the weekly anthropocene



Dispatches From The Wild, Weird World Of Humanity And Its Biosphere

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Niger



Since the 1980s, the people of the West African republic of <u>Niger (see map) have</u> <u>increased their nation's forest</u> <u>cover by over 200 million</u> trees, covering at least 12 million acres-an area larger than Massachusetts and Connecticut combined. (For context, Niger itself is slightly larger than Texas and California combined). *The*

Weekly Anthropocene has touched on this story before, but it's worth discussing further because of both the scale of the accomplishment and the extraordinary context in which it occurred. Niger is one of the poorest countries in the world, a landlocked, subsistence agriculturedependent state spanning the edge of the Sahara and the Sahel that was ranked 189th out of 189 in the UN 2018 Human Development Index. And yet this incredible reforestation was entirely a grassroots campaign conceived of and executed by



ordinary Nigerien citizens, not international aid or a government-led initiative. Indeed, Western scientists only realized it was happening in 2004. The Nigerien government had actually tried a reforestation program in the 1970s, but it failed, with fewer than 20% of the 60 million trees planted surviving. The modern movement was likely sparked in 1983 by a group of farmers who'd traveled to look for work, and didn't return in time to clear their fields of young trees before the rainy season. When they planted around them, they noticed that the crops grew better, fertilized by falling leaves, shielded from sand and sun by the canopy overhead, and moistened by evapotranspiration. This result spread by word of mouth to other local farmers, and was popularized further by Australian missionaries and American Peace Corps volunteers. Notably, the method didn't require planting any trees, which would have meant seedling nurseries and seed collection infrastructure, but simply encouraged farmers to let alone the naturallyregrowing trees amidst their fields. (Pictured at top: herders watering their cattle on a farm in Niger where trees were allowed to naturally regrow). It spread across the country like wildfire: ecologists studying the phenomenon state that "200 million trees" and "12 million acres" are very low-end, conservative

state that "200 million trees" and "12 million acres" are very low-end, conservative estimates, <u>based on aerial imagery analysis</u> since there's no central project database tracking progress on the ground. It's estimated that the newly tree-rich areas of southern Niger have produced an <u>extra half million tons of grain per year, enough to feed 2.5 million people</u>, and kept up a surplus even during the drought of 2011. "Before, farmers often had to sow their crops two, three times after they were destroyed by strong winds that covered the crops with sand," said Aichatou Amadou, a Nigerien farmer from Droum in the Zinder region, to *National Geographic*. "I only have to sow my crop once." Now, aid groups are sending farmers from other countries to Niger to learn about this ultra-low-cost regeneration method, and similar stories of farmer-led agroforestry movements and popping up in Burkina Faso, Mali, and Malawi.

This is a truly inspiring saga of some of the most adversity-stricken people on Earth using their wit and their will to create a better life, and a better world. Great news!



Los Angeles



On April 22, 2022, an array of California dignitaries held a groundbreaking ceremony to mark the start of construction on the Wallis Annenberg Wildlife Crossing. (Pictured: an artist's rendering of the completed crossing). The envisioned forested bridge will be 165 feet wide and span 200 feet over U.S. Highway 101, with 12-foot-high fencing on the sides shielding wildlife from the road and vice versa. It's set to be completed by early 2025, and will likely be the largest wildlife crossing in the world! (Similar projects in Canada's Banff National Park have been wildly successful, reducing elk-vehicle collisions from 100 per year to nearly zero). The crossing's total cost will be \$90 million, \$25 million of which is from the Annenberg Foundation and most of the rest from thousands of other philanthropists. The California state government is overseeing design and construction, but is providing no funding.

This is great for all the local critters, but is particularly excellent news for the imperiled cougar (*Puma concolor*) subpopulation of Los Angeles County. The Santa Monica Mountains cougars have <u>lost at least 20 big cats</u> to collisions with vehicles since 2002-one of which was killed just one day before the groundbreaking ceremony. This wildlife crossing should give them safe passage to the Simi Hills and Santa Susana Mountains subpopulation to the north, which they're currently completely cut off from by Highway 101. This in turn should help keep LA's cougars genetically viable, as some of the cats have been displaying signs of inbreeding recently.

This project is a spectacular example of many of the greatest trends in Anthropocene conservation: innovative problem-solving, public-private partnerships, and humans making the effort to give wildlife the chance to thrive alongside them, sharing the landscape. Great news!



Scientist Spotlight: An Exclusive Interview with Dr. Christopher Anderson

Dr. Christopher Anderson is an environmental data scientist who has conducted extensive research with satellite imagery. Dr. Anderson is currently a researcher at Stanford University's Center for Conservation Biology. He is also the cofounder, CTO, and "resident expert in biodiversity mapping, plant ecophysiology, and satellite image analysis" of Salo Sciences, a conservation technology company. Dr. Anderson's personal website is at <u>earth-</u> <u>chris.github.io</u>, his <u>Google Scholar page is here</u>, and Salo Sciences' website is at <u>salo.ai</u>.

