

# Machine Learning for Structural Analysis in FCC Crystal

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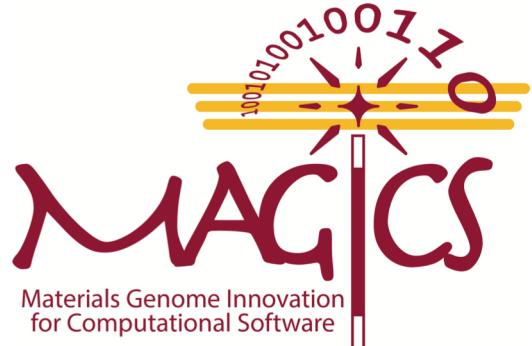
Pankaj Rajak

*Argonne National Laboratory*

Ken-ichi Nomura, Nitish Baradwaj

*Collaboratory for Advanced Computing & Simulations*

*University of Southern California*



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**ENERGY**

Office of  
Science  
Basic Energy Sciences

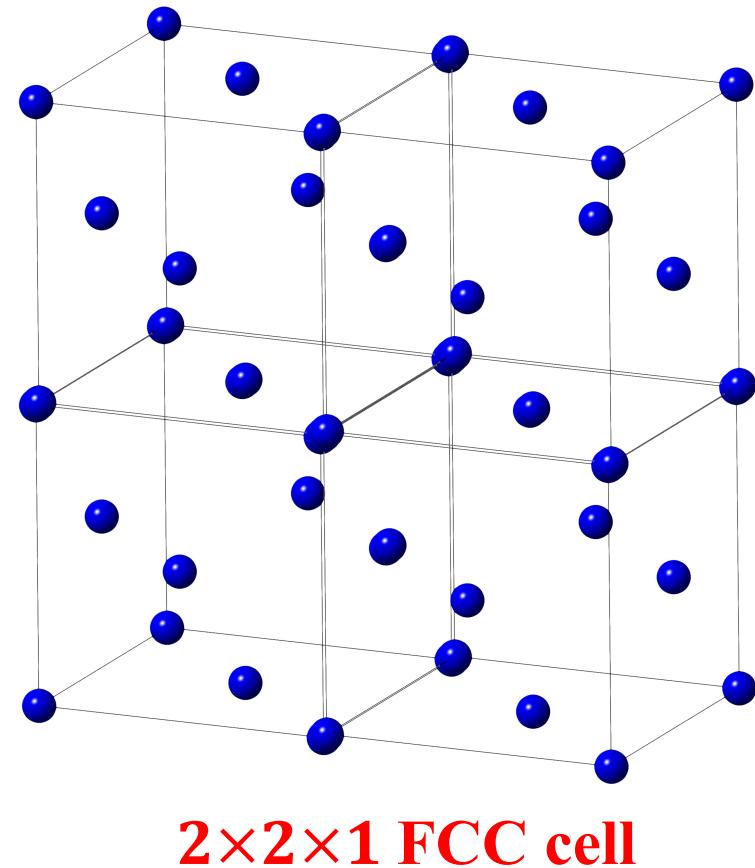
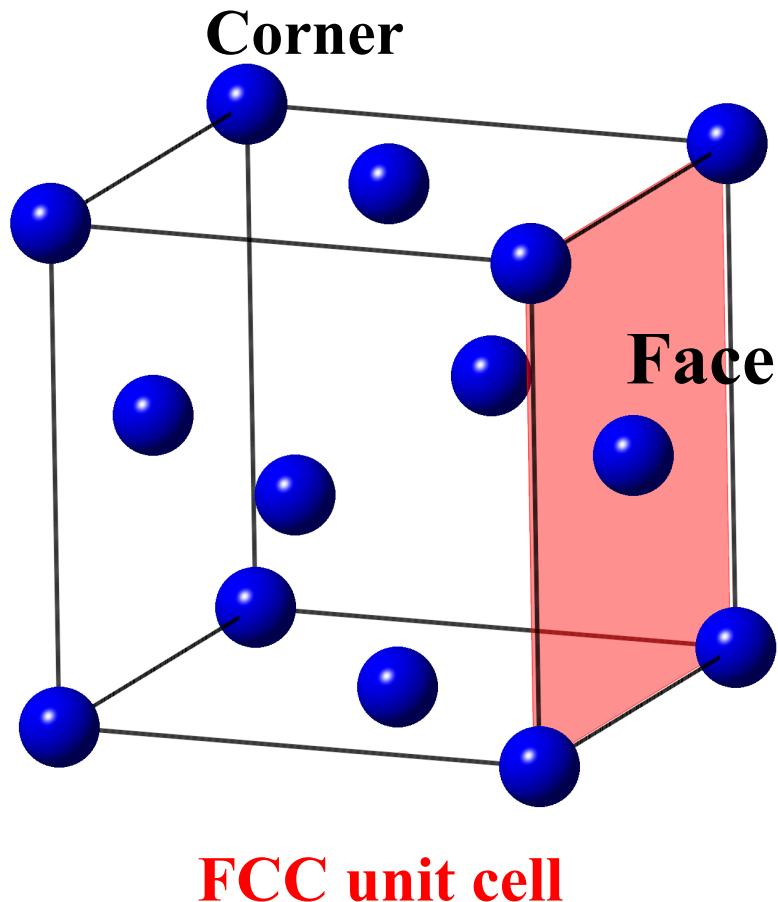
# Outline

