

Risk Analysis of PFAS Contamination in Private Water Wells: a Bayesian Network Model

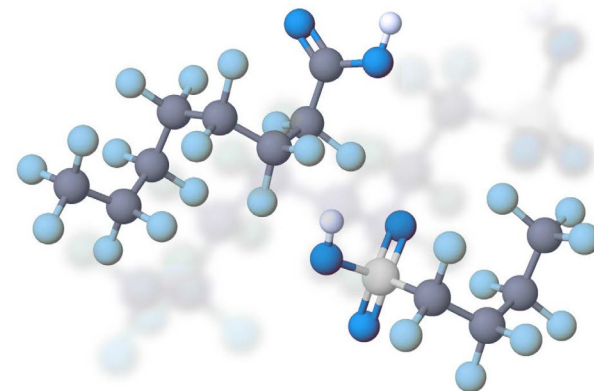
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10/10/2019

Outline

□ Introduction

- PFAS, forever chemicals
- PFAS in NC private wells
- Objectives



□ Methods

- Bayesian network model development
- Model evaluation and cross-validation

□ Results

- Insights on risk factors
- Bayesian network model accuracy

□ Conclusion

- Intervention to decrease risk in private wells



Introduction

Per- and polyfluoroalkyl substances (PFAS)

PFAS are found in many consumer products



<http://www.sixclasses.org/>

Forever chemicals!!!



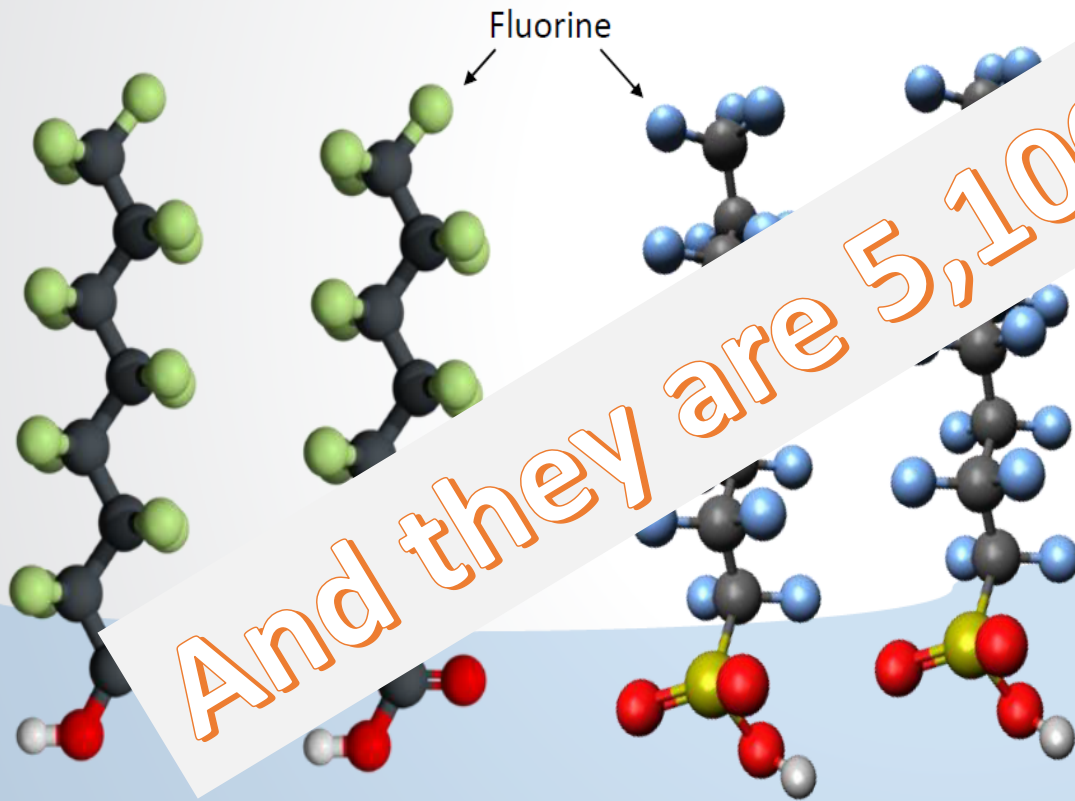
Per- and Polyfluoroalkyl Substances (PFAS)

A class

of carbon (C) atoms
surrounded by fluorine (F)
atoms

- Water-repellent (hydrophobic body)
- Stable C-F bond
- Some PFAS include oxygen, hydrogen, sulfur and/or nitrogen atoms, creating a polar end.

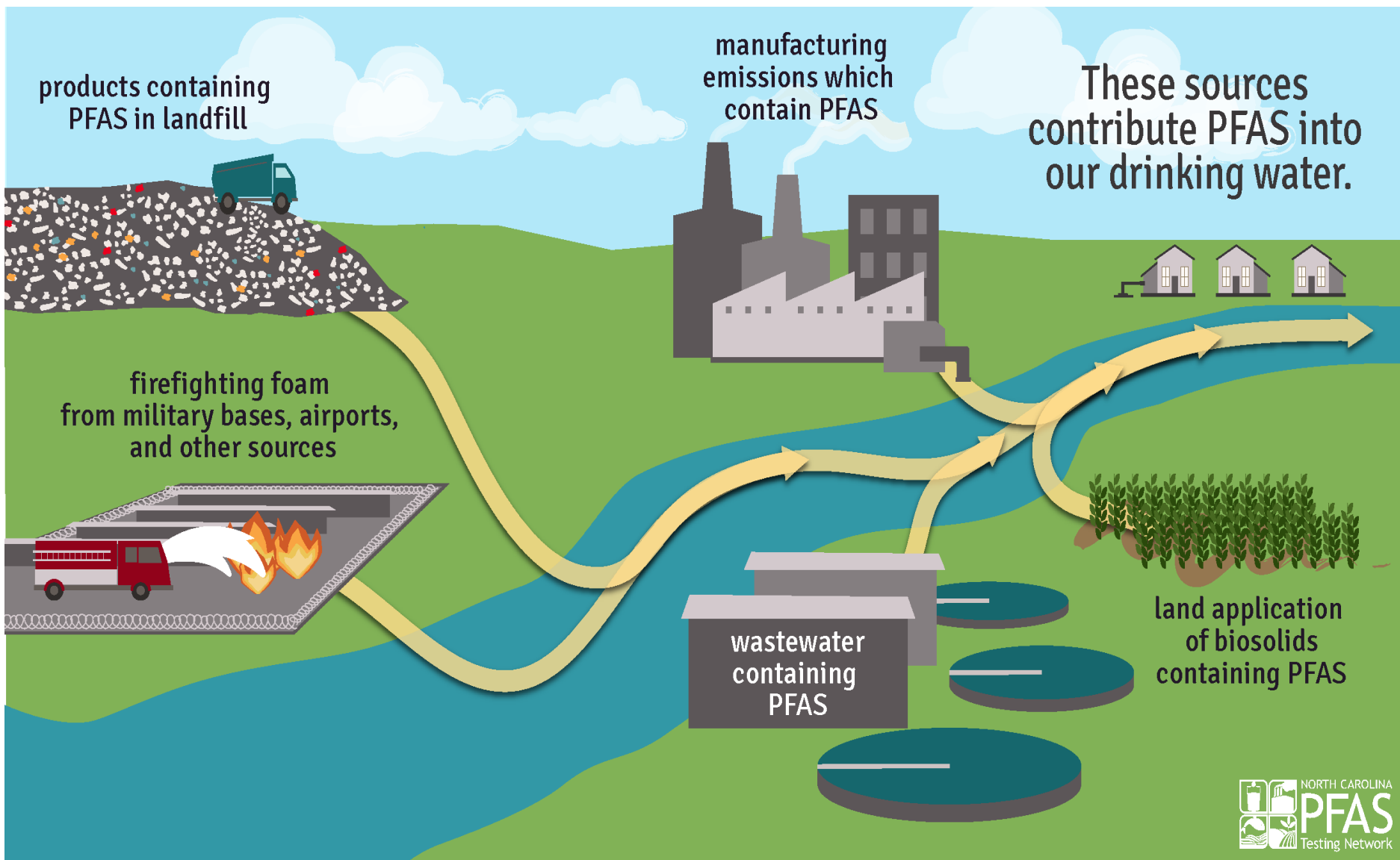
And they are 5,100 of them!!!



Perfluorooctanoic acid (PFOA)

Perfluorooctanesulfonic acid (PFOS)

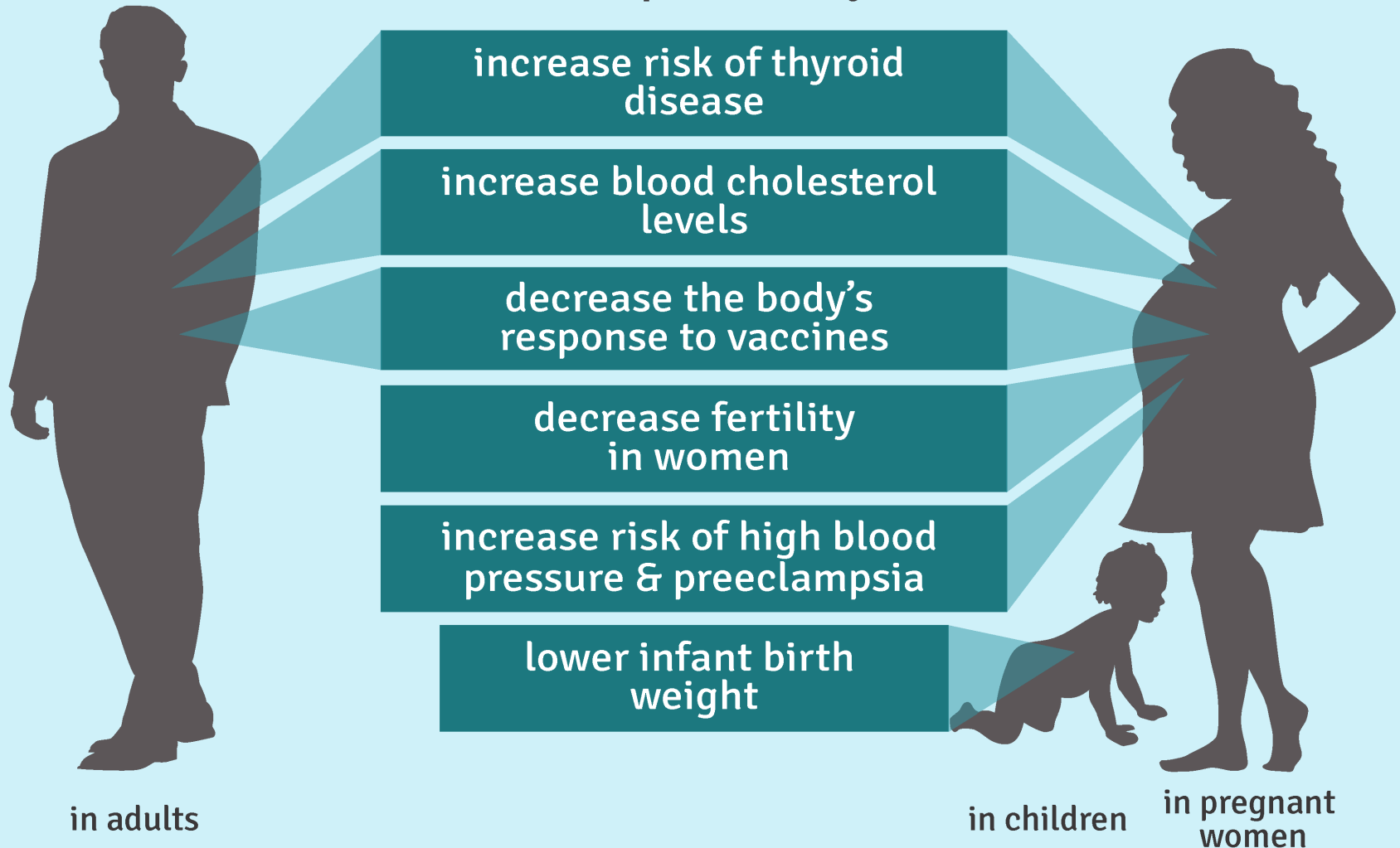
PFAS can reach to our drinking water!!



