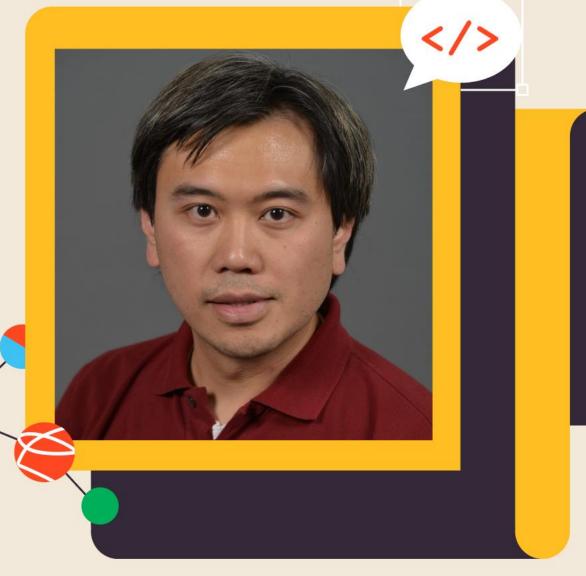


0 N L I N E NOVEMBER 23-25

Making
Neural Network Portable
with ONNX

# Ron Dagdag

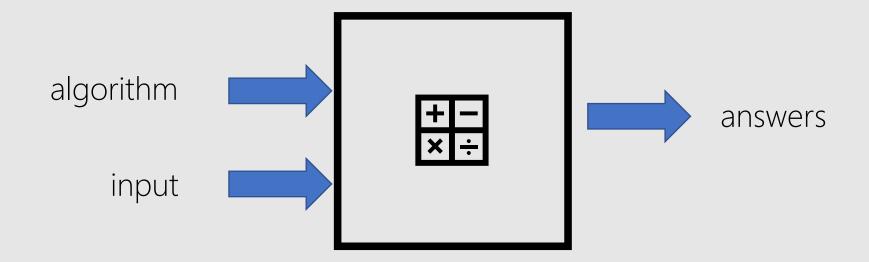


ONNX
Not ONIX
Not ONYX





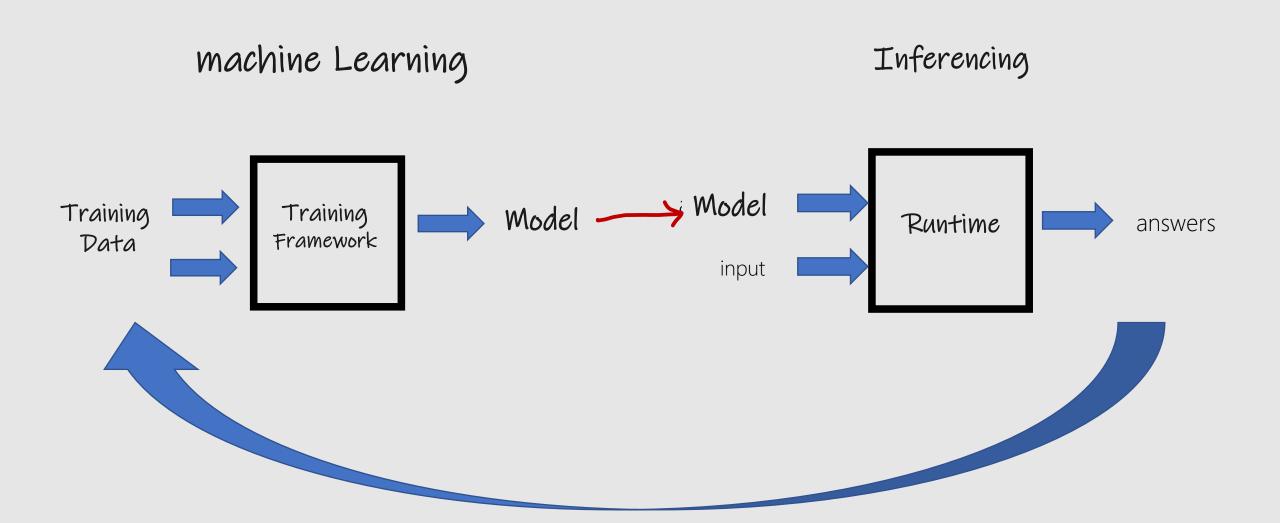
# programming



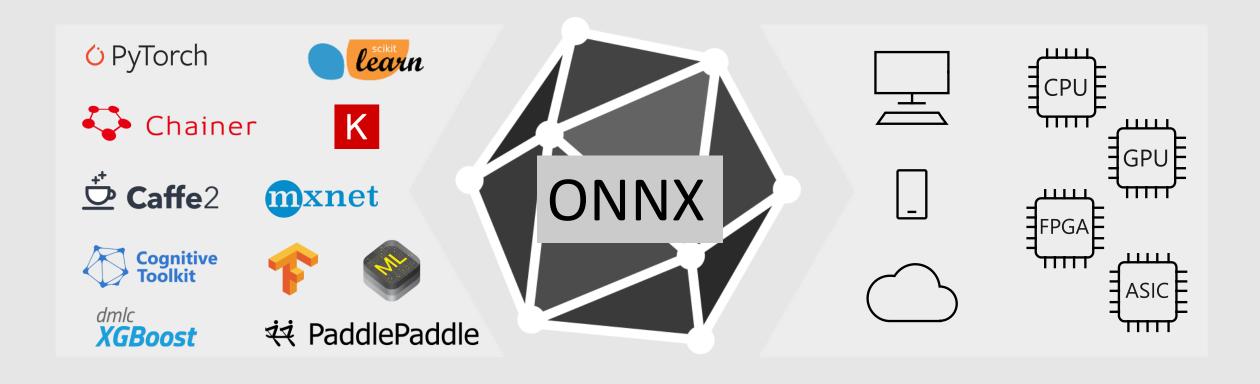
# machine learning



### ML Primer



### Open and Interoperable Al





Open Neural Network Exchange

### Open format for ML models

github.com/onnx onnx.ai/

# ONNX Partners

























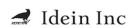




















