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- **FCC crystal and Nanoindentation MD simulation**
- **Machine Learning (ML) model for structural analysis**
- **Hands-on session:**
  - **Structural analysis using ML**
  - **Visualization of predicted label in OVITO**

# Face Centered Cubic (FCC) Crystal

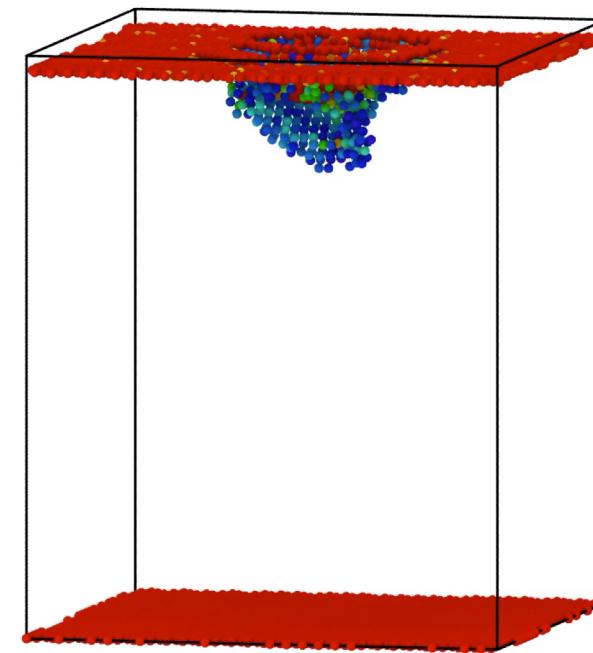
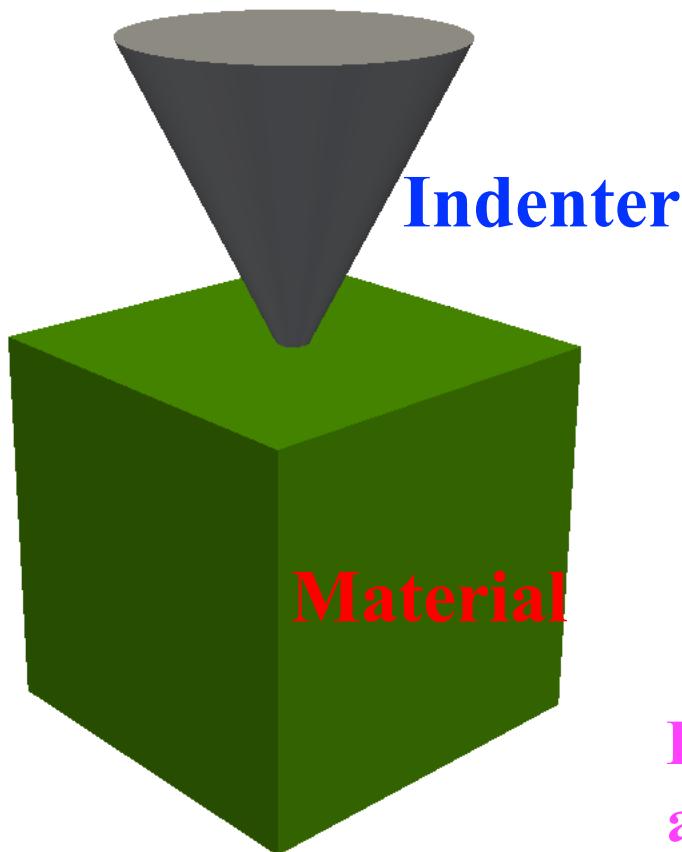
In FCC crystal, each unit cell contain atoms on all the 8 corners and the 6 faces



# Nanoindentation Simulation of FCC Crystal using Molecular Dynamics (MD)

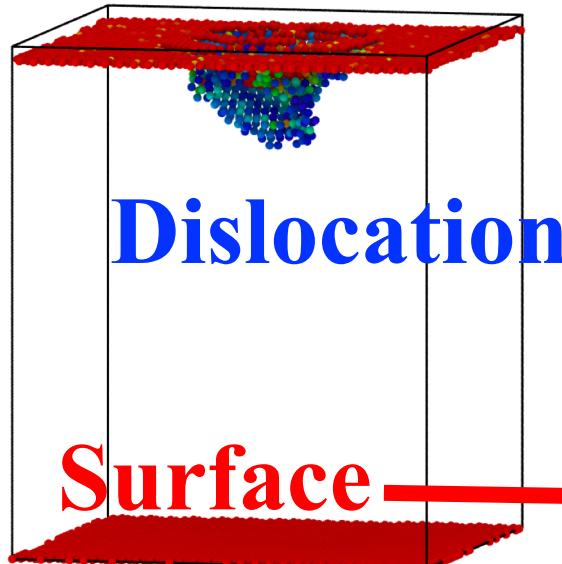
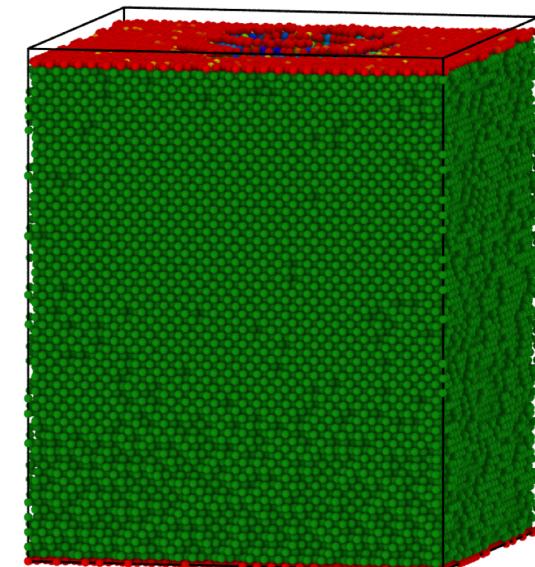
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**Nano-indentation Simulation:** Used to study mechanical properties of material ( hardness, elastic constant)



