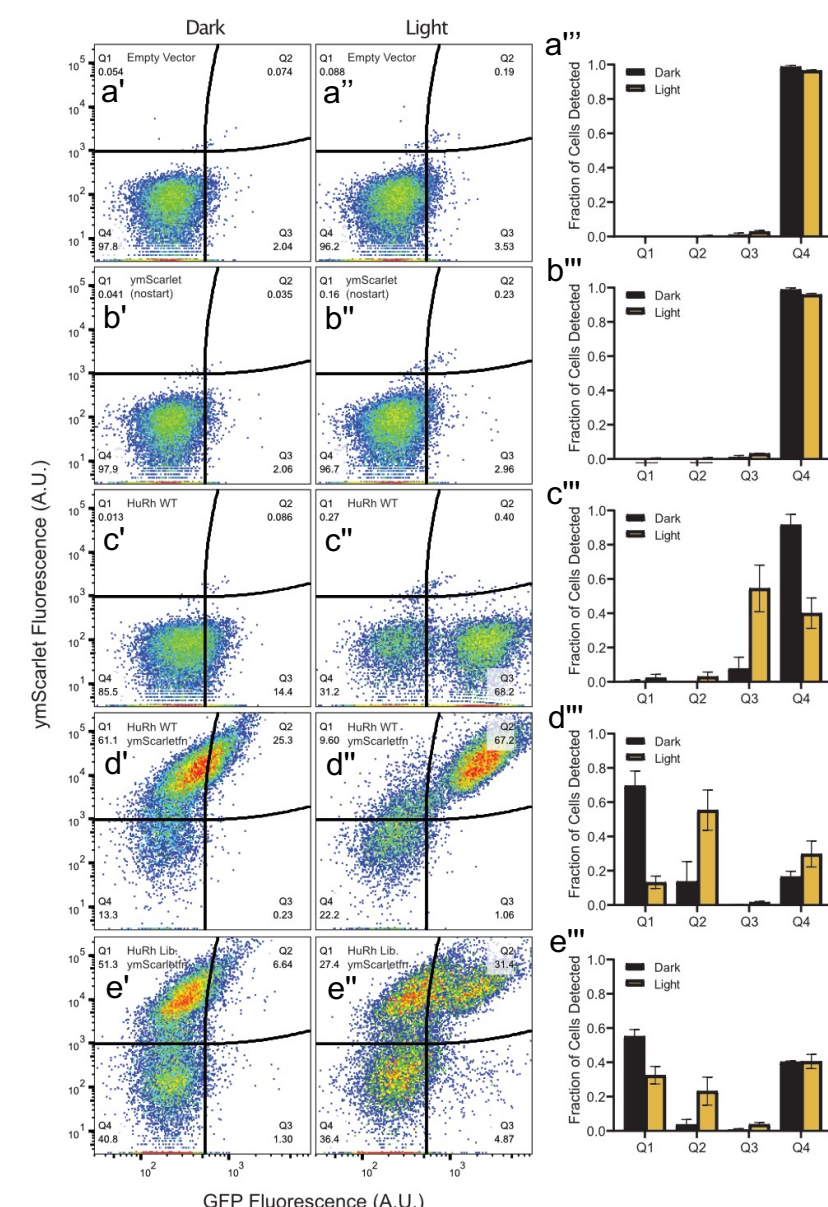


Figure 3. Per cell rhodopsin activation (GFP) and abundance (ymScarlet) under light vs. dark conditions. (a) Engineered yeast not expressing rhodopsin in the dark (a'), light (a''), and fraction of cells detected in each quadrant (a''). (b-e) Same as in (a) except (b) contains ymScarlet with no start codon, (c) is expressing wild type human rhodopsin (HuRh WT), (d) is expressing wild type human rhodopsin fused to ymScarlet (HuRh WT ymScarlet), and (e) is expressing the library of rhodopsin variants fused to ymScarlet (HuRh Lib. ymScarlet).

For most variants, sufficient receptor abundance is required to elicit detectable signaling