PFAS exposure and the health effects

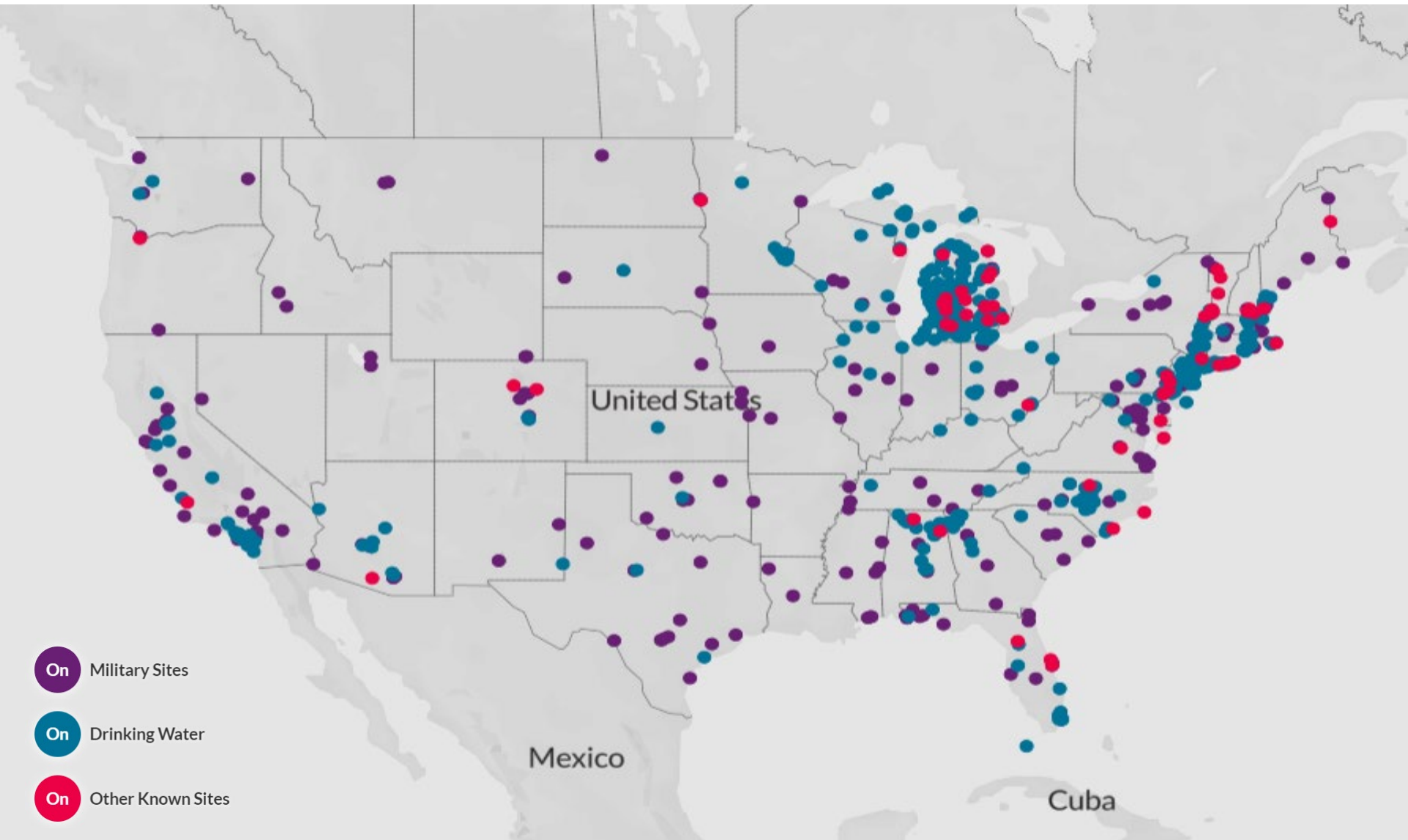


Human studies suggest
PFAS exposure may...



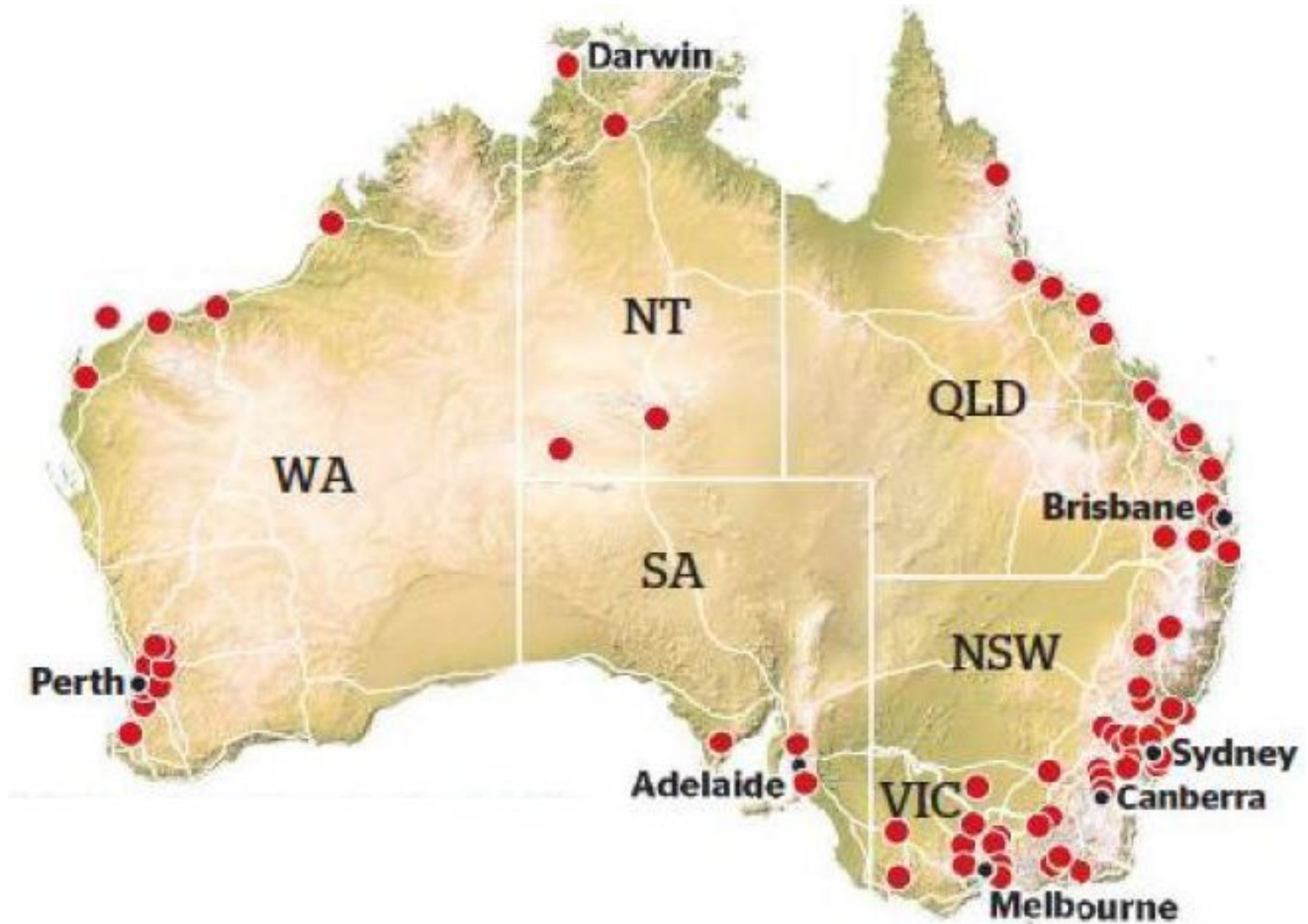
Information sourced from Agency for Toxic Substances and Disease Registry

It has been detected in many site in the US



<https://www.ewg.org/>

And PFAS can be found in other countries too !!!



<https://www.miltoncaine.com>

Private wells around the Chemours Chemical Plant are contaminated with PFAS

- During 2017-2018, PFAS (including GenX), were detected in more than 75% of 803 private water supply wells near the Chemours facility



[NC DEQ, 2017-18, Scruggs, 2019]

GenX concentrations vary widely from well to well

- 803 wells tested
 - 23% > health goal
 - 23% non-detect
- Map key
 - Red = > 140 ng/L
 - Yellow = 0- 140 ng/L
 - Green = non-detect

• Figure courtesy of DEQ



Objectives



- Overall: To uncover **factors** influencing the risk of PFAS contamination in water supply wells near the Chemours plant
- ❖ Determine why **some wells are contaminated** and others are not. What features of the wells, landscape, geology, weather, and geographic location influence risks to wells?
- ❖ Build model to **prioritize** private wells for future testing.
- ❖ Develop **user-friendly web site** to help private well owners assess risks.



Methods

Many variables may influence the PFAS showing up in a private water well

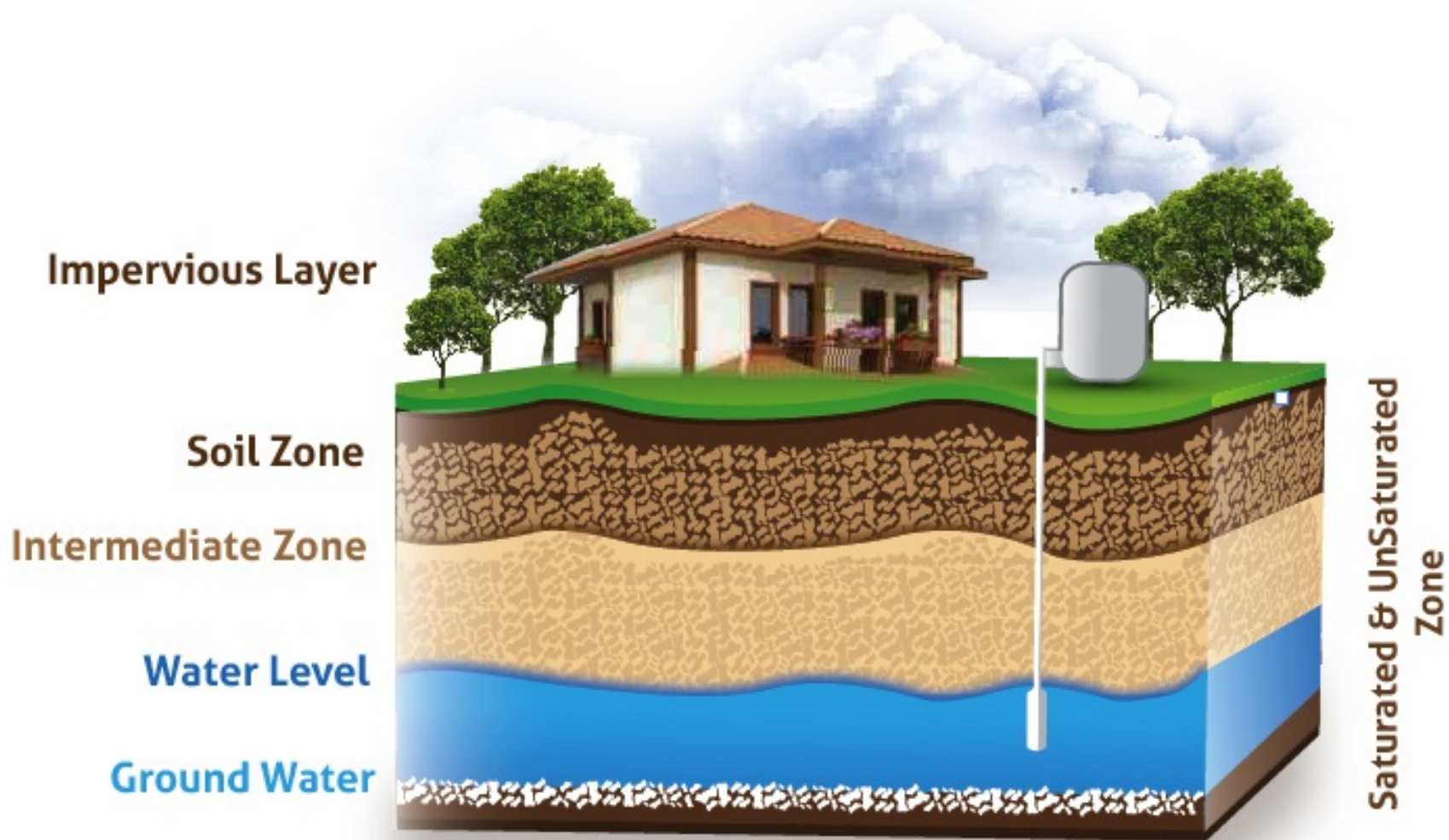


Image from: cleanwaterstore.com

We built a database of multiple factors that might influence GenX in well water

Tests from 803 Wells

Land Cover

Soil Surveys

Waste Sites
Landfills

Air Deposition
Model Output

240 potential predictors

Silt content
Soil Cation

Soil pH

House Age

Well Age

Well depth

Vegetation

Proximity to rivers

Clay content

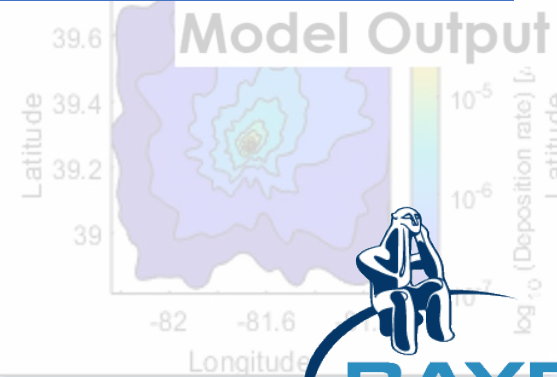
Our curated data set imported to BayesiaLab