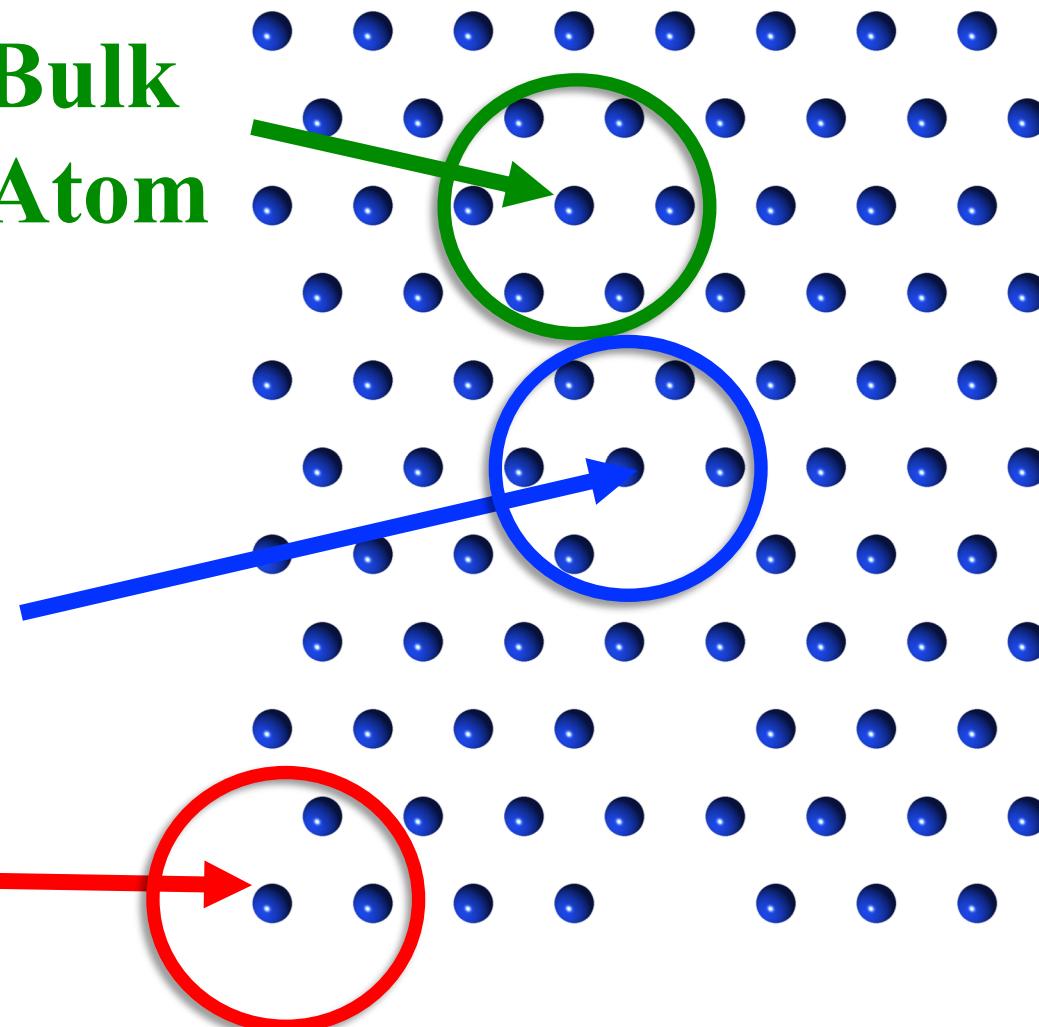
During MD simulation, dislocations are generated inside the material

