

**Universidade Estadual de Goiás**  
**Campus de Ciências Exatas e Tecnologia – Henrique Santillo**  
**Pró Reitoria de Pesquisa e Pós-Graduação**  
**Programa de Pós-Graduação *Stricto Sensu* em Recursos Naturais do Cerrado**

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**Investigando o impacto das mudanças climáticas sobre a produtividade e o serviço de  
polinização em *Caryocar brasiliense***

Anápolis, 2018.

PABLO HENRIQUE DA SILVA

**Investigando o impacto das mudanças climáticas sobre a produtividade e o serviço de polinização em *Caryocar brasiliense***

Dissertação apresentada ao Programa de Pós-Graduação Stricto Sensu em Recursos Naturais do Cerrado, na Universidade Estadual de Goiás para a obtenção do título de Mestre em Recursos Naturais do Cerrado.

Orientador: Paulo De Marco Júnior

Anápolis, 2018.

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**Dados do trabalho**

Título: Investigando o impacto das mudanças climáticas sobre a produtividade e o serviço de polinização em *Caryocar brasiliense*

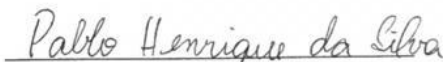
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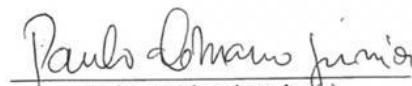
( ) Tese (X) Dissertação

Curso/Programa: Recursos Naturais do Cerrado

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Anápolis, 06/06/2019  
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Assinatura do autor (a)

  
Assinatura do orientador (a)

Elaborada pelo Sistema de Geração Automática de Ficha Catalográfica da UEG  
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SP113i Silva, Pablo Henrique da  
Investigando o impacto das mudanças climáticas sobre a  
produtividade e o serviço de polinização em Caryocar brasiliense / Pablo  
Henrique da Silva; orientador Paulo De Marco Júnior. -- Anápolis,  
2019.  
57 p.

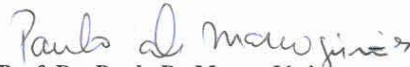
Dissertação (Mestrado - Programa de Pós-Graduação Mestrado  
Acadêmico em Recursos Naturais do Cerrado (RENAC)) -- Câmpus-  
Anápolis CET, Universidade Estadual de Goiás, 2019.

1. Mudanças climáticas. 2. Produtividade. 3. Economia. 4.  
Biodiversidade. 5. Serviços ecossistêmicos. I. Júnior, Paulo De  
Marco, orient. II. Título.

PABLO HENRIQUE DA SILVA

INVESTIGANDO O IMPACTO DAS  
MUDANÇAS CLIMÁTICAS SOBRE A  
PRODUTIVIDADE E SERVIÇO DE  
POLINIZAÇÃO EM *Caryocar brasiliense*

Dissertação defendida no Programa de Pós-Graduação *Stricto Sensu* em Recursos  
Naturais do Cerrado da Universidade Estadual de Goiás,  
para a obtenção do grau de Mestre, aprovada em 26 de fevereiro de 2019, pela  
Banca Examinadora constituída pelos seguintes professores:



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## **DEDICATÓRIA**

A todas as pessoas que acreditam que o conhecimento pode salvar o mundo.

## AGRADECIMENTOS

Gostaria de agradecer... Sim, eu gostaria, mas acho fundamental que vocês me conheçam antes de sair falando o porquê sou grato. Primeiro vou contar para vocês a parte engraçada da minha trajetória e por qual motivo sou lembrado até hoje pelos meus colegas de turma, e claro, por alguns professores. Sempre tive muitas dificuldades em fazer comunicações orais e lidar com o público, mas *“um belo dia resolvi mudar e fazer tudo o que eu queria fazer”*.

É obvio que não tive uma mudança excepcional, mas tudo tem um começo. E lá vou eu mais uma vez encarar o desafio de me apresentar em público. No Seminário de Apresentação de Projetos decidi que iria fazer uma piada modificada a partir de um vídeo viral, o qual eu adaptei para a minha proposta de projeto. A piada não funcionou com todo mundo (é claro!), mas quem sabia do que eu estava falando sorriu. Nesse momento eu não percebi que nem todas as pessoas estão a par do que já aconteceu na internet e me esqueci de esclarecer que era uma piada. Para essas pessoas eu dei um tiro no pé. No entanto, para não te deixar curioso sobre o tamanho dessa munição, é preciso dizer que a piada foi adaptada do vídeo da atriz “Global”, Suzana Vieira, e o seu talento indescritível. Enfim, peguei o meu roteiro de vida e segui em frente. Apesar da minha decisão de continuar com a minha vida, talvez seja possível que no momento em que você esteja lendo esse texto alguém esteja fazendo uma piada sobre o meu talento. Sim, possivelmente foi um erro, mas hoje eu sou grato a todas as piadas do mesmo gênero e assunto que eu tive que ouvir desde então.