Tests from 803 Wells

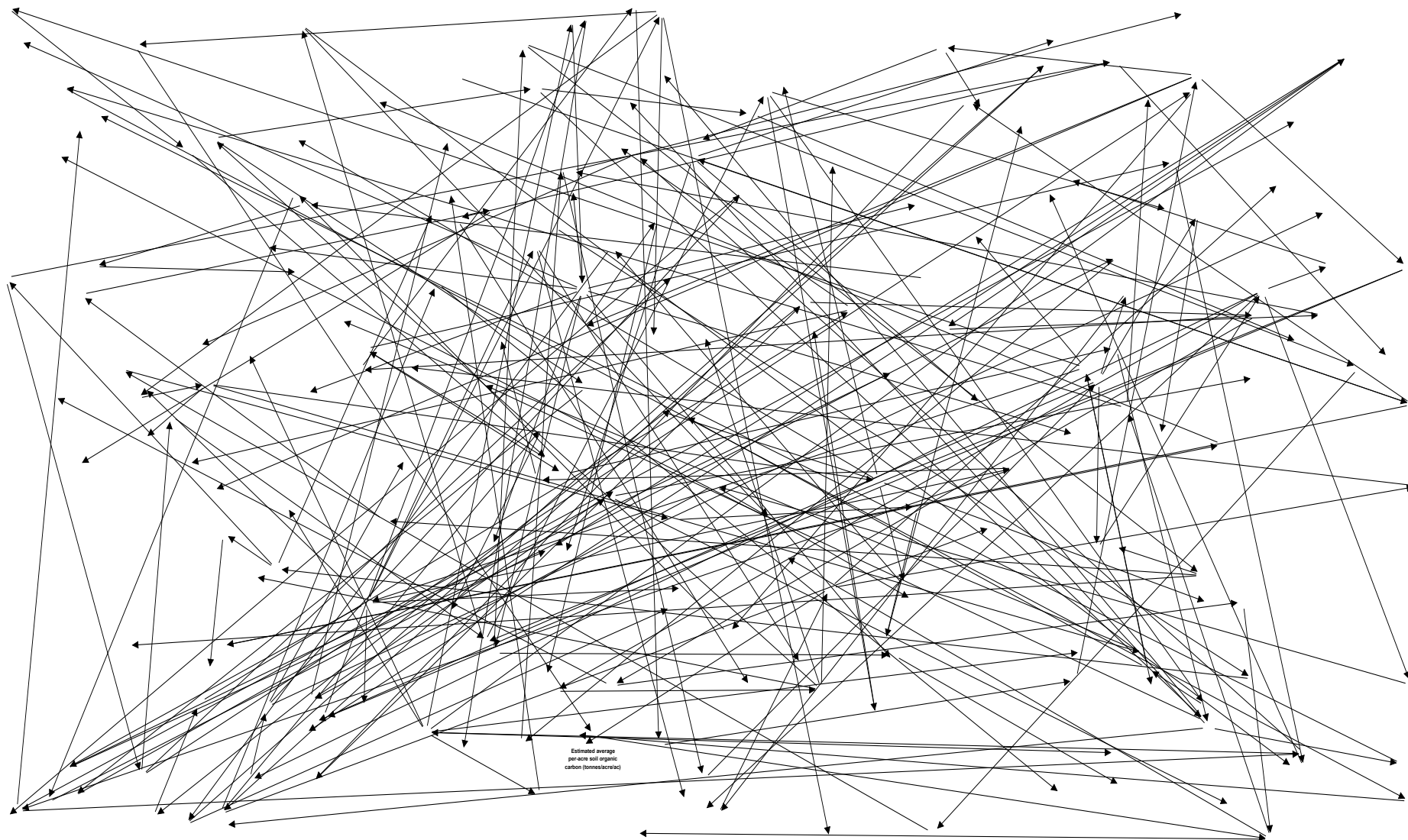
Land Cover

240 potential predictors

A matrix of
803 rows*240 columns



Finding influencing variables among this data set could be complicated



Multiple models built and tested Using BayesiaLab software

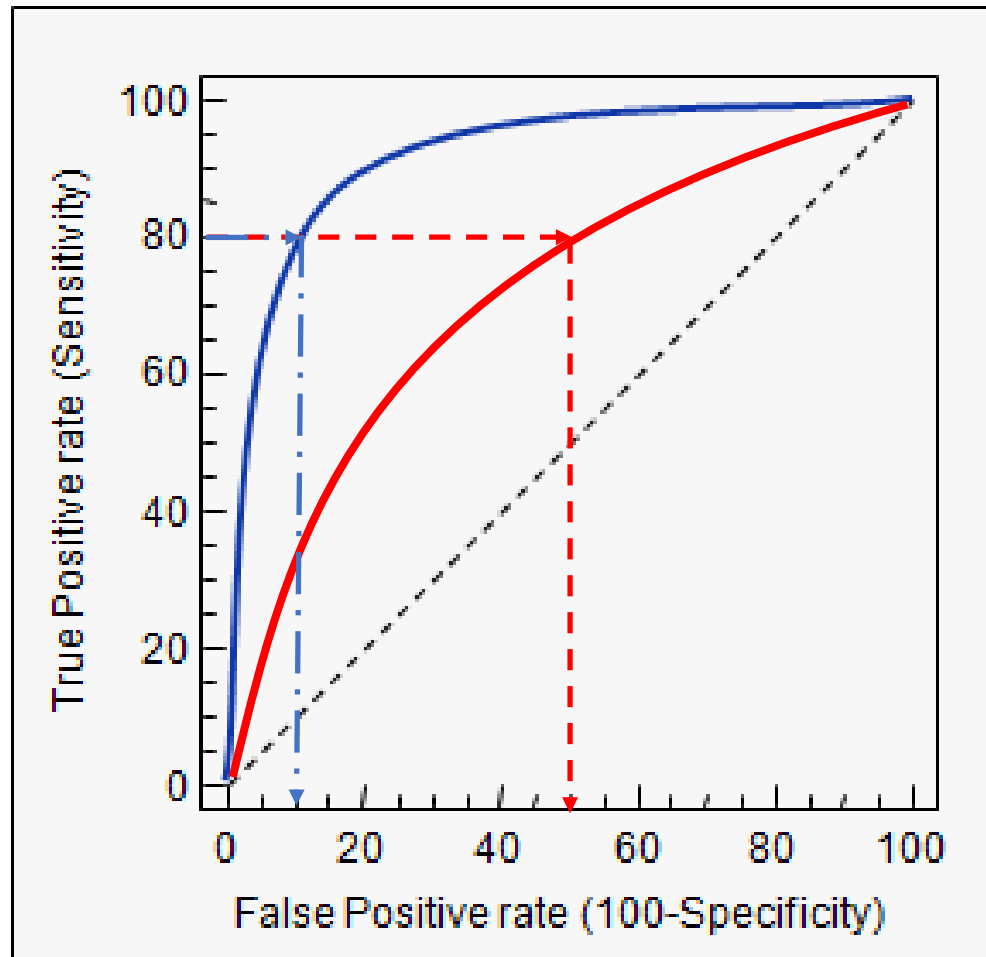
□ Modeling goals:

- Predict which untested wells are at risk of having GenX above NC health goal (140 ng/L)
- Identify factors most influencing risk

□ Model building steps (in BayesiaLab)

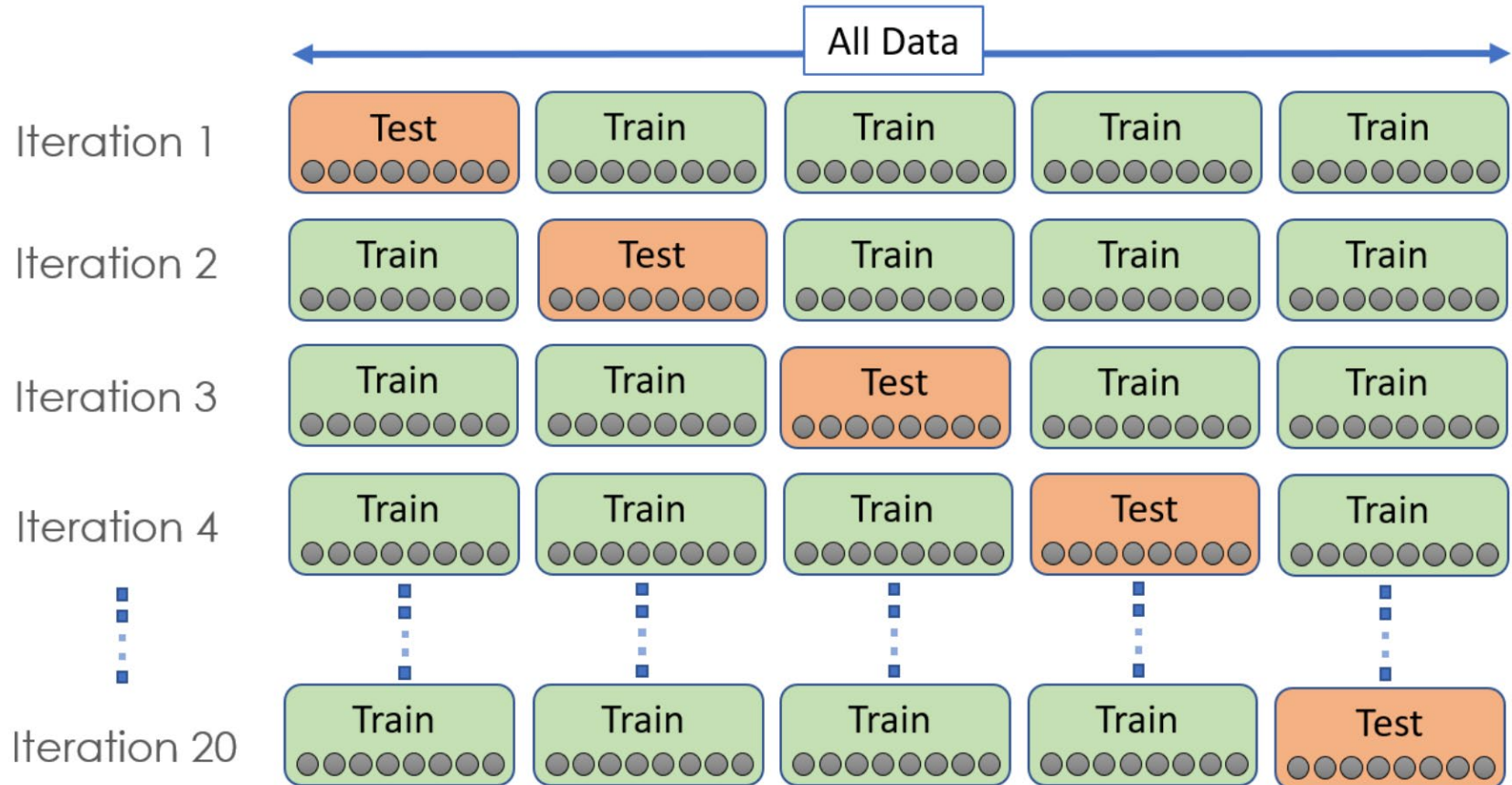
1. “**Unsupervised learning**” and “**variable clustering**” to discover which variables are most closely related to GenX
2. **Elimination** of unrelated variables
3. “**Supervised learning**” to build a predictive model
 - Augmented naïve Bayes algorithm often used for classification problems

Area Under Receiver-Operating Characteristics (ROC) curve used as performance metric



Prediction Accuracy = Area Under Curve (1.00 is perfect)