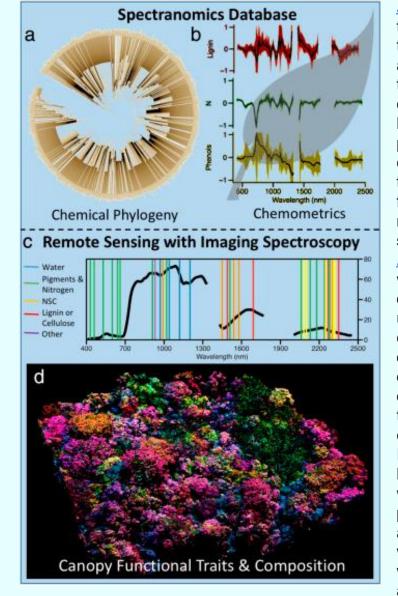


A lightly edited transcript of this exclusive interview follows. This writer's questions and remarks are in **bold**, Dr. Anderson's responses are in regular type. *Bold italics* are clarifications and extra information added after the interview.

Hi, Dr. Anderson! It's an honor to speak with you. I have many questions, but first, can you tell our readers a bit more about yourself? Why did you choose to become an environmental data scientist, and why specifically focusing on the mapping and satellite image analysis track?

Well, I think a lot of it came about from my experience learning to love ecology as a science. I grew up in the redwoods and that was always a sort of key part of my identity as a person. But it wasn't until process of elimination as an undergraduate that I came to ecology as a discipline. I always liked mathematics and science, and that put me into the biology track. It wasn't until I was taking courses in ecology and physiology that I came to appreciate the beauty and the profundity of the natural environment. It actually took me the experience of failing my plant physiology class as an undergrad to kick me into high gear, that was the only course I ever failed. What made [ecology] really compelling was that it was something I could see in my own daily lived experience. I was an undergraduate at <u>UC Santa Cruz</u>, which is just a beautiful campus and a great place to get excited about ecology and about the dynamics of earth's ecosystems. So it was just able to reinforce a lot of the things that I was learning, and thinking about it just became a real holistic part of how I understood the world, and that was really exciting to me.

It wasn't until a couple years after I graduated where I got a job at the Carnegie Institution for Science. I had a friend there who was working on a research project called <u>Spectranomics</u>, looking at the relationship between leaf chemistry, leaf structure, and how you would measure that from satellites. (*Pictured: graphic from full paper on*



spectranomics). And so what they were doing is they were traveling to tropical forests all around the world, climbing to the top of trees, bringing down branches, stripping the leaves off, cleaning them, putting some in the oven to dry them, and grind them up taking some samples. Then freezing them like with liquid nitrogen and sending all that stuff back to the lab. (See *more on this here*). Our job was to then run a series of chemical analyses to understand things like defense compound concentrations, leaf nutrient concentrations, growth compounds, leaf structure, all that kind of stuff. I was offered that position because I had a strong background in leaf biochemistry. When they were telling me about the project, it was like "Oh yeah actually we're flying airplanes with lasers on them. And we're shooting them to trees and we're measuring the

amount of light that's absorbed and reflected by plants and measuring how much nitrogen is in the tree from airplanes." And I went, "What are you talking about, that's crazy!" Like that's so cool. I didn't know any of this technology existed until I was exposed to it by happenstance. I was paid 10 bucks an hour at first, eventually 15 bucks, and then I worked hard enough in the leaf chemistry analysis that as they were growing the program for the airborne data collection, they needed people to come on and do some of the actually flying on the airplanes to collect the data, and process it downstream afterwards. And I just got lucky, they liked the work that I was doing. I was willing to put in the effort and during nights and on the train was commuting down to Stanford I was just reading papers on how remote sensing works. It was really, for me, learning on the job. And being so lucky to get the opportunity to do crazy cool things and start getting my hands on the data, to say nothing at how cool it was to fly airplanes around the world shooting lasers at trees.

How did you get involved with cofounding Salo Sciences, and can you tell our readers exactly what that company does?