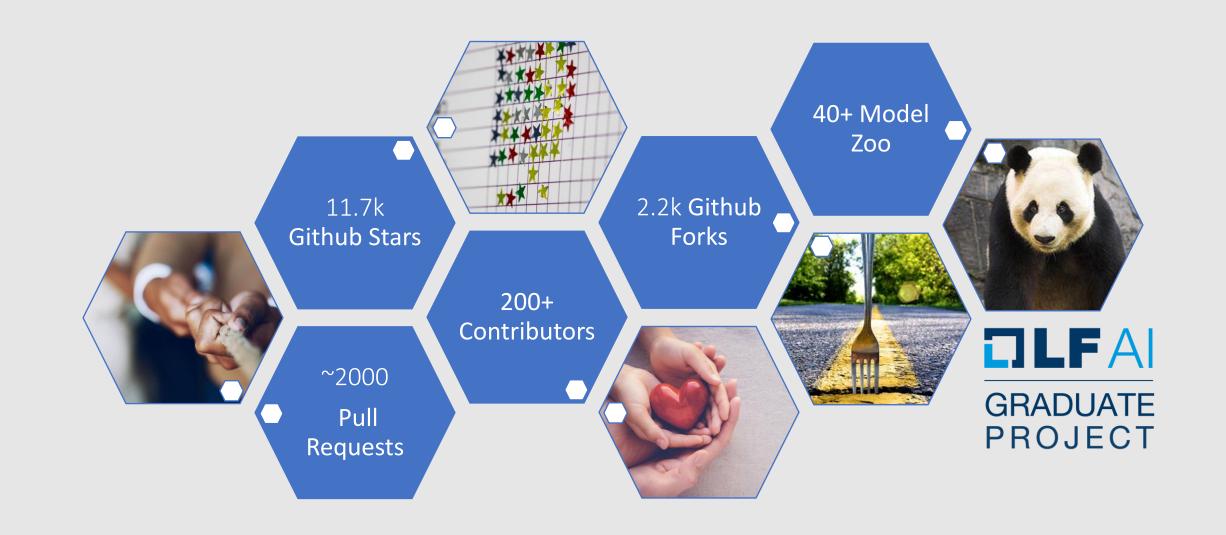












# Agenda

✓ What is ONNX, When to use ONNX

☐ How to create ONNX models

☐ How to deploy ONNX models

### When to use ONNX?

- Trained in Python deploy into a C#/Java/Javascript app
- High Inferencing latency for production use
- Model to run resource on IoT/edge devices
- Model to run on different OS or Hardware
- Combine running models created from different frameworks
- Training takes too long (transformer models)