Accuracy tested in cross-validation



Five-fold cross Validation

Repeated 20 times

- Different random train/test split each time

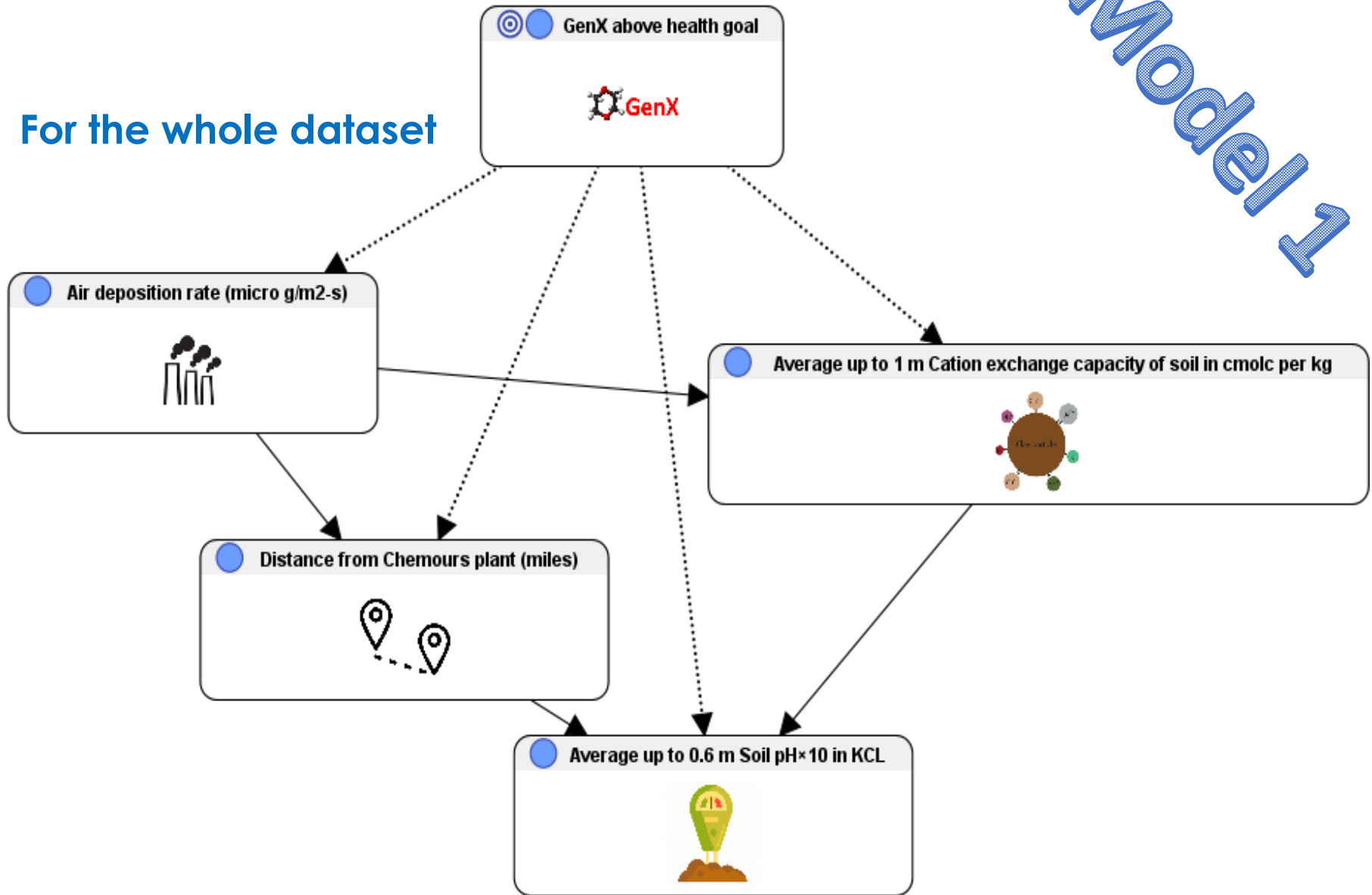


Results

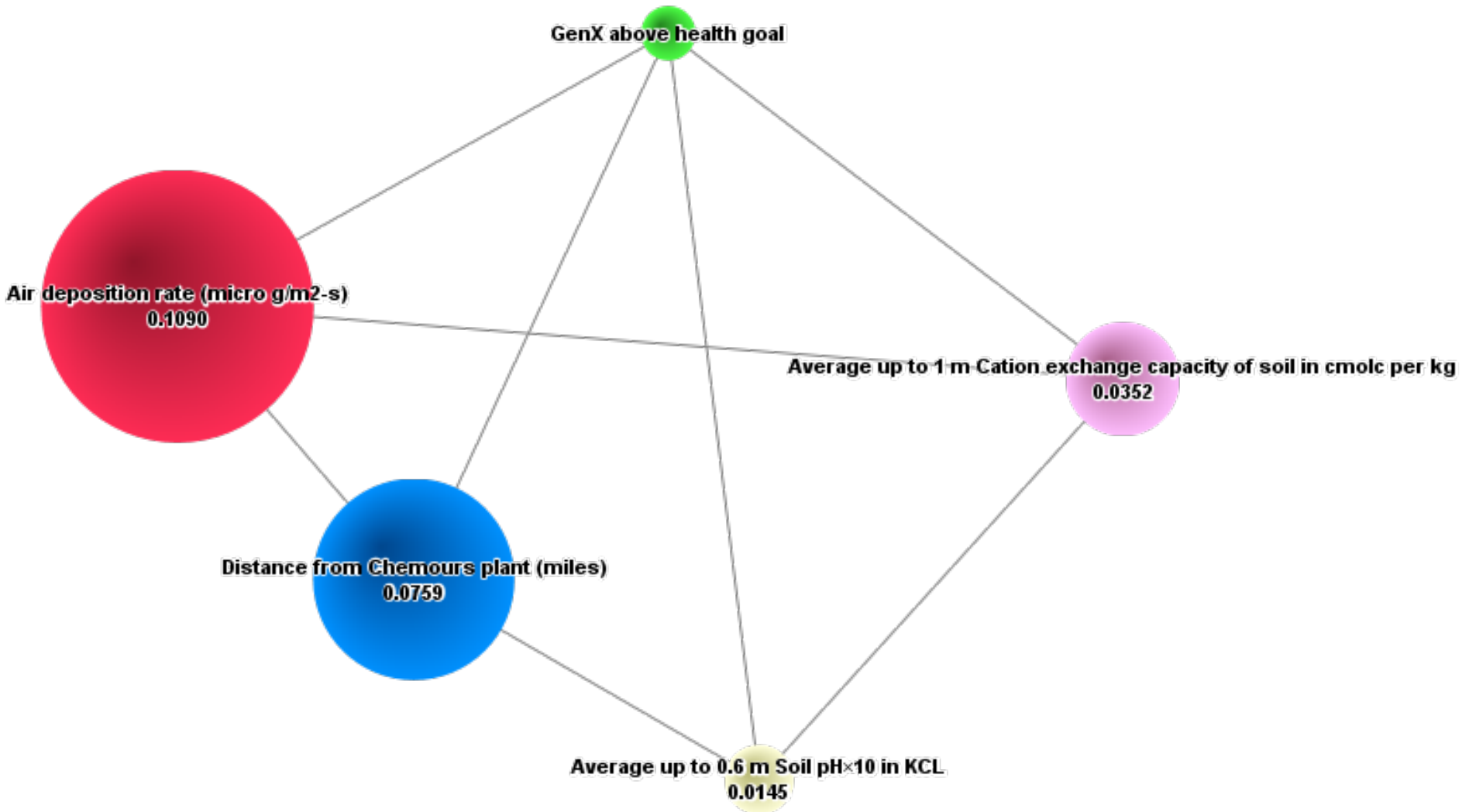
Main variables that influencing higher risk of GenX

Model 1

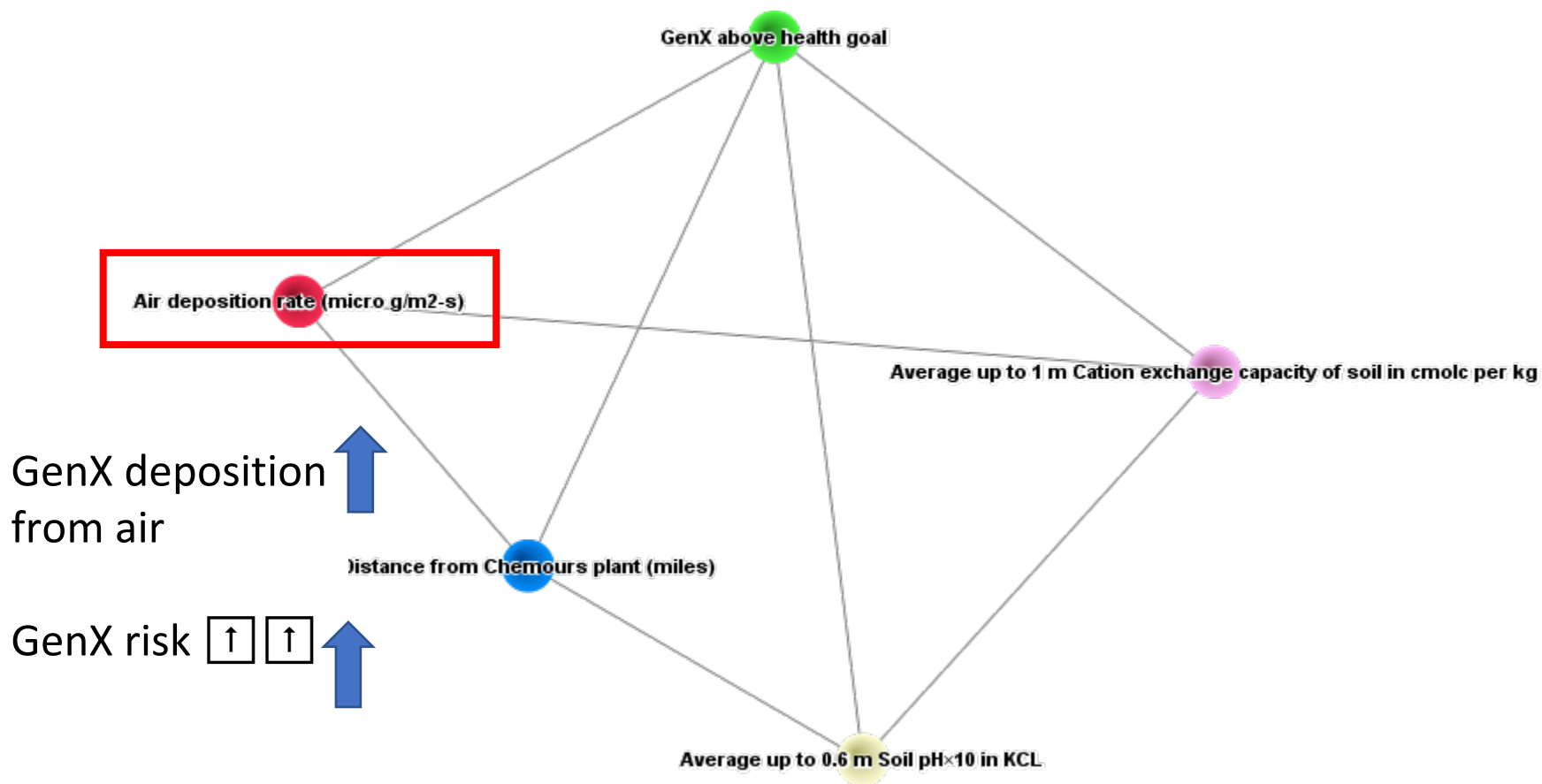
For the whole dataset



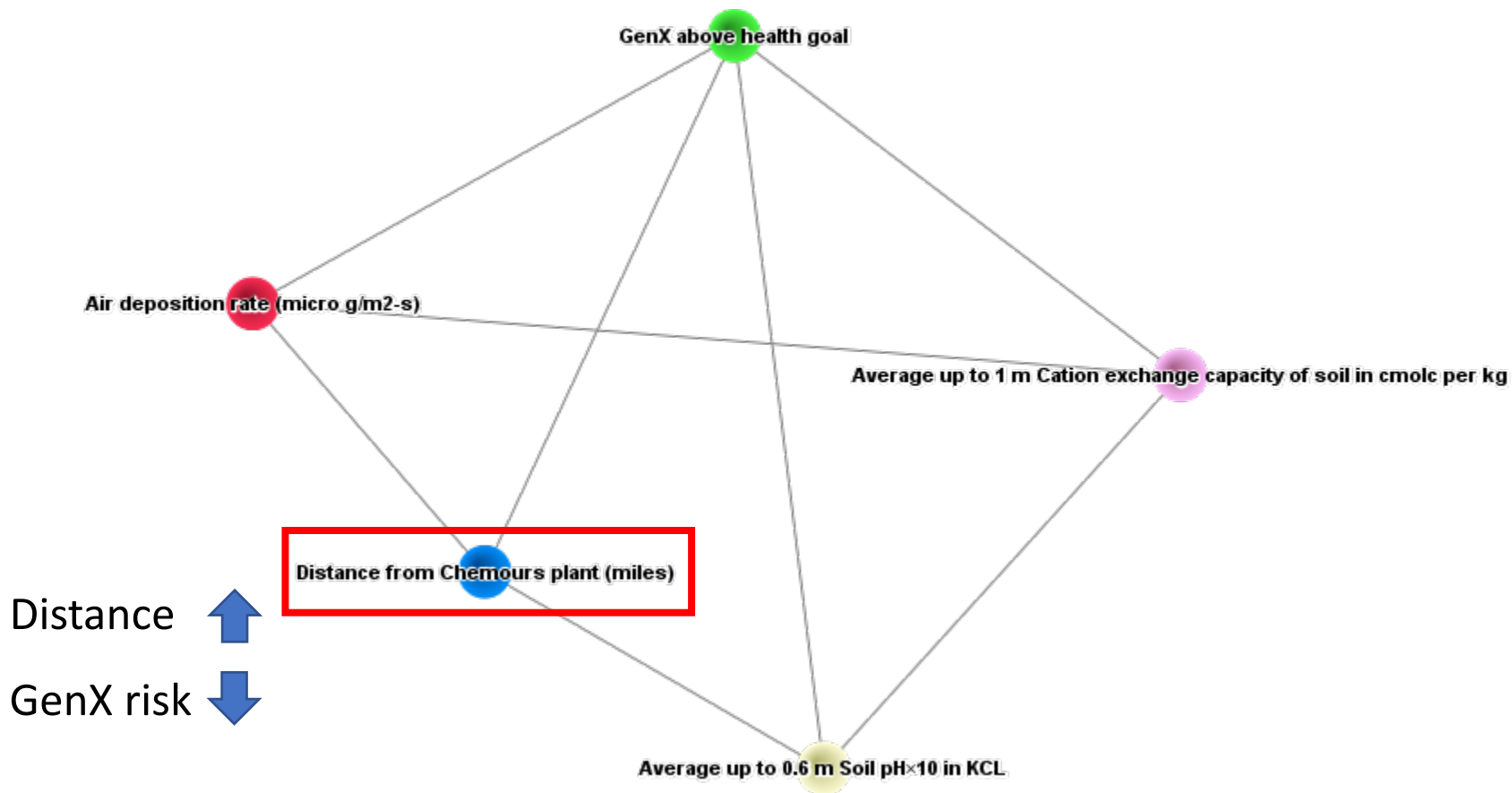
Mutual information with the GenX as target node