# Various Atomic Configurations in FCC crystal

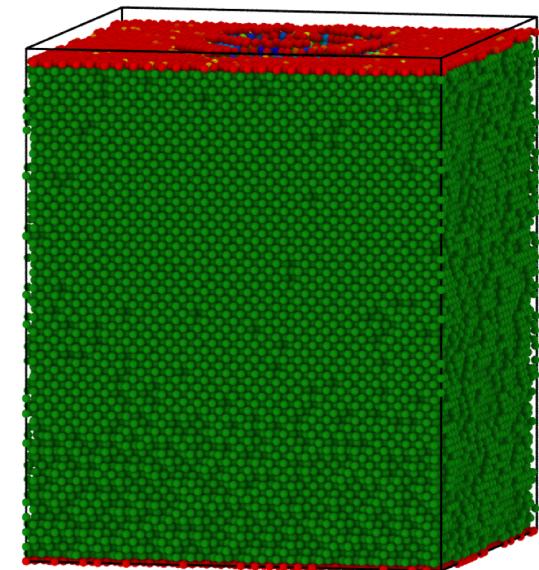


Bulk  
Atom

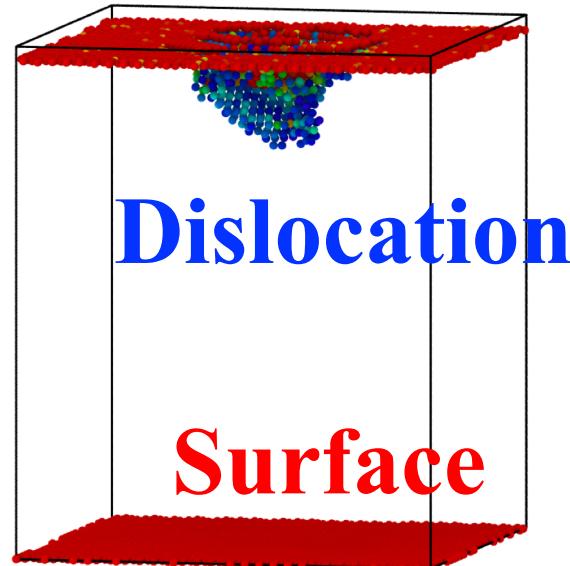
➤ Dislocation & surface atoms  
have missing neighbors



# Various Atomic Configurations in FCC Crystal



Bulk  
Atom



➤ Goal: Build a Machine Learning (ML) model that can identify all these structures

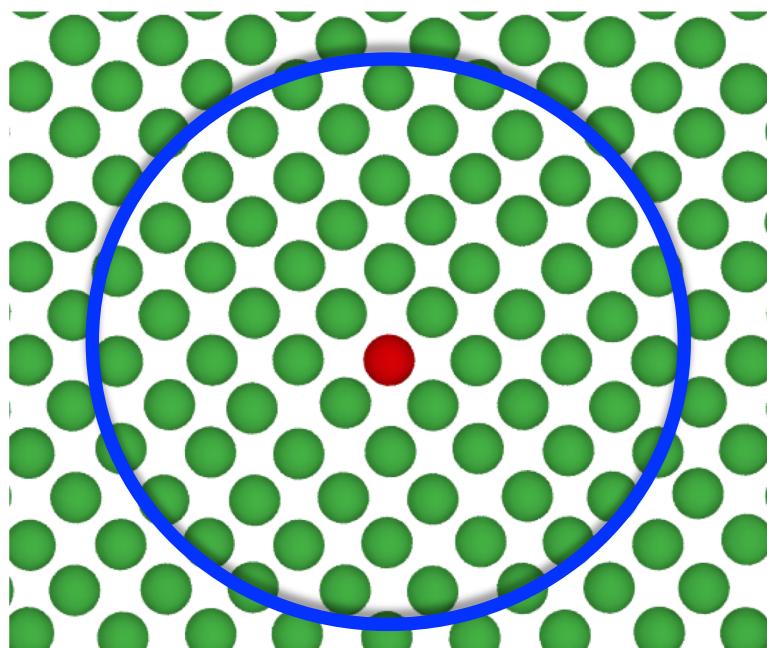
Labels for atomic configurations generated during nanoindentation simulation