### Create

### **Frameworks**







Native support





















Converters

Native support



**ONNX Model** 

### Deploy

**Cloud Services** 

Azure Machine Learning services

Ubuntu VM

Windows VM

**Windows Devices** 

**IoT/Edge Devices** 

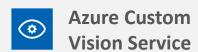
Converters

Native

support

**Other Devices** (iOS, Android, etc)

### **Services**











### 4 ways to get an ONNX model



ONNX Model Zoo



**Azure Custom Vision Service** 



Convert existing models



Train models in Azure Machine Learning

**Automated Machine Learning** 

# ONNX Model Zoo: github.com/onnx/models

#### **Image Classification**

This collection of models take images as input, then classifies the major objects in the images into a set of predefined classes.

Top-5

Model Class	Reference	Description						
MobileNet	Sandler et al.	Efficient CNN model for mobile and embedded vision applications.  Top-5 error from paper - ~10%						
ResNet	He et al., He et al.		Very deep CNN model (up to 152 layers), won the ImageNet Challenge in 2015.					
SqueezeNet	landola et al.	A ligh fewer Top-5	Model	Download	Checksum	Download (with sample test data		
VGG	Simonyan et al.	Deep Challe	ResNet- 18	44.6 MB	MD5	42.9 MB		

Model	Download	Checksum	Download (with sample test data)	ONNX version	Opset version	Top-1 accuracy (%)	Top-5 accuracy (%)
ResNet- 18	44.6 MB	MD5	42.9 MB	1.2.1	7	69.70	89.49
ResNet- 34	83.2 MB	MD5	78.6 MB	1.2.1	7	73.36	91.43
ResNet- 50	97.7 MB	MD5	92.0 MB	1.2.1	7	75.81	92.82
ResNet- 101	170.4 MB	MD5	159.4 MB	1.2.1	7	77.42	93.61
ResNet- 152	230.3 MB	MD5	216.0 MB	1.2.1	7	78.20	94.21

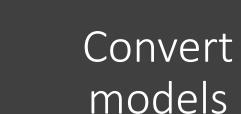
## Custom Vision Service: customvision.ai

1. Upload photos and label X Image upload Add Tags Uploading 2. Train Predic **Training Images** Performance **Training Images Performance Predictions** 4 images will b Delete Export Add some tag 3. Download ONNX model! Add a tag and press enter fruit X Choose your platform **ONNX** ONNX















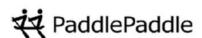






















### Convert models

1. Load existing model

2. (Convert to ONNX)

3. Save ONNX model



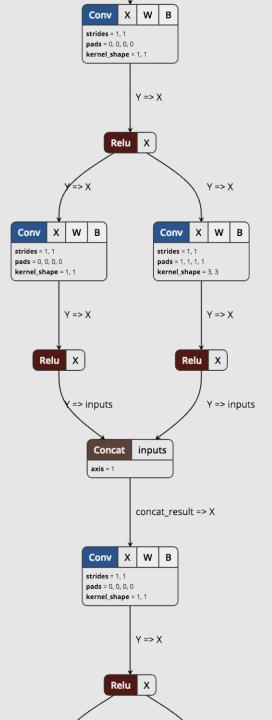
### **ONNX Models**

Graph of operations

Netron

https://netron.app/

https://lutzroeder.github.io/netron/



# Convert models: O PyTorch

```
import torch
import torch.onnx

model = torch.load("model.pt")

sample_input = torch.randn(1, 3, 224, 224)

torch.onnx.export(model, sample_input, "model.onnx")
```



```
In [ ]: import onnxmltools
        from keras.models import load model
In [ ]: # Update the input name and path for your Keras model
        input keras model = 'model.h5'
        # Change this path to the output name and path for the ONNX model
        output onnx model = 'model.onnx'
In [ ]: # Load your Keras model
        keras_model = load_model(input_keras_model)
        # Convert the Keras model into ONNX
        onnx model = onnxmltools.convert keras(keras model)
        # Save as protobuf
        onnxmltools.utils.save_model(onnx_model, output_onnx_model)
```