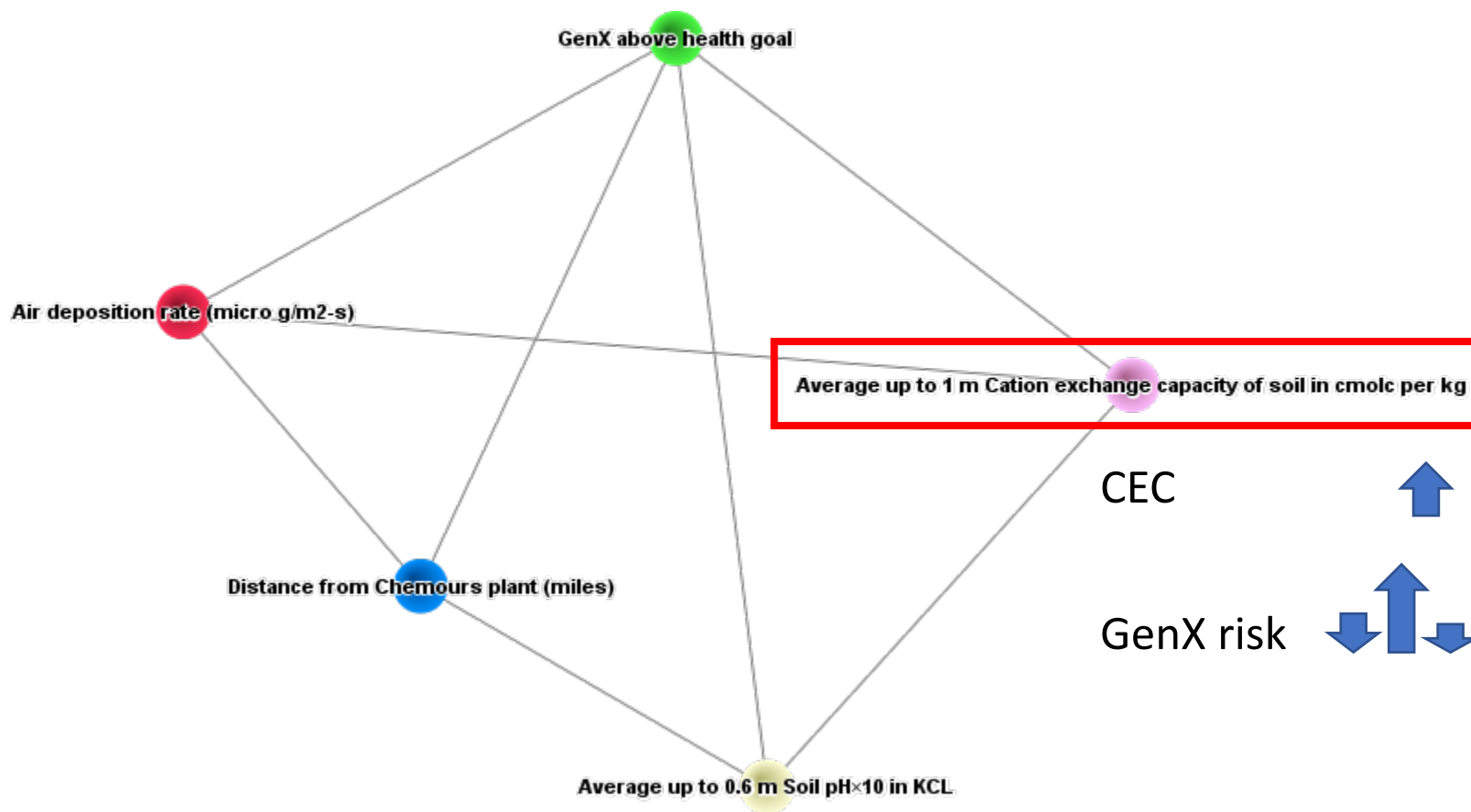
Knowing “Air Deposition Rate” provides the most information about GenX risk



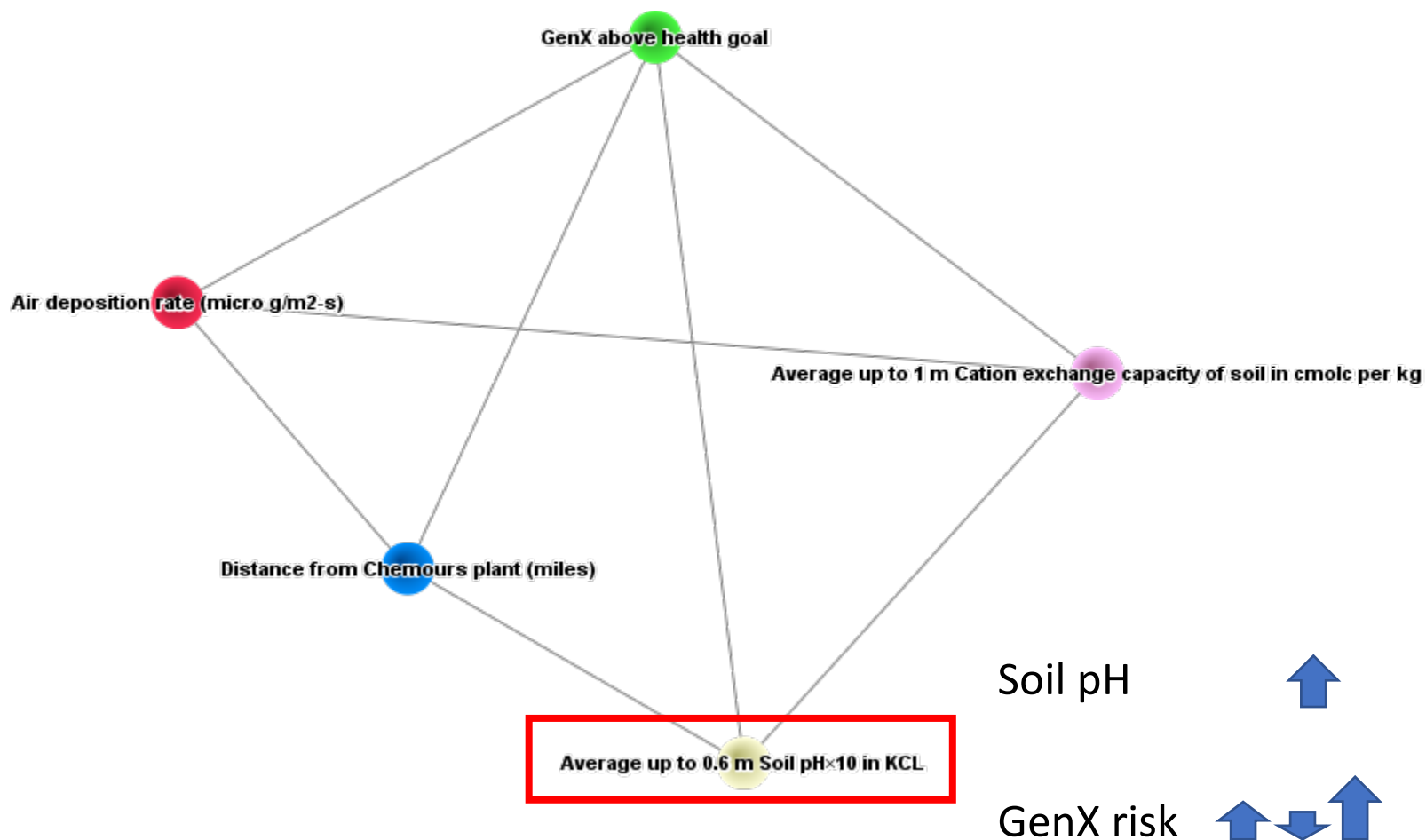
Distance from Chemours is second-most important predictor



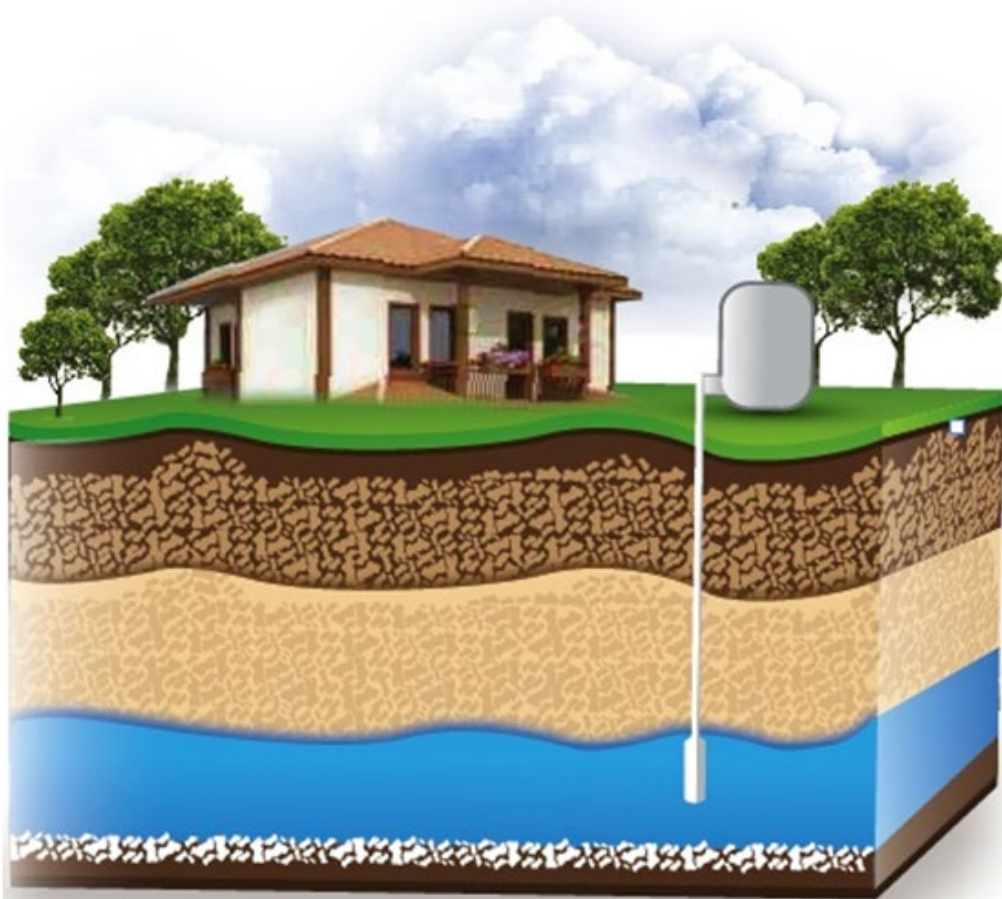
Cation Exchange Capacity is third-most important predictor



Soil pH in KCL is forth-most important predictor



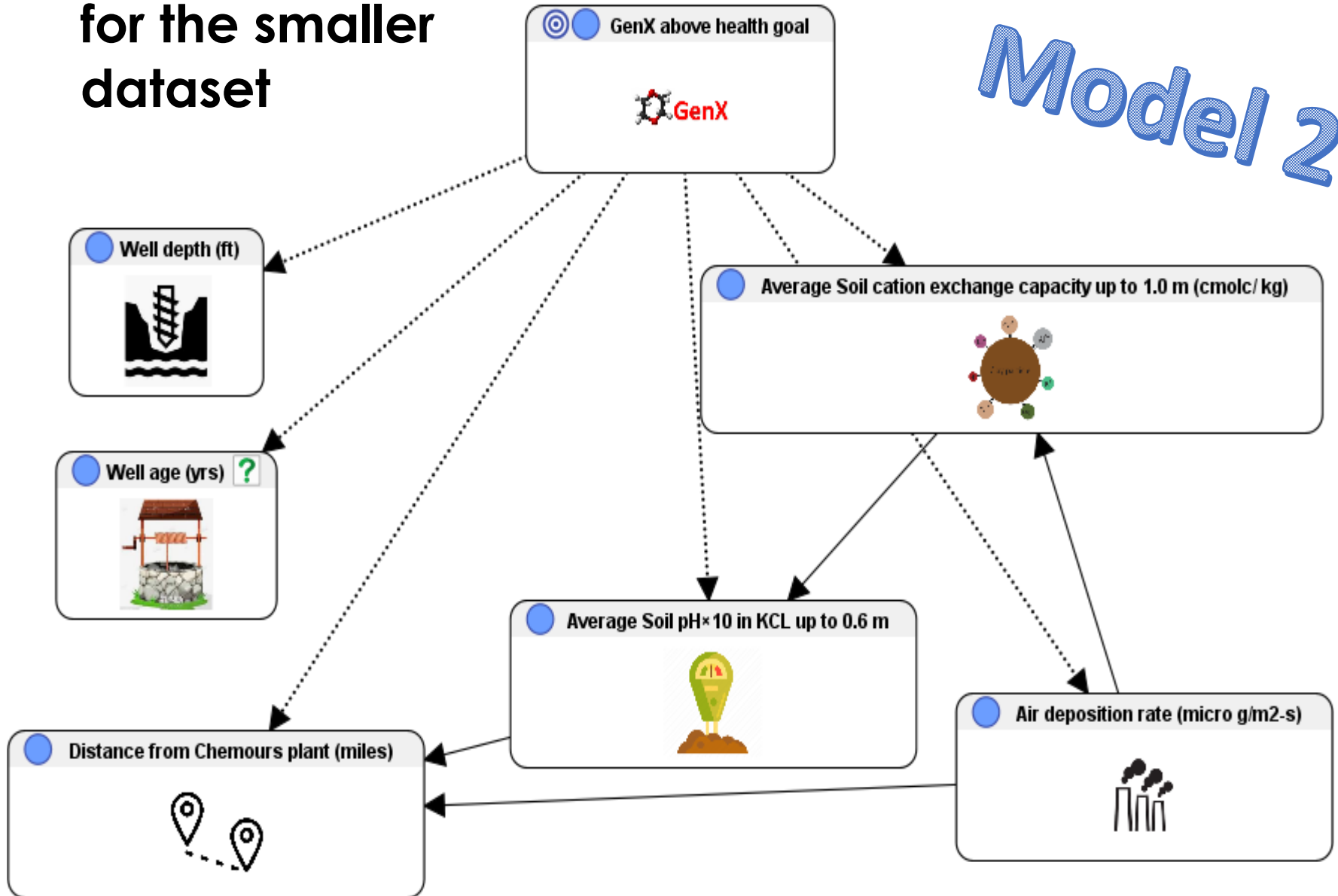
For 25% of houses we have additional information