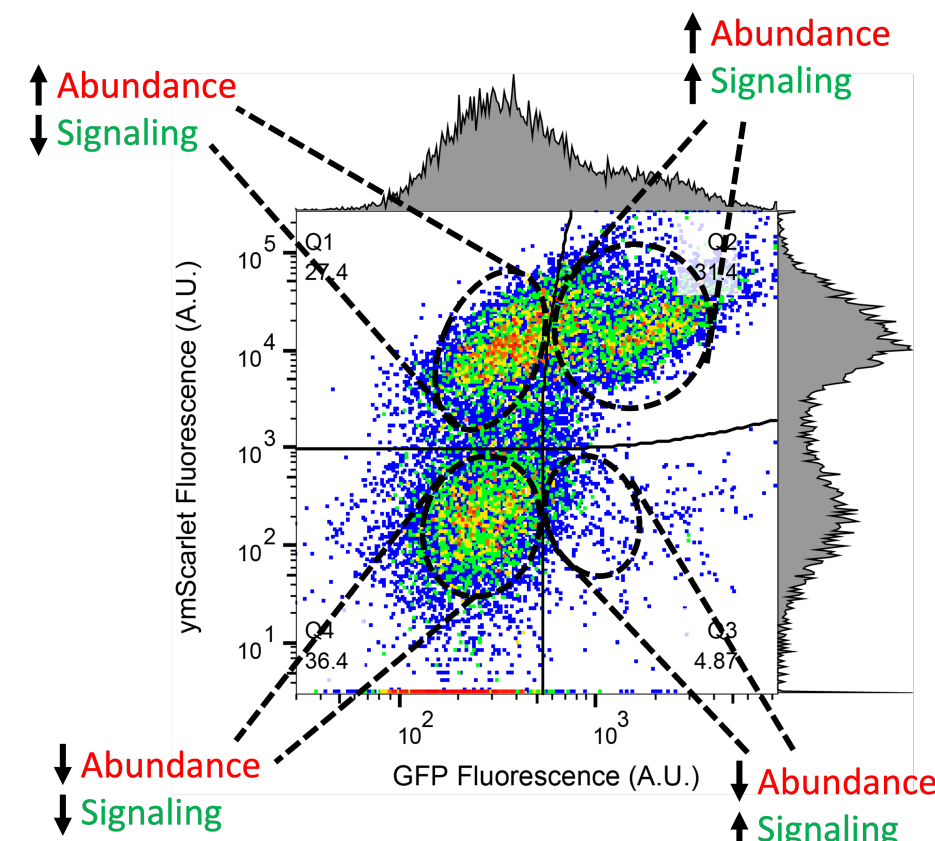


Figure 4. Two-dimensional plot of rhodopsin activation (GFP) and per cell protein abundance (ymScarlet) when our variant library is expressed and light activated in our yeast system.

Imaging rhodopsin in yeast to distinguish misfolded variants from those that fold but do not signal

- Misfolded proteins in yeast are known to form aggregates³
- Can we use differences in membrane heterogeneity to distinguish misfolded variants from those that fold but do not signal?

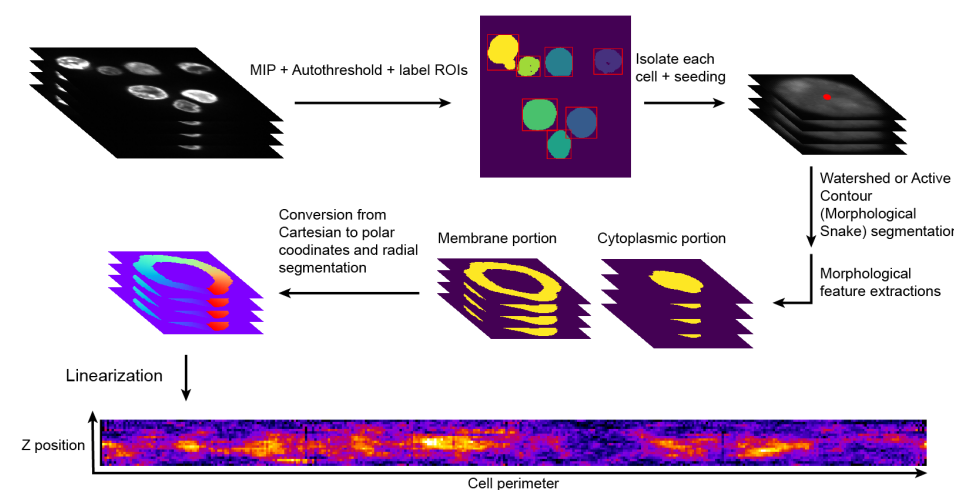


Figure 5. Image processing workflow to segment the plasma membrane of yeast cells that are expressing rhodopsin variants. Maximum intensity projection of the input image is generated, used to create a binary mask of the image by thresholding. Each object in the binary mask is labelled and isolated using its bounding rectangle. Each of the isolated cells is segmented and the resulting segmentation results are separated into binary masks of membrane or the cytoplasmic regions. The membrane mask undergoes further processing where it is linearized and stacked along the depth of the cell, creating a 2D visualization of the protein distribution at the surface.