# Convert models:



- > python -m tf2onnx.convert
  - --saved-model tensorflow-model-path
  - --output model.onnx



### Convert models:



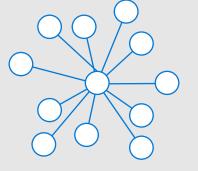
```
# Train a model.
from sklearn.datasets import load iris
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
iris = load iris()
X, y = iris.data, iris.target
X_train, X_test, y_train, y_test = train_test_split(X, y)
clr = RandomForestClassifier()
clr.fit(X train, y train)
# Convert into ONNX format
from skl2onnx import convert sklearn
from skl2onnx.common.data_types import FloatTensorType
initial_type = [('float_input', FloatTensorType([None, 4]))]
onx = convert_sklearn(clr, initial_types=initial_type)
with open("rf_iris.onnx", "wb") as f:
    f.write(onx.SerializeToString())
```

# ONNX as an intermediary format

- Convert to Tensorflow for Android
  - Convert a PyTorch model to Tensorflow using ONNX
- Convert to CoreML for iOS
  - https://github.com/onnx/onnx-coreml
- Fine-tuning an ONNX model with MXNet/Gluon
  - https://mxnet.apache.org/versions/1.3.1/tutorials/onnx/fine\_tuning\_gluon.html

https://github.com/onnx/tutorials

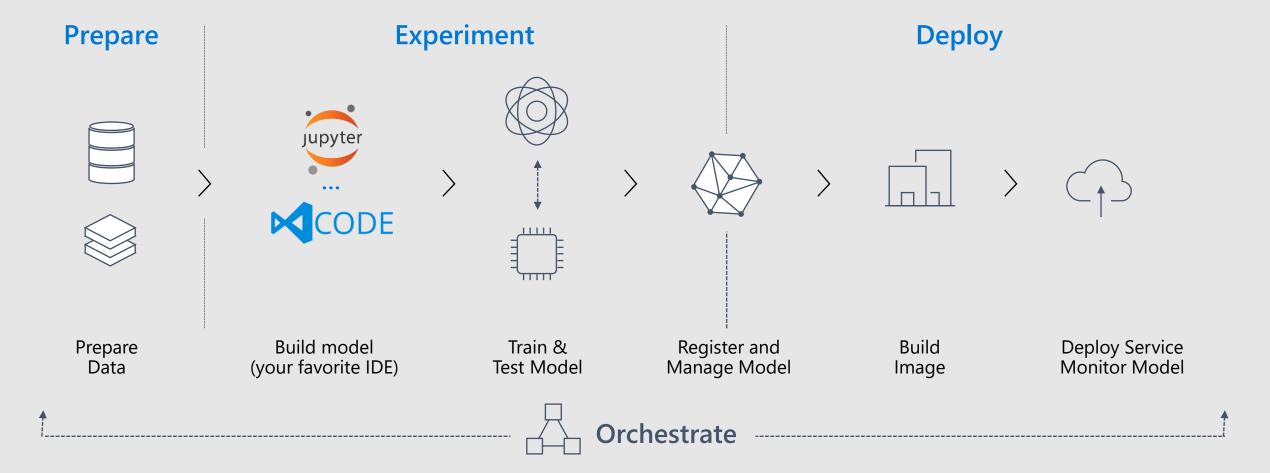




Experiment locally then quickly scale with GPU clusters in the cloud

- Use automated machine learning and hyper-parameter tuning.
- Keeping Track of experiments, manage models, and easily deploy with integrated CI/CD tooling