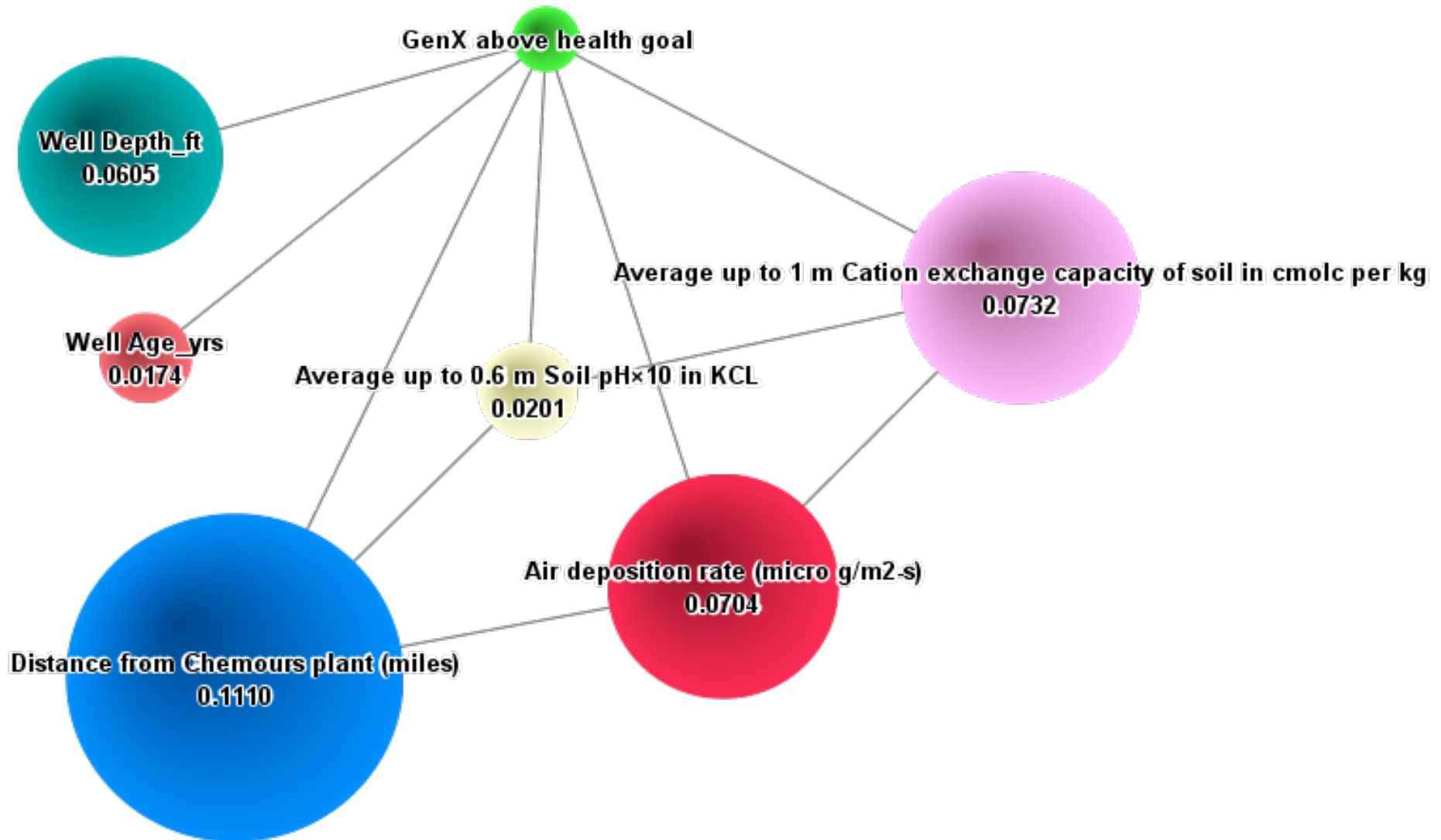
- ✓ Well depth
- ✓ Year of construction
- ✓ Method of construction

Main variables that influencing higher risk of GenX for the smaller dataset

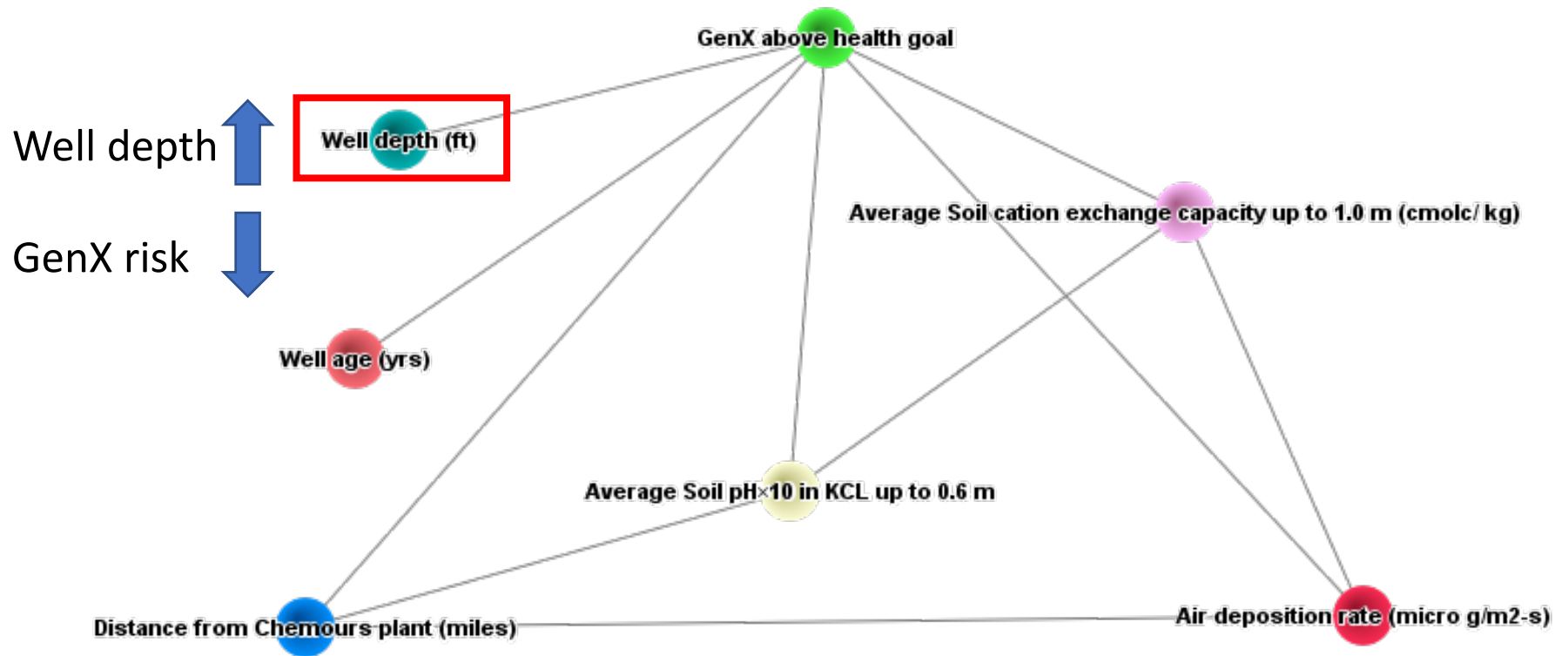
Model 2



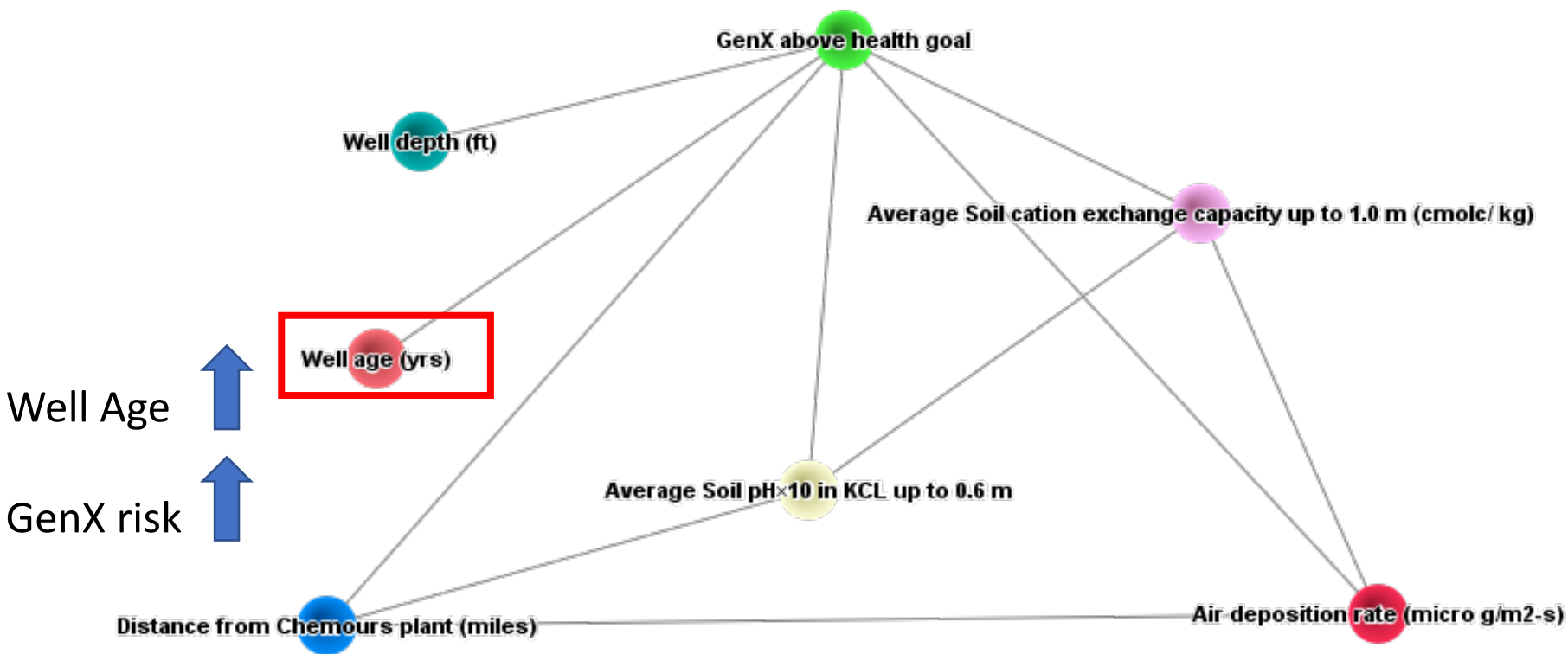
Mutual information with the GenX as target node