	Label (Y)
Bulk Atom	0
Surface	1
Dislocation	2

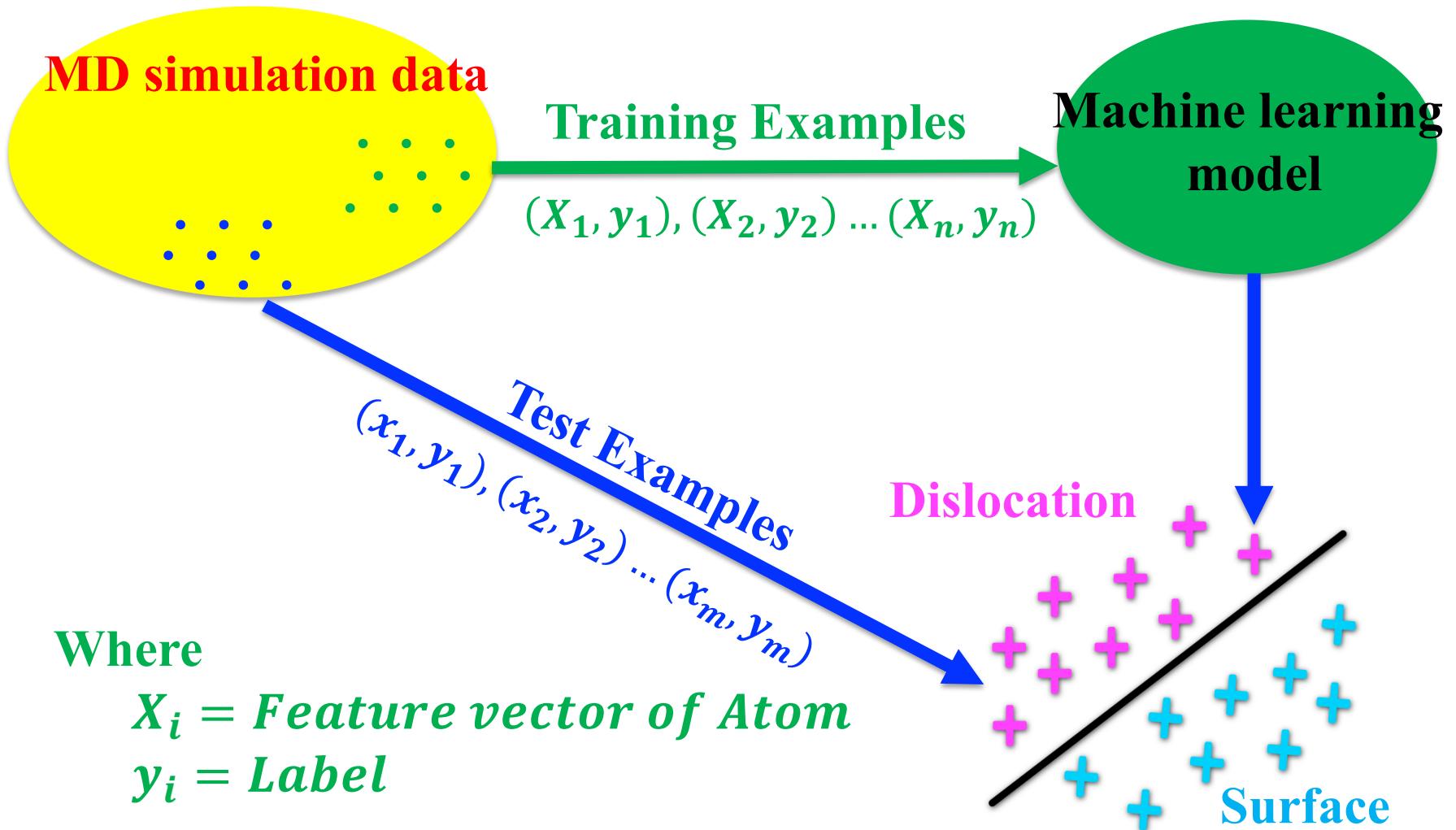
# Feature Vector

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A mathematic representation for each atom which uniquely describes the local environment of an atom



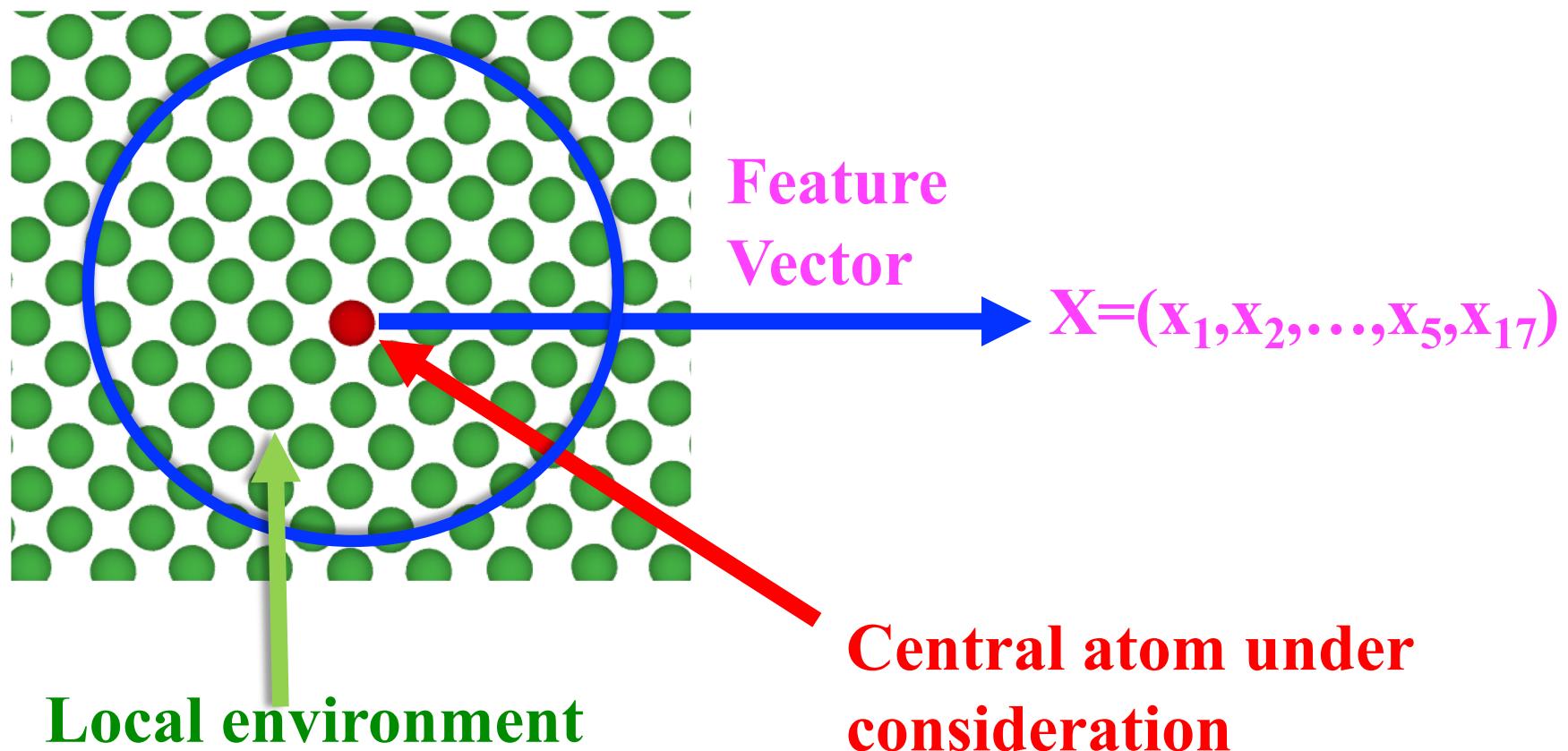
# Objective: Build a Machine Learning (ML) Model for Structural Analysis



# Step 1: Create Feature Vector for Each Atom

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Convert atomic coordinates into a feature vector that captures the local geometry around each atom