# Machine Learning Typical E2E Process











### Create

### **Frameworks**





























Converters

Native



**ONNX Model** 

### Deploy

#### Azure

Azure Machine Learning services

Ubuntu VM

Windows Server 2019 VM

Windows/Linux Devices

**IoT Edge Devices** 

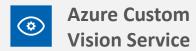
Converters

Native

support

**Other Devices** (iOS, etc)

### **Services**



support



## Deploy in Cloud

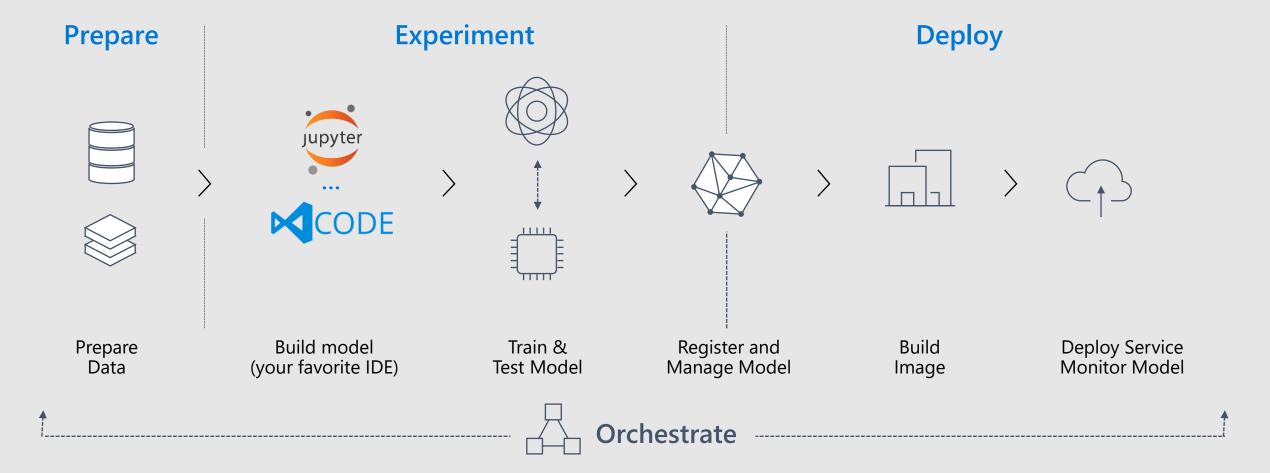
Model management services

- Deploy as web service
- Capture model telemetry



Azure Machine Learning

# Machine Learning Typical E2E Process





# ONNX Docker Image

onnx-base: Use published ONNX package from PyPi with minimal dependencies.

onnx-dev: Build ONNX from source with minimal dependencies.

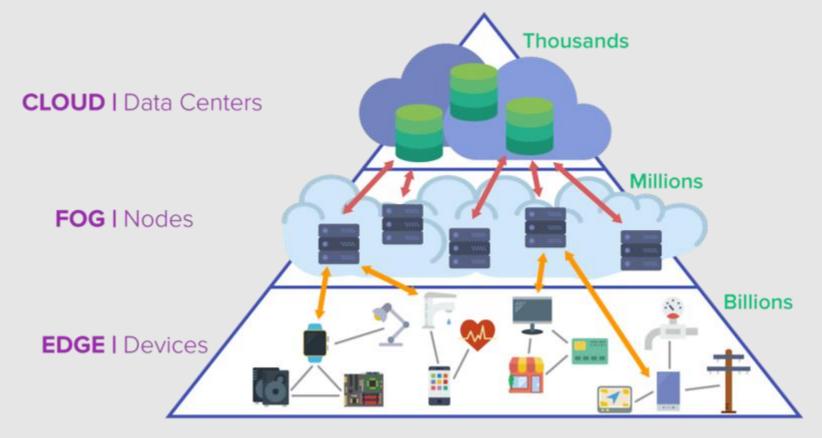
onnx-ecosystem: Jupyter notebook environment

- getting started quickly with ONNX models
- ONNX converters
- inference using ONNX Runtime.