Well depth can reduce the risk of GenX contamination

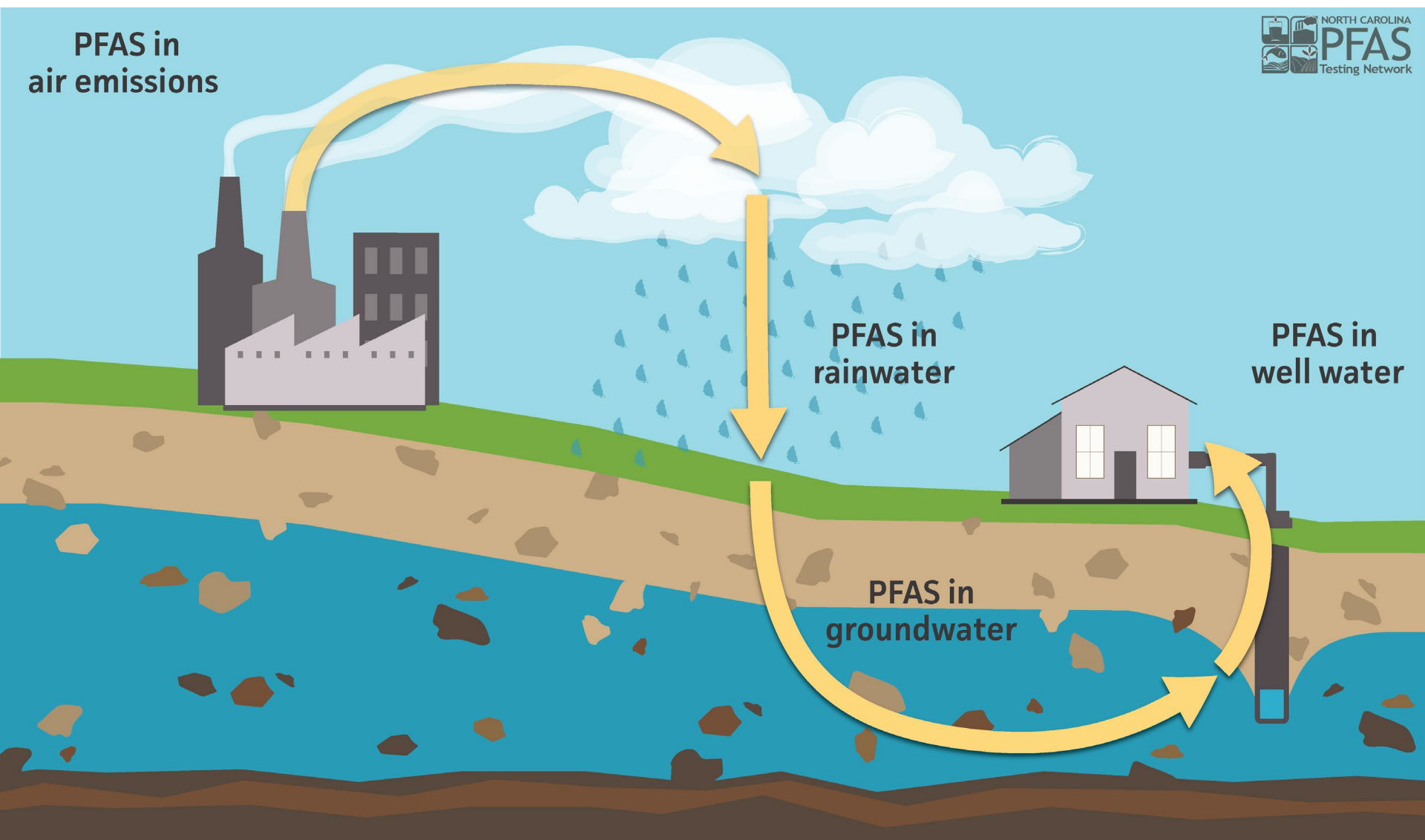


Increase in Well age may cause higher risk of GenX



Our ML model confirmed the source and how it reaches to the houses

PFAS in
air emissions

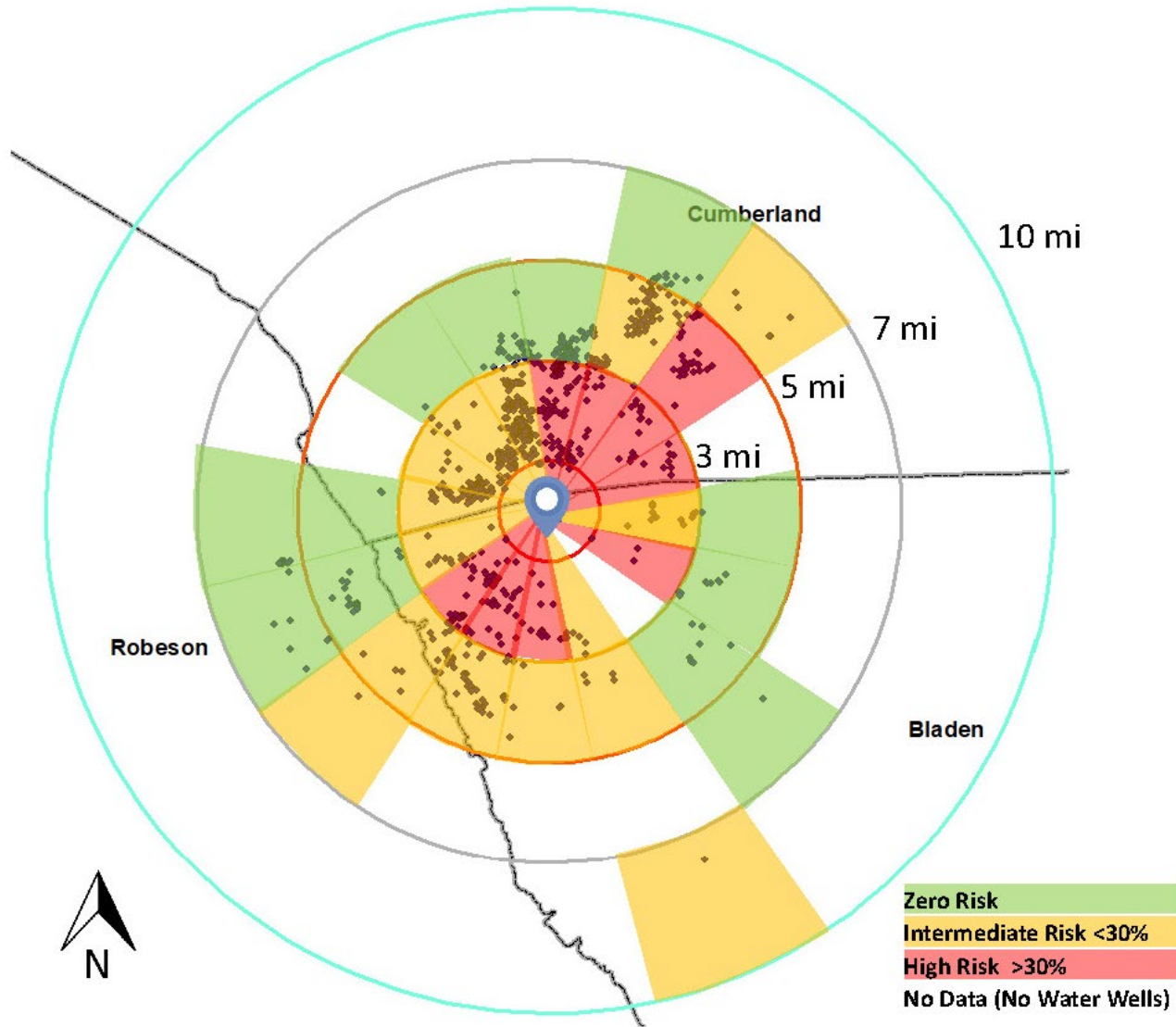


PFAS in
rainwater

PFAS in
well water

PFAS in
groundwater

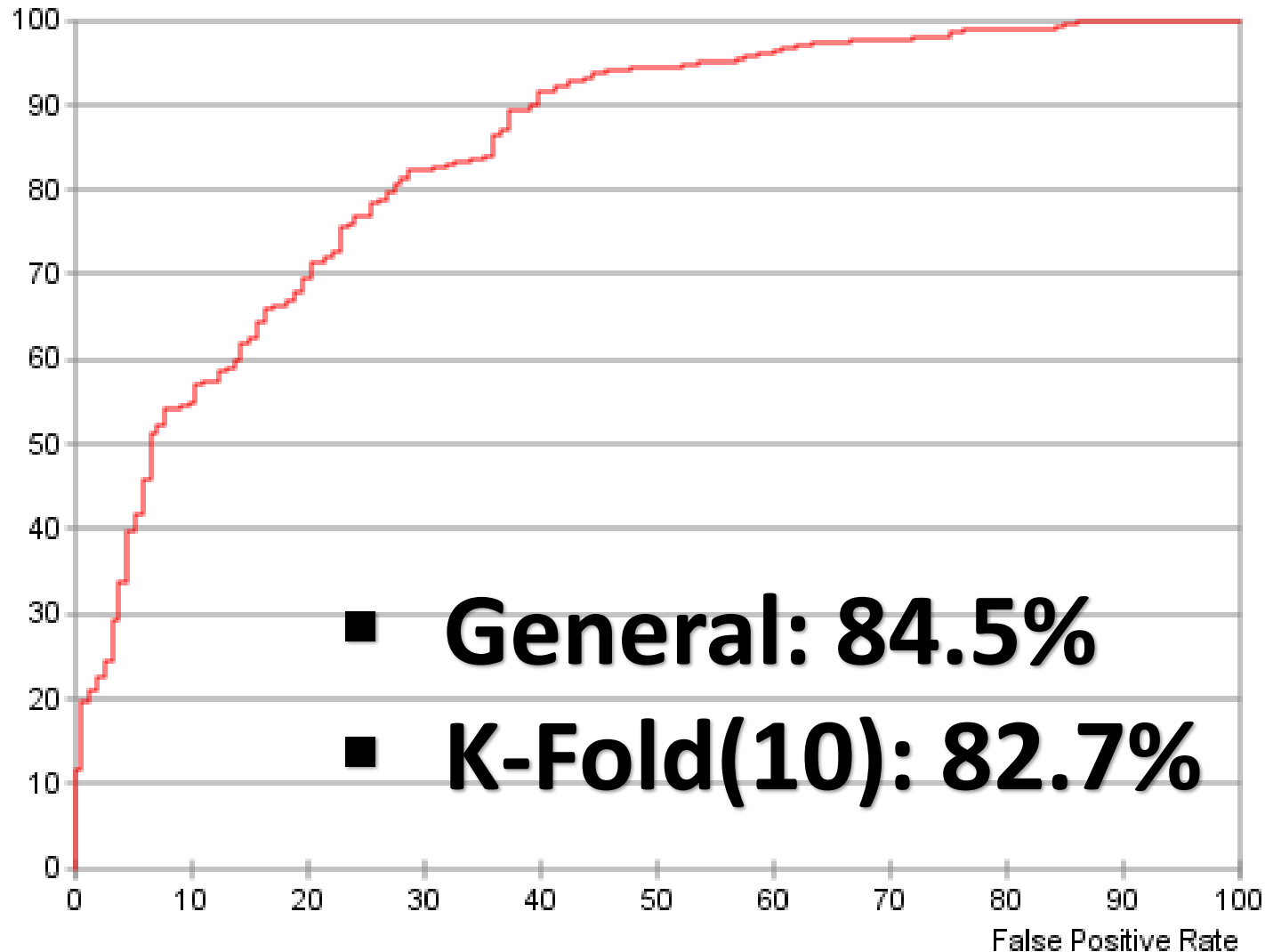
Predictive risk in the area of study



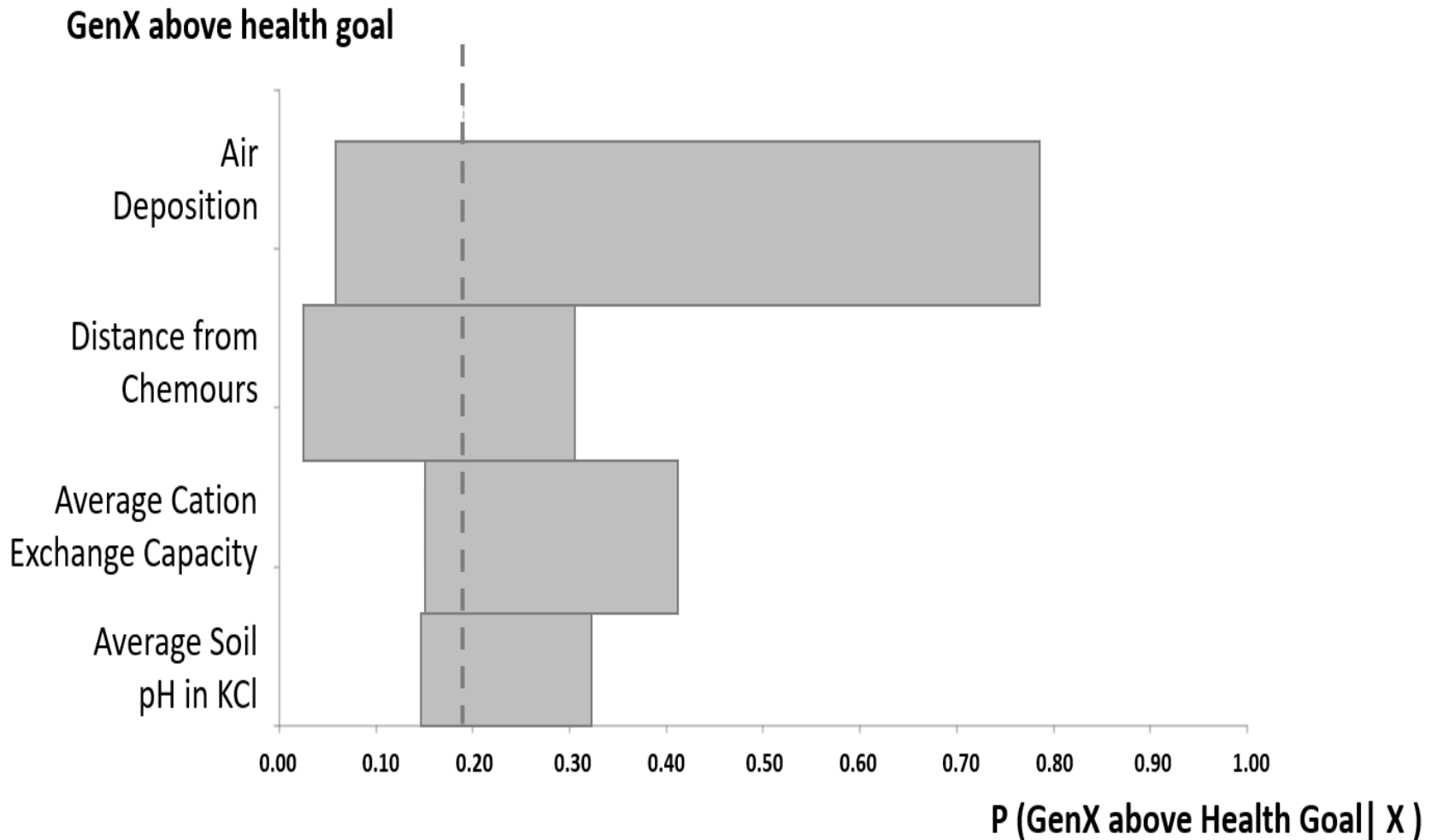
ROC curve for model 1 – whole dataset

True Positive Rate for GenX above health goal = ≤ 140

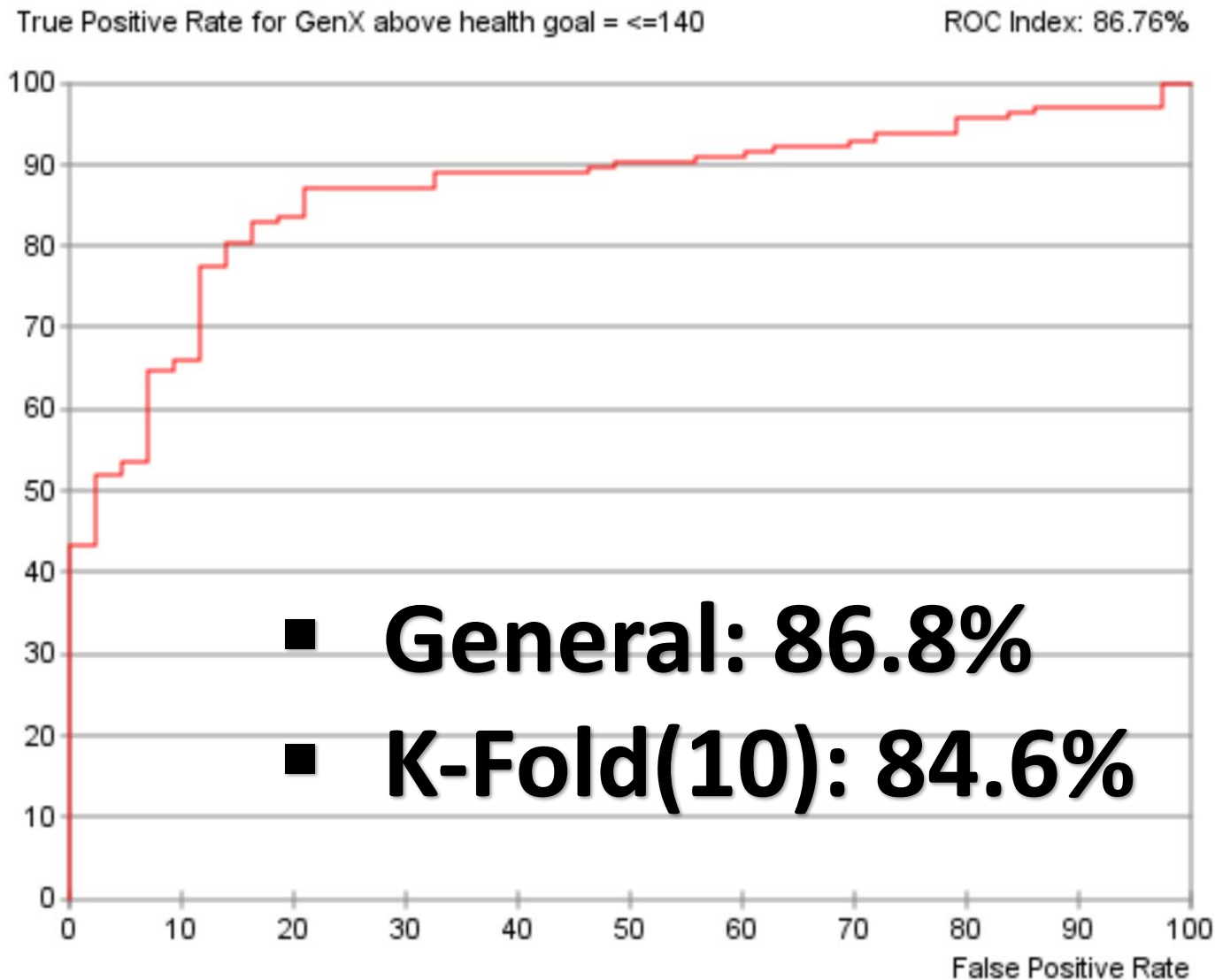
ROC Index: 84.47%



Tornado graph shows variables with most influence on risk – model 1

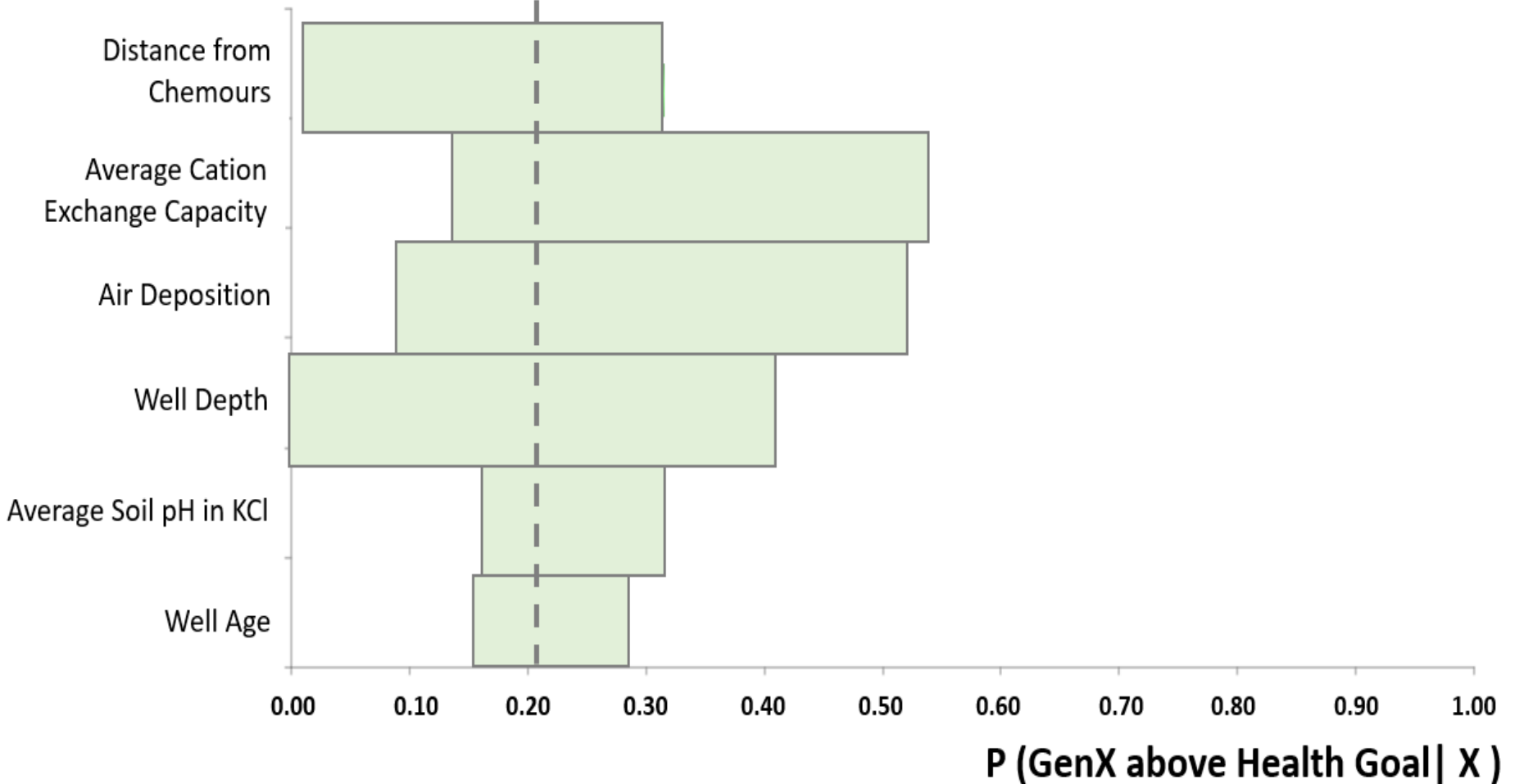


ROC curve for model 2 – 25% of data set



Tornado graph shows variables with most influence on risk – model 2

GenX above health goal



The web simulator for model 1



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UNC Predictive Model 1 (Whole Dataset) for Risk Prediction of GenX

Bayesia Simulator ModelwithoutWellDepthV1

Input 1) Average Soil Cation exchange capacity up to 1.0 m (cmolc/kg)

Mean

Observed

Input 2) Average Soil pHx10 in KCL up to 0.6 m

Mean

Observed

Input 3) Distance from Chemours plant (miles)

Mean

Observed

Input 6) Air deposition rate (micro g/m2-s)

Mean

Observed

Output 1) Probability of GenX above NC health goal of 140 ng/l

Probability of
GenX FC(F)(F)C(F)(F)C(F)(F)OC(F)(F)C(F)(F)O.N

<=140	80.95%
>140	19.05%

Output 2) Predicted Value of GenX (unit ng/l)

Predicted Value of
GenX FC(F)(F)C(F)(F)C(F)(F)OC(F)(F)C(F)(F)O.N Mean

The web simulator for model 2



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UNC Predictive Model for GenX Concentration in Private Wells

Bayesia Simulator

UNC PFAS Model_ With Well De



Average Soil cation exchange capacity up to 1.0 m (cmol/ kg)



Mean



12.35

Observed

Average Soil pHx10 in KCL up to 0.6 m



Mean



42.06

Observed

Well depth (ft)



Mean



91.108

Observed

Distance from Chemours plant (miles)



Mean



3.12

Observed

Well age (yrs)



Mean



18.1

Observed

Air deposition rate (micro g/m2-s)