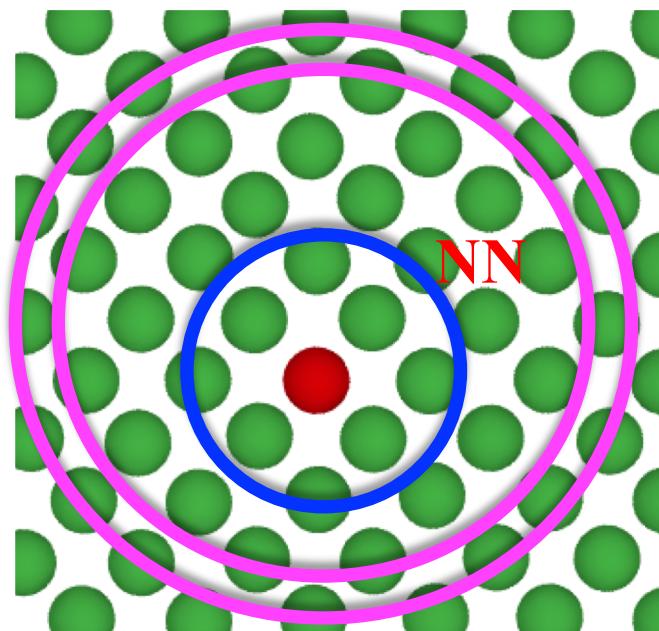


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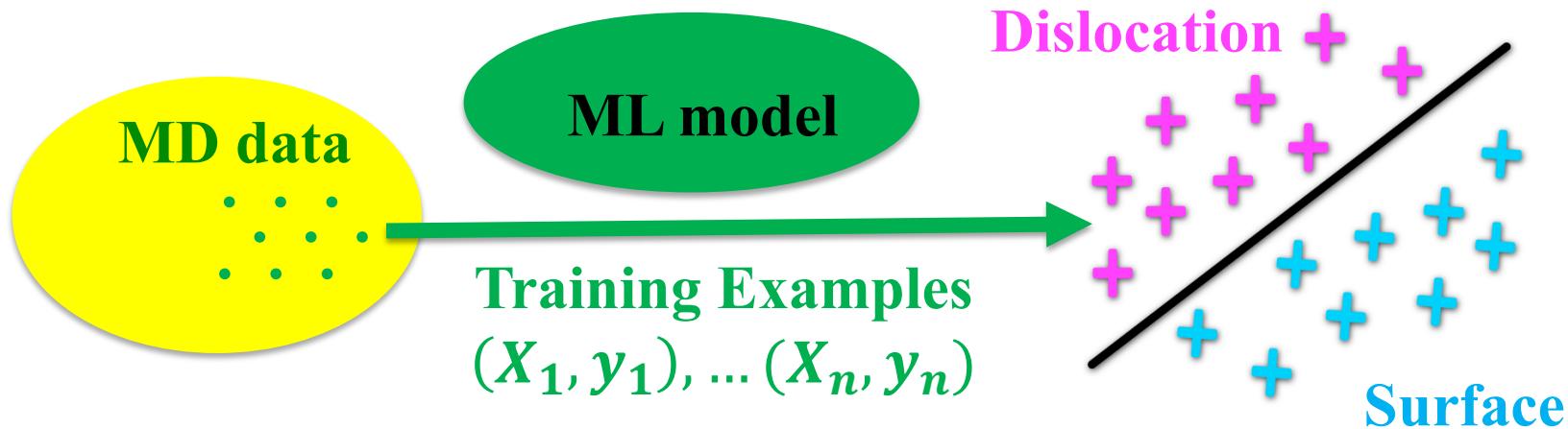
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**Feature Vector:** Each atoms local environment will be represented using 17 different properties

1. No. of nearest neighbor (NN)
2. Average distance of NN
3. Minimum distance of NN
4. Maximum distance of NN
5. Average distance between NN
6. Minimum distance between NN
7. Maximum distance between NN
8. NN's average numbers of neighbor
9. NN's neighbor's average distance
10. NN's neighbor's minimum distance
11. NN's neighbor's maximum distance
- 12-14 Number of neighbor's between 3-4, 4-5, 5-6A
- 15-17 Average distance of neighbor's between 3-4, 4-5, 5-6A



# Step 2: Build a Linear Classifier using Machine Learning



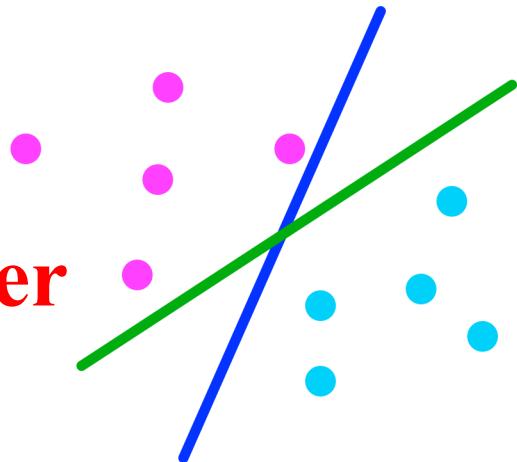
## ML model (a linear classifier)

$$y_{predicted} = w_1 x_1 + \dots + w_{17} x_{17} - b \begin{cases} > 0 \text{ (Dislocation)} \\ \leq 0 \text{ (Surface)} \end{cases}$$

