



Tell us what your idea is.

Describe in 250 words what the feature or service will do and how you'll use Machine Learning to push the bar:

Smartphones are the world's largest Augmented Reality platform. Google's ARCore and Apple's ARKit have made immersive experiences accessible to anyone with a recent smartphone. To create a realistic experience, understanding the user's physical environment is crucial, and these frameworks detect real-world planes, images, and faces. However, the world is a wonderfully chaotic place, and no two environments are the same. EnviARment allows AR to adapt its content to fit any room, park, or store just like it was designed for that space, seamlessly integrating with real objects.

Reimagine a biology classroom. Jumbo cross sections of plant cells rest on each student table. Photons stream through the windows in golden rays as students witness photosynthesis firsthand. Upon closer inspection, a basil plant on the teacher's desk shows the process on the plant level. An overview of the experiment is available on the whiteboard, and attached to the bookshelf are materials that dive deeper into the science behind the concept. Nobody measured the classroom or arranged these virtual objects themselves; in fact, the same experience works in a school halfway around the world with no additional setup.

EnviARment will use machine learning to detect specific physical objects and features to accurately anchor virtual content in appropriate places. This scene understanding applies to many concepts including games like AR soccer or an AR scavenger hunt. With semantic segmentation, EnviARment can identify continuous surfaces like grass, concrete, floors, and walls, enabling AR indoor and outdoor games that adapt to the amount and layout of available space.

Tell us how you plan on bringing it to life.

Describe where your project is, how you could use Google's help in the endeavor, and how you plan on using On-Device ML technology to bring the concept to life. The best submissions have a great idea combined with a concrete path of where you plan on going, which should include:

- (1) any potential sample code you've already written,
- (2) a list of the ways you could use Google's help,
- (3) as well as the timeline on how you plan on bringing it to life by May 1, 2020.

EnviARment has many applications including education, retail, gaming, and utilities. Being able to attach



virtual content to specific real objects makes AR much more useful and easier to integrate. Currently, the only way to anchor content on a specific item is recognizing a unique image or a prescanned 3D object. EnviARment leverages machine learning to ensure, for example, any nightstand can serve as a real world anchor for virtual content.

EnviARment is currently in the beginning prototyping stage. I plan on using the TensorFlow object detection API and transfer learning to train a model that detects objects to use as content anchors. To cover many objects and variations, I could use Google's help gathering a large and diverse dataset of images. [DeepLab](#) for semantic segmentation is a perfect architecture to detect the extent and location of surfaces including indoor floors, grass, and concrete. From a past project, I know converting DeepLab models to TFLite can be challenging, and would appreciate assistance running that model through the new model optimization and conversion tools available. While the 2D object bounding boxes and hit test distance information is sufficient to place virtual content on a detected object, I would like to investigate if Google AI's [research](#) in depth estimation from video can be applied to EnviARment for more accurate AR placement. Access to computing resources, especially for training the DeepLab model, would be very helpful.

To initially bring the concept to life, I'm choosing an educational experience as the first demo of EnviARment's capabilities. The rough outline of steps include:

Machine Learning

1. Collect and label a dataset of objects in indoor scenes, possibly from existing datasets or new data (2-3 weeks)
2. Retrain an object detection model with this dataset (1 week)
3. Label the dataset of indoor scenes with segmentation masks including the classes: floor, wall, door, window, table (1-2 weeks)
4. Retrain a DeepLab model on this dataset (1 week)
5. Convert the models to TFLite and run inferences with the TFLite GPU delegate in a sample Android app (1 week)
6. Conduct real world tests and collect more data to update the models if necessary (1-2 weeks)

The app (Augmented Reality)

Build a working prototype using existing object detection and segmentation models while the custom dataset and models are developed and training.

1. Use the TFLite GPU Delegate to quickly run inferences on both models (3-5 days)
2. Calculate the 3D locations of objects and surfaces using the predicted bounding boxes and masks, adding depth with hit testing in ARCore (3-5 days)
3. Leverage ARCore and Sceneform to render virtual content at the calculated positions (1-2 weeks)
4. Design the UI and AR interface for the educational experience (3-4 weeks)
5. Solicit feedback and update the UI accordingly (3-4 weeks)

When the custom models are trained, they will replace the pretrained models.



Tell us about you.



[Pi AR on the App Store](#)

[Mav World on the App Store](#)