<u>Salo Sciences</u> is a climate technology company. The simplest description is we are very good at analyzing pictures of trees taken from space. One of the motives for founding a company was that we saw that there is this incredible technology [remote sensing] that is being used, often in sort of academic and research context, to be able to answer questions like:



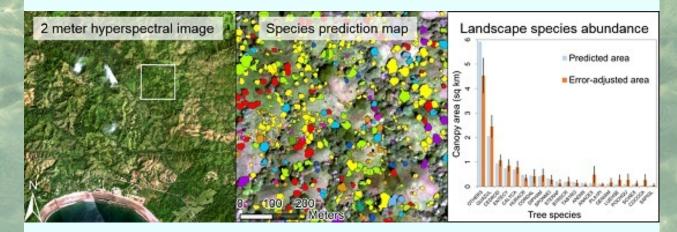
How is biodiversity changing or distributed across large spatial scales? How are carbon stocks changing in response to disturbance and wildfire, how can we map them across large areas. But it seemed like a lot of this was-it's not a particularly original critique-often funded for the purpose of research, a few people get the benefits of understanding how to work with the information and publish the results, and then it will kind of sit there in a database or be published as a paper. And we decided that there were lot of opportunities to use this information and this technology to help inform a lot of the large-scale conservation challenges around the world.

So we wanted to build something, from the ground up. Our own version of NASA, for mapping the earth, really focused on trees, carbon, biodiversity, and how they change. I would've worked for a global organization doing it if one existed. You'd think it would be a global governance system, but it doesn't really exist. We did what most people do when the thing they want doesn't exist, we had to create it ourselves. So that was how we set out to build Salo to be the premier forest monitoring technology company. And it's taken a long time to build it from scratch and to do that, from the perspective of being a private organization as opposed to a public institution. We don't take outside investment, because that would be viewed as swaying our neutrality as an organization. We've tried to really build the organization from the ground up and bootstrap it based on the needs and demands of the people who need the products that we create. **That's really inspiring, an incredible vision**.

You coauthored a very interesting paper a few years back, <u>Tree species</u> <u>abundance predictions in a tropical agricultural landscape with a supervised</u> <u>classification model and imbalanced data</u>." I've done some supervised classifications myself-can you tell me more about your process here and your results for the Panamanian landscape you analyzed?

A lot of that was led by Sara Graves, a Ph.D. student at the University of Florida, now at the University of Wisconsin as a professor. At the time there was a lot of effort into developing novel species classification algorithms. Understanding the spatial distribution and tree species is really important for a lot of different applications, like the rate at which they accumulate carbon, how they support wildlife, just for the value of the information. The airborne imaging spectroscopy data is one of the best resources for mapping tree species, as a remote sensing resource. These datasets measure the light reflected by vegetation, from the visible spectrum further into the infrared spectrum. And that's really valuable information. Say you have two different tree species, one [has leaves that are] really bright green, one is really dark green. The really dark green species is absorbing a lot more red and blue light, the bright green is reflecting a lot more light. Ultimately, that's a function of the amount of chlorophyll that's in each tree. More chlorophyll, darker green trees. And as you move further into the spectrum into the near infrared, shortwave infrared, there are other things like how much nitrogen there is, how much cellulose they have, how many leaves are on the

tree, the total water content, this is all information that you can derive from that spectroscopy signal. What makes this really critical is that different species have different underlying chemical fingerprints; there's a consistent chlorophyll distribution, consistent nitrogen use, consistent water content within trees of the same species. So put those things together, and you can use that information in the spectral signal to discriminate between different tree species.



(*Pictured above: graphical abstract of paper*). And so, one of the big challenges is being able to get enough data to make enough observations on the ground to correctly say that we have this many species, here's the sort of representative fingerprint of the data sets that they occupy and how we can try to scale out over large areas. One problem that you run into in data science that's really common is the issue of imbalanced data. Having some class really well represented, other classes really poorly represented. This is fairly classical in ecology, because there are all sorts of exponential distributions where you tend to have species that are very common and species that are very rare. So, how would you come up with a data science based approach to address some of that imbalance, where you have some species with a lot of information and some species with very little information? That was the goal of the paper, to try to think about how you might understand these data sets and how you might choose the algorithm that you use to parse through that data to handle some of those different challenges. And then what you can get as a result is a map of tree species across agricultural landscapes, which helps predict carbon storage on them and the wildlife available in those areas.

Can you tell me more about your and Salo Sciences' current California Forest Observatory <u>project</u>? What's it like working with Planet and their cubesats? How are you planning to work with communities to make this data useful? What languages and platforms are you coding in-JavaScript, QGIS, Leaflet?

The Forest Observatory product itself is really three primary components. The first is the underlying datasets themselves. This information is 3 meter spatial scale, which is accessible through our partnership with Planet. And these are maps of the patterns of forest structure that drive wildfire behavior. So things like, how tall the tree is, what is the density of <u>ladder fuels</u>-which is the vegetation that's on the floor of the canopy and could lead to surface fire transitioning into a <u>canopy fire</u>-as well as the distance between the forest floor and the lowest layers in the canopy. And then also a surface field model that predicts things like spread rate and how tall flames could be, and maps that across the whole state. It's updated on a regular basis using satellites.