$w_i$  : tunable parameters that we will learn using training examples

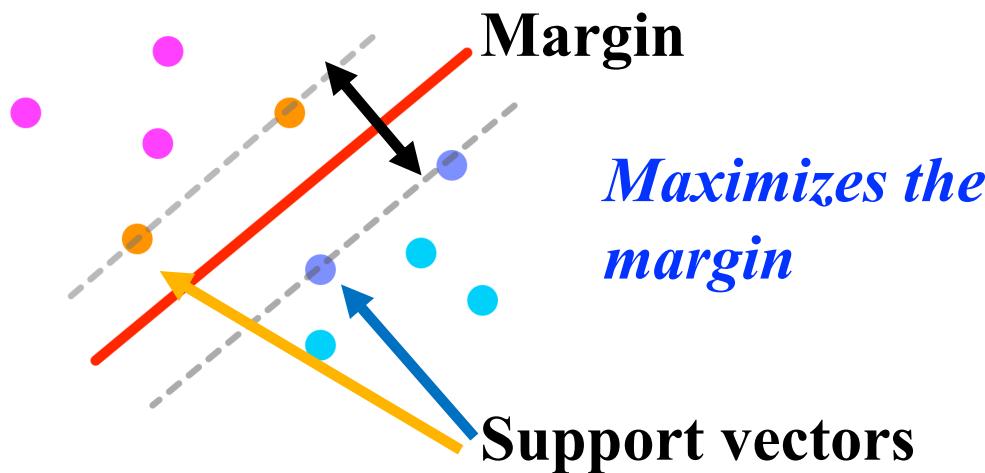
# Linear Support Vector Machine (SVM)

Linear classifier



*Many possible  
choices for decision  
boundary*

Linear SVM



*Maximizes the  
margin*

# Step 3: Evaluate Accuracy of Model

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- Compare the accuracy of the model on training and test data.

$$Error = \frac{1}{N} (y_{predicted} - y_{true})^2$$

$$Accuracy = 1 - Error$$

- Visualize the predicted labels by the ML model on test data using OVITO

# Machine Learning Module for Structural Analysis

---

# Files under Ml\_module

```
$ ls Ml_module
```

Makefile	Ni_train.xyz	createfeature.h
Ni_ML.ipynb	SVM_model.py	<b>feature/</b>
Ni_ML.py	atom_property.c	readinput.py
Ni_test1.xyz	atom_property.h	
Ni_test2.xyz	createfeature.c	

1) **atom\_property.c** and **createfeature.c**: C code to read atomic configuration and create feature vector for atoms

2) Ni\_ML.py: python code to create and train ML classifier

3) SVM\_model.py: python script to build the ML model

4) Ni\_train.xyz, Ni\_test1.xyz and Ni\_test2.xyz : Atomic coordinate for training and test data

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createfeature.h  
**feature/**  
readinput.py

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# Compute Feature Vector from Atomic Coordinates for Training and Validation

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- Using training data (**Ni\_train.xyz**), compute feature vector for each atom
- First, build executable (**c\_feature**) from the C codes (**atom\_property.c** and **createfeature.c**).

**make**

- Using **c\_feature**, compute feature vector from training data

```
./c_feature Ni_train.xyz feature/train.txt
```

**Ni\_train.xyz** : Input atomic coordinate in XYZ format

**feature/train.txt** : Output that contains feature vector for each atoms

# File Format that Contains Atomic Coordinates and Features

- Each line of the created feature vector contains atom type, x y z coordinates, label and 17 dimension feature vector for each atom

```
$ ls feature/train.txt
```

114376																	
17																	
Ni	101.481003	130.103226	101.481003														
Ni	1.756660	4.199579	1.764920	2	8.000000	2.471503	2.444679	2.500690	3.391102								
Ni	1.746970	5.941798	0.018508	0	11.000000	2.480045	2.458448	2.507447	3.551417								
Ni	5.244970	4.214318	1.754720	2	8.000000	2.467677	2.449782	2.491687	3.391526								
Ni	3.489960	5.923518	1.752180	0	11.000000	2.474269	2.449782	2.492165	3.541959								
Ni	3.495490	4.193375	0.006788	2	7.000000	2.476350	2.457604	2.497069	3.304576								
Ni	8.770810	4.207564	1.757480	2	8.000000	2.473605	2.448607	2.490758	3.395039								
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Ni	6.998650	4.182088	0.009597	2	8.000000	2.480736	2.468257	2.497350	3.387989								
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Ni	10.503100	5.947286	1.755110	0	12.000000	2.474824	2.455090	2.500750	3.544996								
Ni	12.237800	5.942111	0.005140	0	12.000000	2.474072	2.455739	2.493437	3.543886								
Ni	15.734100	4.192092	1.764940	2	8.000000	2.480301	2.461069	2.498584	3.390528								
Ni	13.984100	5.945300	1.749540	0	12.000000	2.472308	2.454947	2.483916	3.541379								
Ni	13.987700	4.219180	0.002308	2	8.000000	2.467989	2.455739	2.481780	3.391241								
Ni	15.745700	5.948635	0.000045	0	12.000000	2.474543	2.451119	2.490067	3.544575								
Ni	17.493999	5.937821	1.738920	0	12.000000	2.474458	2.447819	2.490233	3.544428								

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x,y,z coordinate

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Feature vector

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type

x,y,z coordinate

label

Feature vector

# Python Class to Build SVM Linear Classifier

**build\_classifier** provides three member functions; **train**, **prediction** and **accuracy**.

```
class build_classifier:
```

```
    def __init__(self, trainX, trainY):
```