Algumas coisas são extremamente necessárias, algumas pessoas também, mas tem aquela pessoa que faz a sua vida ter um significado grandioso. Para mim existe apenas uma pessoa e sem ela eu não conseguiria chegar onde estou. Vocês devem estar se perguntando quem é essa pessoa, e a resposta é o mais simples possível. Minha mãe dedicou uma vida inteira ao trabalho e aos filhos. Se hoje estou aqui é porque tive apoio, e continuo sendo apoiado todos os dias. Obrigado mãe! Sou muito grato aos seus esforços para que eu pudesse atingir os meus objetivos.

Da mesma maneira que o mundo não é binário e que não existe apenas uma alternativa correta para as nossas perguntas, é claro que o quadro de pessoas importantes não acaba por aqui. Durante a minha trajetória acadêmica tive a oportunidade de conhecer diversas pessoas, com pensamentos e maneiras diferentes de enxergar o mundo. Isso realmente teve um diferencial enorme na minha vida. Especificamente, no início do meu

mestrado surgiu uma pessoa fantástica. Ela é realmente excepcional! Acredito que me tornei o maior fã dela. Certo dia, estávamos cursando uma disciplina e decidimos fortalecer a nossa parceria fazendo o trabalho final em dupla. Durante o processo dissemos “no sábado a gente submete”. Por incrível que pareça estamos fazendo esse trabalho até hoje. Acho que merecemos uma camiseta com essa frase. Gracielle Higino, você vai mudar o mundo com a divulgação científica (\\o//). Sou muito grato por ter te conhecido e aprendido diversas coisas contigo.

“*Só para os loucos, só para os raros*”. A letra dessa música define essa pessoa. Mas afinal, por que eu sou tão grato? Aqui eu poderia listar os “*13 reasons why*”, mas não estou disposto a dar esse gostinho para ele. “*Confesso impressionado, nunca vi ‘ninguém’ igual. O ‘laboratório’ era um refúgio, um lugar ‘primordial’*”. Foi no laboratório TheMetaLand que tive o prazer de me aproximar dessa pessoa incrível. Nesse mesmo lugar, desenvolvi diversas habilidades que de uma maneira ou de outra foram estimuladas por ele. Acho que eu estava no lugar certo e com a pessoa certa, mas a jornada acabou. Sinto uma necessidade enorme de agradecer por todos os seus ensinamentos e perspectivas de vida. Hoje eu enxergo o mundo de uma maneira diferente e muito disso eu aprendi com ele. Sou muito grato ao meu orientador, o Dr. Paulo De Marco Júnior, por ser uma inspiração.

Aproveitando o gancho, gostaria de agradecer a galera mais sensacional que existe. Sou grato a todos do laboratório TheMetaLand por serem o que são. Com vocês eu tive conversas maravilhosas e contei absurdos sobre a minha vida. Vocês fazem parte da minha construção como indivíduo, e a ajuda de vocês me possibilitou atingir alguns dos meus principais objetivos. Obrigado por sempre estarem dispostos. Sou muito grato a todos vocês. Especialmente, gostaria de agradecer ao Santiago Velazco pelo suporte na obtenção dos dados de solo para esse trabalho.

Apesar da ordem dos parágrafos, esse foi um dos últimos que eu escrevi pelo simples fato de não saber o que dizer nele. Estava querendo contar uma história excepcional, mas acho que não há como selecionar apenas uma sobre essa pessoa. Foram tantos momentos juntos que já se tornou cotidiano, mas no final de tudo ela merece um prêmio por me tolerar todo esse tempo. Gostaria de agradecer a Rafaela Gonçalves pela sua amizade e dicas para ganhar descontos em aplicativos. Sou muito grato por todos os desafios que encaramos juntos. Fundamos uma Empresa Júnior e lidamos praticamente sozinhos com ela durante um ano, e a partir daí compartilhamos a nossa paixão pelo empreendedorismo. Até hoje nos aturamos. Muito obrigado por ter caminhado comigo durante todo esse trajeto.