Second component of it is an <u>API</u> (*application programming interface*), which I didn't really know what it was until I had to build one myself. I think that's more or

less the lesson of all the stuff that we do, it's really just about being curious and capable about learning new technologies as they come up. The API is a way to allow command-line or programmatic access to the datasets we have. Instead of downloading a 10-gigabyte <u>GeoTIFF</u> (*big map image*), you can just say, "Oh, I need to run <u>zonal statistics</u> in this area, can I get the value for this latitude and longitude? Can I actually just show the web map tiles, display this on a web map?" So the API is the way you can do that and get more fine-grained control over the datasets.



Third component is the user interface, <u>forestobservatory.com</u>. (*Available to all at forestobservatory.com*, *pictured above*). The goal for that was to provide good visual access so that people were able to understand the datasets, to visualize how the patterns of forest structure change over space and how they interact with the other drivers of wildfire behavior. We decided there was a really good opportunity to create more or less Google Maps for wildfire risk. You can just look at something and it will tell you here's where I am, here's the fuels around the landscape, here's what fire parameters are and here's where the wind speeds are blowing. There are lot of choices there you have to make, in terms of what sort of information you represent, how you represent it.

It came together in partnership with <u>Planet</u> (*pioneering private satellite company* <u>*Planet Labs*</u>), because after the <u>Camp Fire in 2018</u> (*the most expensive natural* <u>*disaster in the world that year, killing at least 85 people and destroying at least*</u>

<u>18,000 structures</u>) and the major destruction and devastation of the past couple of wildfire seasons, everyone was like "Holy shit, this is an all hands on deck moment." We need to be able to bring all this technology together. We need to improve our resilience and to improve our understanding of these dynamics and help us be better prepared for the future. And there are very good opportunities for better data to help us do that.

That's obviously not the whole story, but it came together. It was supportive. We were leading a lot of the technical development, <u>Vibrant Planet</u> (*a different company from Planet Labs*) was doing a lot of outreach.

And part of why their involvement was so critical is that we needed to do interviews with a series of stakeholders in wildfire science. New data products can come and go all the time, but people need to know: does this data solve my problem? Is it designed in a way that I can access it? Does it tell me what I need to know? Do I trust these people? We just had to work through a long process of interviewing 70 people to figure out what the data needs were. How they would use it, how we should present it, timescales and all that sort of stuff. So that was how we built the system, and since then we've been working on expanding the geographic coverage. We started in California as pilot. And we are expanding this summer to all of the Mountain West, this is eleven states of Colorado and points west. And we've been doing some work in Australia as well, learning about new fire models Down Under.

What major software and programming language are you using and what skills are you looking for at Salo Sciences right now?

We mostly do a lot of work in Python for our software development, and all of our computing resources are built on top of <u>Google Cloud</u>. That's the alternative to AWS; Google Cloud platform has a series of cloud storage features, so you can host data in public buckets and access it in cloud native data formats. Otherwise we mostly built everything as a custom software system, using TensorFlow, rasterio and gdal. Geopandas is great, that's super cool. The thing we experience in geospatial is that every problem is a custom problem, it's hard to build things that are one size fits all. There are some particular dynamics to geospatial data, where there are a lot of deep learning packages in TensorFlow for doing image recognition, but they're all based on RGB data, like maybe three channels of information. And a lot of stuff is hard coded for how you would do image sampling and resizing. And you don't want to mess with that, different things emerge if you resample ecological data or remote sensing data, so we've built our own custom in-house software systems to be able to work with a lot of that information. We're kind of remote sensing generalists, we work with hyperspectral data, LIDAR, JEDI, with field data. So being confident in a lot of different technologies and knowing some of the advantages and disadvantages of them in ways to pull them together, that's sort of the bread and butter. We've grown relatively slowly, but we are intentional with the people that come on board. Being willing to learn new technologies and having a good enough grasp to get started tends to work pretty well. We're becoming a bit more specialized now as we are developing more specific data products, but that's been sort of our set priorities. Nobody has Google Cloud experience, everyone comes from AWS, so that's often a learning experience for people, but you know we love [Google] Earth Engine. (This writer loves Earth Engine too and has used it a lot recently-it's a really cool, mostly free, global satellite imagery platform). Earth Engine is amazing. It was such a revelation when it first arrived, we're huge fans of Earth Engine, but commercial support is limited, and we spent so much time poking at the individual pixels in our datasets that we need to be operating at that really low level. And earth engine requires a lot of abstractions to be able to understand the data sets so that hasn't been our priority for building out sort of scalable solutions, vet.

Thank you so much for sharing your wisdom today!

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