Mean



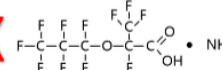
0.0000039

Observed

GenX above health goal

Probability of

GenX



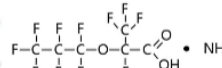
≤ 140 97.88%

> 140 2.12%

GenX above health goal

Predicted Value of

GenX



Mean 39.3



Conclusion

Conclusion

- ❑ Bayesian network identified **factors** associated with the GenX Concentration
 - Proximity to Chemours Plant
 - Air Deposition
 - Soil pH
 - Cation Exchange Capacity

- ❑ BN model can classify the private wells with above **average sensitivity (84%)** and reasonable false-positive rate.

- ❑ Adding more well permit **data and more measurement** to the model can help increase ROC index.

Main takeaways

- Although PFAS issue is complicated, we have **tools** and datasets to build ML models which can help us understand PFAS better.
- Bayesian network modeling has the capability to show us the **influencing variables** in PFAS contaminations
- The **accuracy** of the ML model is reasonably good, however **more data** (such as well-depth, age, and measurements) can help it increases the accuracy.

Acknowledgement

❑ Funding Source

- North **Carolina** Policy Collaboratory



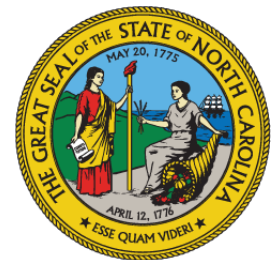
❑ Collaborators

- Dr. Detlef Knappe, NCSU
- Dr. David Genereux, NCSU
- Riley Mulhern, UNC
- Dr. Frank Stillo, UNC
- Ariel Lam, UNC
- Alex Shoaf, UNC



❑ Organizations

- Local Health Department at Cumberland, Robeson and Bladen County
- NC DEQ



thank
you!

Questions?



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References

S. Conrady;, Bayesian Networks & bayesialab. 2015.

SHEENA SCRUGGS, 2019, PFAS — a problem in North Carolina drinking water, <https://factor.niehs.nih.gov/2019/3/feature/2-feature-pfas/index.htm>

Urban Forest Data for North Carolina

<https://www.nrs.fs.fed.us/data/urban/state/?state=NC>

Soil survey ,

http://www.ncmhtd.com/arcgis/rest/services/NRCS/NRCS_SoilData/MapServer/4

National Land Cover Database (NLCD)

https://www.usgs.gov/centers/eros/science/national-land-cover-database?qt-science_center_objects=0#qt-science_center_objects

Fire Detection GIS Data

<https://fsapps.nwcg.gov/gisdata.php>