Durante o mestrado tive a companhia de um amigo em particular. Ele conseguiu entender o meu nível de pobreza e me ajudou várias vezes. Sempre demos risadas das coisas mais ridículas que conversávamos. Ajudávamos um ao outro sempre que fosse necessário, principalmente em questões acadêmicas. Gostaria de agradecer ao Edgar Lima pela companhia durante esses dois anos. Ter alguém próximo e que compartilhasse o dia a dia comigo me ajudou a lidar com algumas crises acadêmicas. Muito obrigado!

Agora vou falar de uma pessoa que esteve comigo durante parte do meu desenvolvimento pessoal e que me ajudou bastante em questões relacionadas à língua inglesa. Gostaria de agradecer a Chris Borges por todas as conversas ácidas. Gosto tanto de conversar contigo que nem sequer vejo o tempo passar, quando vejo já se foi uma hora de procrastinação. Sou muito grato pela sua amizade e por borrifar repelente no meu corpo quando passeávamos pelo Pantanal.

Ainda não chegamos ao fim, mas estamos quase lá. Antes eu preciso agradecer a um grupo específico de amigos, o “Gangster Manjericão”. Nesse grupo de *whatsapp* tem uma galera incrível com pensamentos diversos. Sempre foi muito divertido me reunir com vocês e conversar sobre a vida. Muito obrigado por partilharem o tempo de vocês comigo. É muito bom quando estamos todos juntos. Sou muito grato a esse círculo de amizade.

Acima, eu contei sobre algumas pessoas que fazem parte da minha trajetória, mas agora vou falar das pessoas que entraram na minha vida da maneira mais divertida possível. “*Quem são eles? Quem eles pensam que são?*” Bom! Eles são exatamente quem eu gostaria que vocês conhecessem. Mas enfim, para não reforçar todo esse mistério, gostaria de agradecer a todas as bocas que beijei. Um beijo, uma história. Cada pessoa que eu me aproximei tem uma história fantástica por trás. A diversidade de pensamentos e modos de vida que me fizeram refletir sobre o meu posicionamento diante da sociedade brasileira atual. Hoje eu tenho também tenho uma história e vocês fazem parte dela. Muito obrigado!

Gostaria de agradecer ao Programa Próprio de Bolsas da UEG pelas duas bolsas concedidas durante o meu mestrado. Essas bolsas permitiram que eu chegasse até o final do mestrado sem muitas complicações financeiras.

Por fim, mas não menos importante, gostaria de agradecer ao Programa de Pós-Graduação em Recursos Naturais do Cerrado (RENAC) e a Universidade Estadual de Goiás por terem me acolhido durante esses dois anos. Claro que em alguns momentos eu preferia estar vendendo churros na feira, mas não consegui herdar a barraquinha do meu avô. O mais

importante é que não desisti de uma das minhas principais ambições. Sou grato por todas as amizades que construí e pelas risadas que demos no corredor. Acredito que fomos a turma mais barulhenta que já passou por ali. Sou grato ao RENAC por isso.

*I'm convinced I can swim  
I'm pretty sure that I have a fin  
I dream about going in a pool  
I kick and splash and I look real cool  
Okay, I've never seen a pool  
But I'd like to visit one with you  
Yeah, I'm sure it's a fin  
I'm convinced I can swim*

*Art Smelly*

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## RESUMO

A produção de frutos em cultivos agrícolas é resultado de diversos fatores externos que contribuem para o aumento da produtividade ao longo do tempo. Por exemplo, o serviço ecossistêmico de polinização realiza a troca de gametas entre indivíduos de plantas, aumentando a taxa reprodutiva dessas espécies. No entanto, as mudanças climáticas podem alterar a distribuição de plantas e polinizadores, ocasionando uma incompatibilidade espacial que pode resultar em uma menor taxa de visitação, e consequentemente, produção de frutos. O Pequi (*Caryocar brasiliense*) é uma espécie de planta endêmica e amplamente distribuída no Cerrado brasileiro. É polinizada por dois morcegos nectarívoros (*Anoura geoffroyi* e *Glossophaga soricina*) também com distribuições amplas por toda a região Neotropical. O nosso objetivo é avaliar se as mudanças climáticas podem alterar a distribuição dessas espécies, ocasionando uma incompatibilidade espacial entre plantas e polinizadores. O segundo objetivo é avaliar se os cenários futuros de mudanças climáticas podem comprometer a produtividade de frutos de Pequi. Para avaliarmos o efeito das mudanças climáticas sobre a produtividade e serviço de polinização nessa planta, realizamos a Modelagem de Nicho Ecológico. Nesses modelos usualmente utiliza-se registros de ocorrência de espécies para extrair uma faixa de tolerância ambiental onde as espécies são capazes de sobreviver, e após isso é estima-se áreas adequadas para com base nessa informação. Nós encontramos que as mudanças climáticas não exercem um efeito muito forte na distribuição da planta e de seus polinizadores. Assim, o nosso resultado indica que o serviço de polinização se mantém em cenários futuros de mudanças climáticas. O mesmo acontece quando avaliamos somente a produtividade. Em nossos resultados foi possível observar apenas uma retração de áreas com alta produtividade em cenários de mudanças climáticas. Isso pode indicar que as mudanças climáticas não exercem um efeito muito grande em sua distribuição, e consequentemente, sua produtividade. Portanto, o Pequi pode ser um cultivo rentável de acordo com as projeções climáticas que utilizamos.