### Caffe2/PyTorch Docker

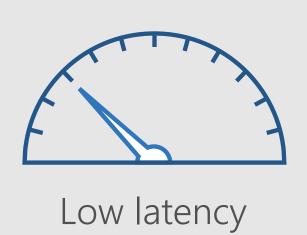
docker run -it --rm onnx/onnx-docker:cpu /bin/bash

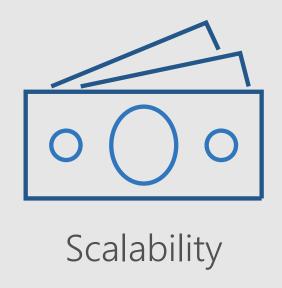
# What is the Edge?

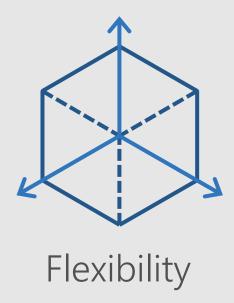


**Imagimob AB** 

# Al on the edge







#### ONNX Runtime

- High performance inference engine for ONNX models
- Founded and Open Sourced by Microsoft under MIT License
- Supports full ONNX-ML spec
- Extensible architecture to plug-in hardware accelerators
- Ships with Windows 10 as WinML
- onnxruntime.ai

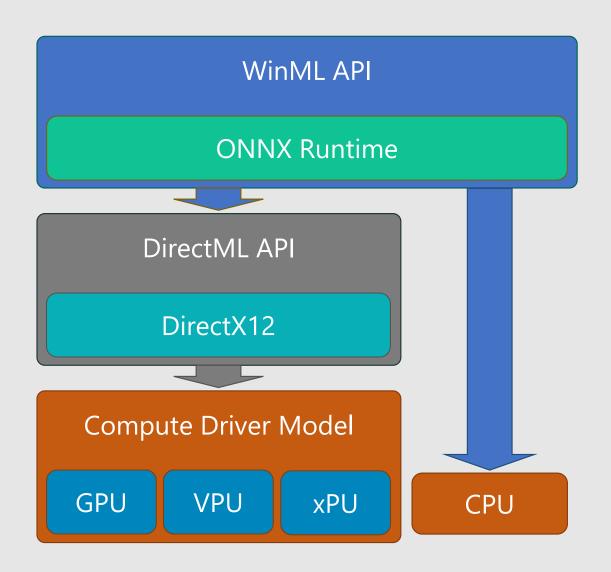


## **ONNX** Runtime

#### **Get Started Easily**

Optimize Inferencing	Optimize Trainii	ng									
Platform	Windows	dows		Мас		Android		iOS		Web Browser (Preview)	
API	Python	C++	C#		С	Java		i	Obj-C	Obj-C WinRT	
Architecture	X64		X86		ARM64		ARM32		I	IBM Power	
	Default CPU	Default CPU			DirectML		oneDNN		(	OpenVINO	
Hardware Acceleration	TensorRT	TensorRT		NNAPI		ACL (Preview)		ArmNN (Preview)		CoreML Preview)	
	MIGraphX (Preview)	•			Rockchip NPU (Preview)		Vitis AI (Preview)				
Installation Instructions	Install Nuge Refer to <b>doc</b>		ge <u>Microsoft.ML.</u> quirements.	OnnxRu	ıntime. <u>Gpu</u>						

## Windows AI platform



- WinML
  - Practical, simple model-based API for ML inferencing on Windows
- DirectML
  - Realtime, high control ML operator API; part of DirectX family
- Compute Driver Model
  - Robust hardware reach/abstraction layer for compute and graphics silicon

# ONNX Runtime JavaScript

# Node.js binding

Web

React Native

# ONNX Runtime Node.js

- Node.js binding
- ONNX model inferencing
- Electron
- Uses web assembly

#### Install

```
# install latest release version
npm install onnxruntime-node
```

#### **Import**

```
// use ES6 style import syntax (recommended)
import * as ort from 'onnxruntime-node';

// or use CommonJS style import syntax
const ort = require('onnxruntime-node');
```

# ONNX Runtime Web (ORT-Web)

- JavaScript library for running ONNX models on browsers
- adopted Web Assembly and WebGL technologies
- optimized ONNX model inference runtime for both CPUs and GPUs.

