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AT ALL SCALES OF LIFE

In 1930, the Italian biologist D'Ancona had a problem. D'Ancona was an Italian biologic statistician who studied the abundance of fishes in the Mediterranean during World War I. He had been asked to increase fish stocks, but rather than increasing the number of all fish, the increased fishing had had ecological effects on fish populations. The fished fisheries seemed to have more large predators—like shark and dolphin, whose numbers went up—while smaller species, like “sand eels” (the anchovies and sea bream), whose numbers were once abundant, disappeared. When the fisheries recovered, the effects reversed. Frustrated, D'Ancona turned to mathematics to try to understand the mathematics behind patterns that corresponded to what he saw. In 1931, D'Ancona and Gause (1933) in a journal of fisheries—was the catalyst for a revolution in ecology.

What D'Ancona did was to start the first mathematical model of a predator-prey system. He hypothesized that in the absence of predators, prey would grow exponentially, while predators would have high mortality rates due to the lack of food. When there was enough food, the population would stabilize—“zero growth” (see also Lotka-Volterra, 1926). Thus, when prey were abundant, predators could flourish; if the prey grew in number, they would deplete prey resources, so the prey population would start to die off, enabling prey to implement a “gatekeeper” mechanism that systems can oscillate without external oscillatory predators. Indeed, the model predicted that endogenous and non-linear feedback—specifically, the fact that the population frequency was a product of the prey and predator populations—was the key to solving his “shoal-in-the-bell” puzzle, in that it predicted that prey populations would fluctuate while predator populations would go up to a function of the prey, which would then fluctuate.

This model was developed and developed in parallel by the Austrian physicist Alfred Lotka and the Italian biologist Vito Volterra. This model, now known as the Lotka-Volterra model of predator-prey dynamics, became one of the most famous mathematical models of all time. It predicted exponentially parabolic rates of change in the early twentieth century, and it has since become a harbinger of a far broader phenomenon. Yet the model also has some interesting quirks. If the system is initialized with two different initial conditions entered, a fact that is, of course, to begin with the starting numbers of predators and prey, the model predicted oscillations that would return precisely to that initial point. Similarly, if the system had started with no predators, then the model would again predict oscillations that would return to that new starting point. Formally speaking, these are termed “ergodic” systems, because they are systems that have conserved quantities in the system. Such conservation laws may be important, but to this point they point to pathologies in the ecological model. In 1925, Lotka and Volterra showed how predator-prey dynamics could