Well-characterized misfolded variants are heterogeneously expressed on the cell surface

- P23H and V81del amino acid substitutions are known to cause rhodopsin to misfold⁴ while N200I is characterized to be superactive (Chen et al. in prep)

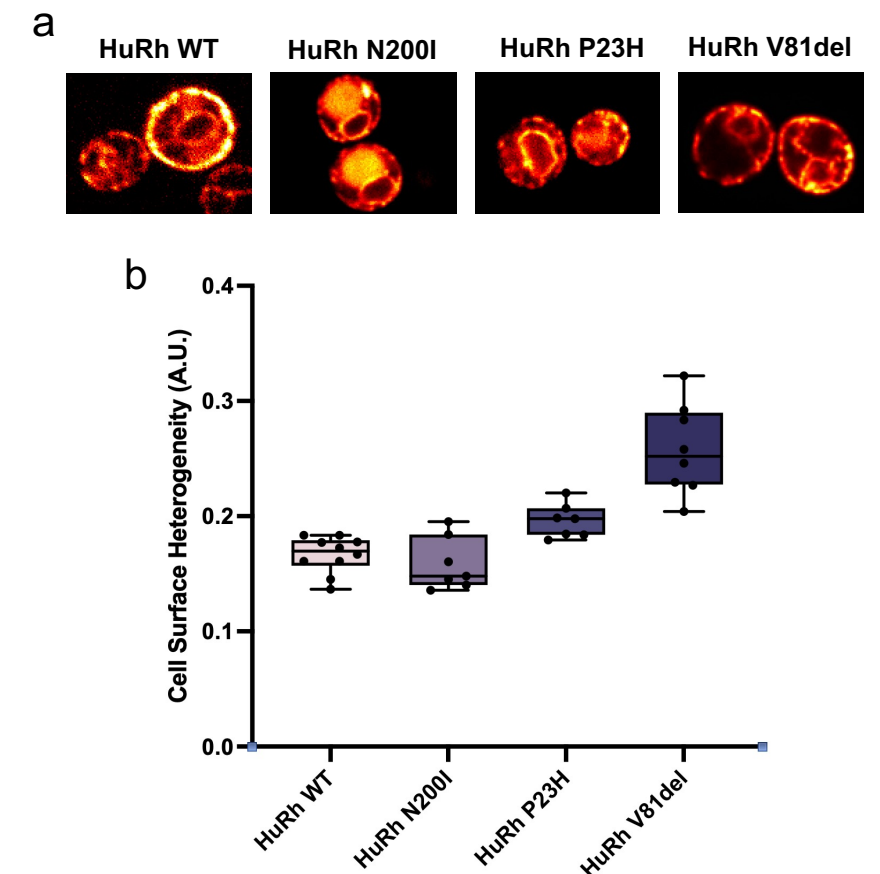


Figure 6. Cell surface heterogeneity for well-characterized variants. (a) Representative confocal micrographs for wild type human rhodopsin (HuRh WT) and three human rhodopsin variants carrying N200I, P23H, and V81del substitutions respectively. (b) Cell surface heterogeneity measurements for the variants in (a).

Future directions

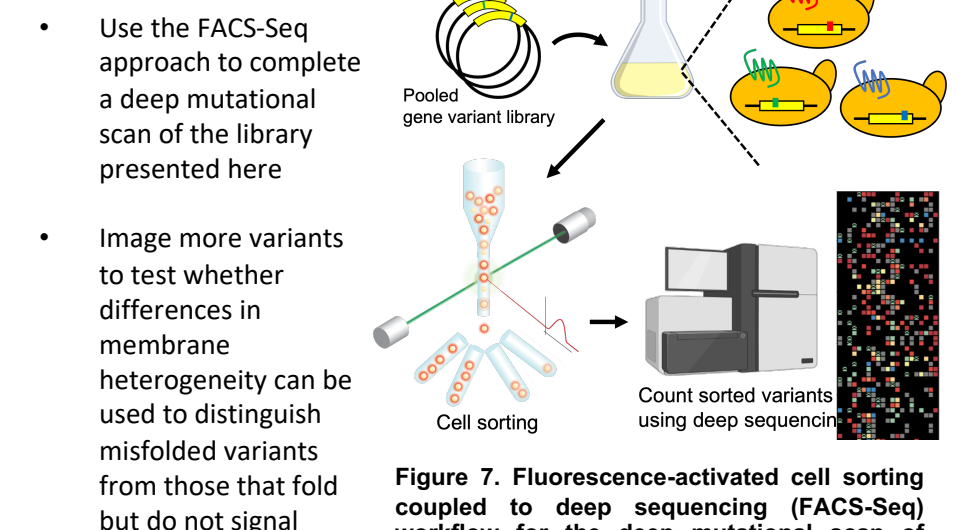


Figure 7. Fluorescence-activated cell sorting coupled to deep sequencing (FACS-Seq) workflow for the deep mutational scan of protein variants.