#### Install

```
# install latest release version
npm install onnxruntime-web

# install nightly build dev version
npm install onnxruntime-web@dev
```

#### **Import**

```
// use ES6 style import syntax (recommended)
import * as ort from 'onnxruntime-web';

// or use CommonJS style import syntax
const ort = require('onnxruntime-web');
```

# Why inference in the browser



It's faster



**It's safer** and helps with privacy



It works offline



It's cheaper

# Why not in the browser?





THE MODEL IS TOO LARGE AND REQUIRES HIGHER HARDWARE SPECS.

DOWNLOADED ONTO THE DEVICE

## **React Native**

- score pre-trained ONNX models
- ONNX Runtime Mobile
- light-weight inference solution
- Android and iOS

#### Install

```
# install latest release version
npm install onnxruntime-react-native
```

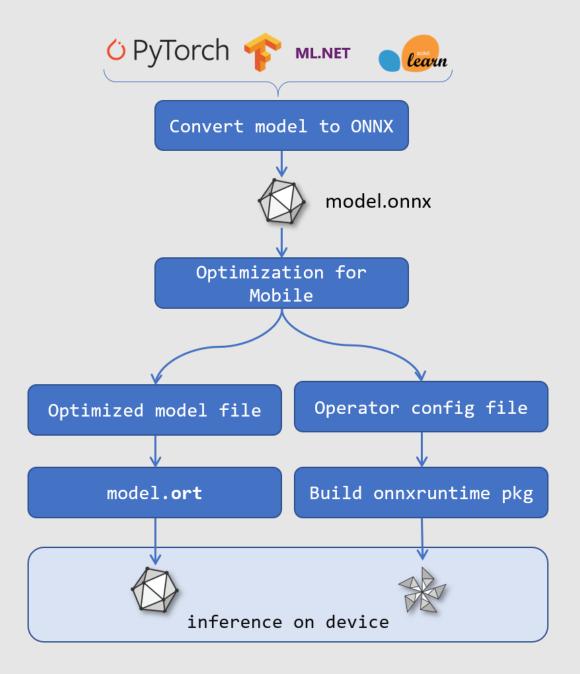
#### **Import**

```
// use ES6 style import syntax (recommended)
import * as ort from 'onnxruntime-react-native';

// or use CommonJS style import syntax
const ort = require('onnxruntime-react-native');
```

## **ONNX Runtime Mobile**

- minimizes the binary size
- pre-optimized ONNX model to an internal format ('ORT format model')



# Compatibility Chart

#### Compatibility

OS/Browser	Chrome	Edge	Safari	Electron	Node.js
Windows 10	wasm, webgl	wasm, webgl	-	wasm, webgl	wasm
macOS	wasm, webgl	wasm, webgl	wasm, webgl	wasm, webgl	wasm
Ubuntu LTS 18.04	wasm, webgl	wasm, webgl	-	wasm, webgl	wasm
iOS	wasm, webgl	wasm, webgl	wasm, webgl	-	-
Android	wasm, webgl	wasm, webgl	-	-	-



https://bit.ly/onnxportable



## Recap

✓ What is ONNX

ONNX is an open standard so you can use the right tools for the job and be confident your models will run efficiently on your target platforms

✓ How to create ONNX models
 ONNX models can be created from many frameworks

✓ How to deploy ONNX models

ONNX models can be deployed with Node, Web Browser, React Mobile using high performance ONNX Runtime

### **About Me**

#### Ron Dagdag





#### Ron Lyle Dagdag

Immersive Experience Developer Cell: 682-560-3988 ron@dagdag.net



www.dagdag.net
@rondagdag

Experience AR http://ron.dagdag.net

Lead Software Engineer at Spacee

5<sup>th</sup> year Microsoft MVP awardee

Personal Projects www.dagdag.net

Email: ron@dagdag.net Twitter @rondagdag

Connect me via Linked In www.linkedin.com/in/rondagdag/

Thanks for geeking out with me about ONNX

https://linktr.ee/rondagdag

## Hackster Portfolio

#### www.dagdag.net @rondagdag