- load training data (trainX)
- Load true labels (trainY)
- Normalize training data (trainX)

```
    def train(self):
```

- Train the ML model using normalize training data

```
    def predict(self):
```

- predict labels of test data using trained model

```
    def accuracy(self):
```

- computes prediction accuracy of the model

# Python Class to Build SVM Linear Classifier

Training data (*trainX*) and true label (*trainY*) is necessary to instantiate the class.

```
class build_classifier:  
    def __init__(self, trainX, trainY):  
        ▪ load training data (trainX)  
        ▪ Load true labels (trainY)  
        ▪ Normalize training data (trainX)  
  
    def train(self):  
        ▪ Train the ML model using normalize training data  
  
    def predict(self):  
        ▪ predict labels of test data using trained model  
  
    def accuracy(self):  
        ▪ computes prediction accuracy of the model
```

# Python Class to Build SVM Linear Classifier

Trains the ML model given training data and true labels.

```
class build_classifier:
```

```
    def __init__(self, trainX, trainY):
```

- load training data (trainX)
- Load true labels (trainY)
- Normalize training data (trainX)

```
    def train(self):
```

- Train the ML model using normalize training data

```
    def predict(self):
```

- predict labels of test data using trained model

```
    def accuracy(self):
```

- computes prediction accuracy of the model

# Python Class to Build SVM Linear Classifier

Using the trained model, predict label for test data.

```
class build_classifier:
```

```
    def __init__(self, trainX, trainY):
```

- load training data (trainX)
- Load true labels (trainY)
- Normalize training data (trainX)

```
    def train(self):
```

- Train the ML model using normalize training data

```
    def predict(self):
```

- predict labels of test data using trained model

```
    def accuracy(self):
```

- computes prediction accuracy of the model

# Python Class to Build SVM Linear Classifier

Calculate error and accuracy of the developed model.

```
class build_classifier:
```

```
    def __init__(self, trainX, trainY):
```

- load training data (trainX)
- Load true labels (trainY)
- Normalize training data (trainX)

```
    def train(self):
```

- Train the ML model using normalize training data

```
    def predict(self):
```

- predict labels of test data using trained model

```
    def accuracy(self):
```

- computes prediction accuracy of the model

# 1. Train the Linear SVM Classifier

---

- Instantiate `build_classifier` class

```
Ni_model = build_classifier (train_x, train_y)
```

- Trains a linear classifier using `svm.LinearSVC()`, and store the model into a variable `self.model`.

```
Ni_model.train()
```

- `train()` also computes error and accuracy of the developed model.

Number of training examples: 18686

Training error = 3.16%

Training accuracy = 96.83%

## 2. Compute Accuracy of the Model using Test Data

- Convert atomic coordinates of test data (`Ni_test1.xyz`) into feature vector.

```
./c_feature Ni_test1.xyz feature/test_1.txt
```

- Predict labels of the test data using the trained model by `predict()` and `accuracy()` function.

```
label_pred = Ni_model.predict(testX)
accuracy = Ni_model.accuracy(textY,label_pred)
```

Output:

Test error = 1.03%

Test accuracy = 98.96%

### 3. Visualize the Predicted Label in OVITO

---

- `writexyz()` function creates an output file (`output.xyz`) that contains atomic coordinates and the true and predicted labels

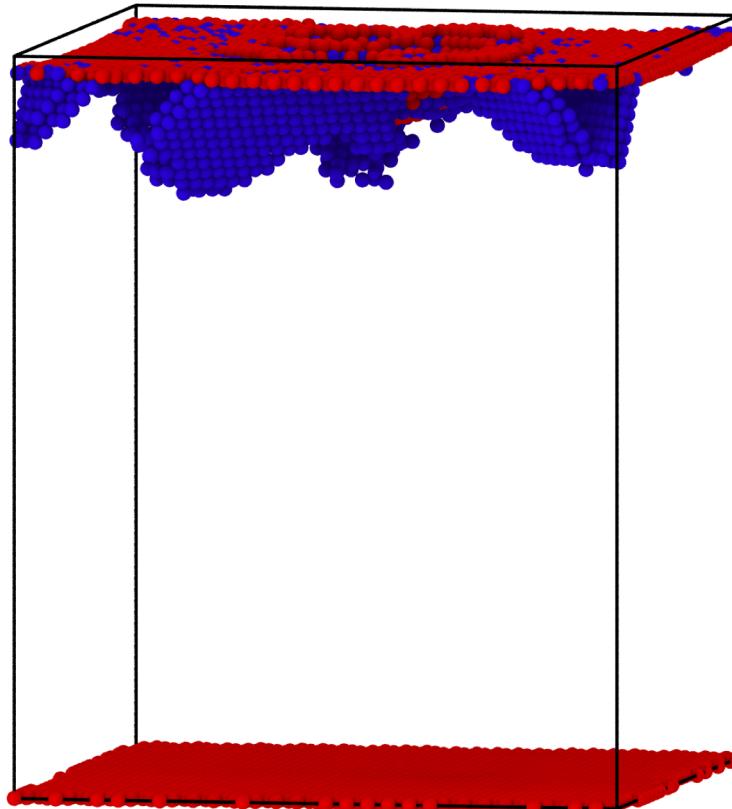
```
writexyz(Natoms, position, labelpred, labeltrue)
```

- Visualize the predicted label (`output.xyz`) in OVITO

### 3. Visualize the Predicted Label in OVITO

---

True Label



ML Predicted Label

