# Predicting the disease phenotypes of *de novo* ATP1A3 variants

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#### 1 Generating PROVEAN and SIFT scores for variants

The following code cell reads the spreadsheet containing the ATP1A3 mutations and generates an output file compatible with PROVEAN. file\_reader reads an input delim-delimited file and generated an ordered dictionary.

PROVEAN requires input data to be space-delimited and formatted like this: <Protein\_ID> <position> <reference\_residue(s)> <variant\_residue(s)> <comment (optional)>. Multiple variants can be analyzed at once if they are line-delimited. To achieve appropriate formatting, the provean\_input\_formatter function has been implemented. Finally, an output .txt file is generated by the function output\_writer. This file can be directly uploaded to the online PROVEAN server.

```
In [1]: import csv
        NCBI_ref_id = "NP_689509.1"
        def file_reader(infile, delim):
            data = []
            with open(infile, "r") as fh:
                for row in csv.DictReader(fh, delimiter = delim):
                    data.append(row)
            return data
        def provean_input_formatter(1):
            provean_input = []
            for variant in 1:
                temp = []
                if variant["type"] == "Missense_Mutation":
                    ref_aa = variant["substitution"][0]
                    alt_aa = variant["substitution"][-1]
                    pos = variant["substitution"][1:-1]
                    disease = variant["disease"]
                    temp = [NCBI_ref_id, pos, ref_aa, alt_aa, disease]
                    provean_input.append(temp)
                else:
                    pass
            return provean_input
        def output_writer(l, outfile, delim):
            with open(outfile, "w") as fh:
```

### 2 Making a feature matrix

The tabulated PROVEAN output is then related back to the originial mutation data, so that disease outcome and protein domain data can be added. The dictonary domains maps certain amino acid positions to NCBI-annotated protein regions, properties maps single letter amino acid codes to a set of chemical properties, and diseases maps amino acid changes (in the format A[pos]B, where A is the reference residue and B is the altered residue) to disease phenotypes according to published data.

```
In [2]: '''
        NCBI_domains = {37: "phosphorylation site",
                   56: "phosphorylation site",
                   range(72,75): "interaction with phosphoinositide-3 kinase",
                   range(78,99): "transmembrane region",
                   range(122,143): "transmembrane region",
                   218: "phosphorylation site",
                   265: "phosphorylation site",
                   range(279,299): "transmembrane region",
                   range(311,329): "transmembrane region",
                   442: "phosphorylation site",
                   548: "phosphorylation site",
                   range(763,783): "transmembrane region",
                   range(834,857): "transmembrane region",
                   range (909, 929): "transmembrane region",
                   933: "phosphorylation site by PKA",
                   range (942, 961): "transmembrane region",
                   range(976,997): "transmembrane region"}
         , , ,
        pfam_domains = {range(33,102): "cation_ATPase_N",
                        range(153,345): "E1-E2_ATPase",
                        range(416,512): "cation_ATPase",
                        range(789,999): "cation_ATPase_C"}
        properties = {"R": "charged",
                       "K": "charged",
                       "D": "charged",
                       "E": "charged",
                       "Q": "polar",
                       "N": "polar",
```

```
"H": "polar",
              "S": "polar",
              "T": "polar",
              "Y": "polar",
              "C": "polar",
              "W": "polar",
              "A": "hydrophobic",
              "I": "hydrophobic",
              "L": "hydrophobic",
              "M": "hydrophobic",
              "F": "hydrophobic",
              "V": "hydrophobic",
              "P": "hydrophobic",
              "G": "hydrophobic"}
def get_domain(position):
    for key in pfam_domains.keys():
        if position in key:
            return pfam_domains[key]
    return "unannotated domain"
def get_chemical_properties(amino_acid):
    return properties[amino_acid]
def prop_features(prop):
    if prop == "charged":
        return [1, 0, 0]
    elif prop == "polar":
        return [0, 1, 0]
    elif prop == "hydrophobic":
        return [0, 0, 1]
def ncbi_domain_features(domain):
    if domain == "phosphorylation site":
        return [1, 0, 0, 0]
    elif domain == "interaction with phosphoinositide-3 kinase":
        return [0, 1, 0, 0]
    elif domain == "transmembrane region":
        return [0, 0, 1, 0]
    elif domain == "phosphorylation site by PKA":
        return [0, 0, 0, 1]
def pfam_domain_features(domain):
    if domain == "cation_ATPase_N":
        return [1, 0, 0, 0, 0]
    elif domain == "E1-E2_ATPase":
        return [0, 1, 0, 0, 0]
    elif domain == "cation_ATPase":
```

```
return [0, 0, 1, 0, 0]
    elif domain == "cation_ATPase_C":
        return [0, 0, 0, 1, 0]
    elif domain == "unannotated domain":
        return [0, 0, 0, 0, 1]
def provean_sift_features(provean, sift):
    if provean < -2.5:
        p_pred = 1
    else:
        p_pred = 0
    if sift > 0.05:
        s_pred = 1
    else:
        s_pred = 0
    return [provean, p_pred, sift, s_pred]
diseases = {}
for variant in variants:
    diseases[variant["substitution"].strip()] = variant["disease"]
provean_output = file_reader("provean_output.tsv", "\t")
```

Here, I define the custom object Variant, which contains a method get\_feature\_vector which constructs a 16-dimensional feature vector for that variant. These feature vectors are stored in the list feature\_vectors, and their corresponding disease state is stored in the list disease\_list.

```
In [3]: class Variant():
            def __init__(self, ref_aa, pos, alt_aa, provean, sift):
                self.ref_aa = ref_aa
                self.alt_aa = alt_aa
                self.pos = pos
                self.domain = get_domain(pos)
                #self.domain_features = pfam_domain_features(self.domain)
                self.ref_aa_prop = get_chemical_properties(ref_aa)
                self.alt_aa_prop = get_chemical_properties(alt_aa)
                self.provean = provean
                self.sift = sift
                try:
                    self.disease = diseases[ref_aa + str(pos) + alt_aa]
                except KeyError:
                    self.disase = "null"
            def get_feature_vector(self):
                aa_prop_vector = prop_features(self.ref_aa_prop) + prop_features(self.alt_aa_prop_seatures)
                domain_vector = pfam_domain_features(self.domain)
                provean_sift_vector = provean_sift_features(self.provean, self.sift)
```

### 3 Training the Support Vector Classifier (SVC)

## 4 Making predictions

```
In [5]: pos = input("enter position of the amino acid variant: ")
    ref_aa = input("enter the single letter code of the reference amino acid: ")
    alt_aa = input("enter the single letter code of the altered amino acid: ")
    provean = input("enter the PROVEAN score of the variant: ")
    sift = input("enter the SIFT score of the variant: ")

In [6]: v = Variant(ref_aa, int(pos), alt_aa, float(provean), float(sift))
    print("predicted disease phenotype: " + classifier.predict([v.get_feature_vector()])[0])

predicted disease phenotype: AHC
```