Palavras-chave: serviço de polinização, produtividade, economia, Modelagem de Nicho Ecológico.

## ABSTRACT

The production of fruits in agricultural crops is the result of several external factors that contribute to increase productivity over time. For instance, the pollination ecosystem service performs the exchange of gametes among plant individuals, increasing the reproductive rate of these species. However, climate change can alter the distribution of plants and pollinators, leading to a spatial incompatibility that can result in a lower rate of visitation and, consequently, fruit production. “Pequi” (*Caryocar brasiliense*) is a plant species endemic and widely distributed in the Brazilian Cerrado. It is pollinated by two nectarivorous bats (*Anoura geoffroyi* and *Glossophaga soricina*) also with broad distributions throughout the Neotropical region. Our aim is to evaluate if climatic changes can alter the distribution of these species, causing a spatial incompatibility between plants and pollinators. The second objective is to evaluate if future scenarios of climate change can compromise the productivity of “Pequi” fruits. In order to evaluate the effect of climatic changes on productivity and pollination service in this plant, we performed the Ecological Niche Modeling. In these models, species occurrence records are usually used to extract a range of environmental tolerance where species are able to survive, and after that, it is estimated that adequate areas are based on this information. We found that climate change does not have a very strong effect on the distribution of the plant and its pollinators. Thus, our result indicates that pollination service remains in future scenarios of climate change. The same is true when we evaluating only productivity. In our results, it was possible to observe only a retraction of areas with high productivity in scenarios of climate change. This may indicate that climate change does not have a very large effect on its distribution, and consequently its productivity. Therefore, “Pequi” can be a profitable crop according to the climate projections we use.

Keywords: pollination service, productivity, economy, Ecological Niche Modeling.

## APRESENTAÇÃO GERAL

A presente dissertação intitulada “**Investigando o impacto das mudanças climáticas sobre a produtividade e o serviço de polinização em *Caryocar brasiliense***” está dividida em dois capítulos. No primeiro capítulo investigamos o efeito das mudanças climáticas no serviço ecossistêmico de polinização, no qual testamos a hipótese de que as mudanças no clima provocarão uma incompatibilidade espacial entre a distribuição da planta e seus respectivos polinizadores. No segundo capítulo é feita uma abordagem econômica sobre a produção de frutos. Para isso, investigamos a capacidade dos modelos de nicho ecológico em prever a produtividade. Com isso, também estimamos o efeito das mudanças climáticas na produtividade econômica em *Caryocar brasiliense*. Em ambos os capítulos incorporamos o uso de dados de solo ao modelarmos a planta, assumindo a premissa de que a adição de variáveis de solo pode aumentar a precisão dos Modelos de Nicho Ecológico. Os capítulos seguem as normas de formatação das revistas *Biological Conservation* e *Ecological Economics* respectivamente.

## INTRODUÇÃO GERAL

A conservação da biodiversidade tornou-se um tema frequente na comunidade científica (Correa Ayram et al., 2016; Willis and Birks, 2006). Diversos estudos tentam indicar soluções para a conservação, a fim de maximizar a persistência das espécies na natureza (Oliveira et al., 2017; Terrado et al., 2015; Whitehead et al., 2017). No entanto, diversos fatores podem resultar em consequências para a manutenção dessas espécies, tal como a degradação de habitat (Wilson et al., 2016), sobre-exploração (Gallo and Pejchar, 2016), introdução de espécies exóticas (Miller et al., 2016) e mudanças climáticas (Bellard et al., 2012). Identificar esses problemas é uma etapa fundamental para o estabelecimento de ações práticas de conservação. As mudanças climáticas, por exemplo, podem alterar o tamanho da distribuição das espécies, restringindo populações em pequenos fragmentos de áreas adequadas (Martínez-Meyer et al., 2004). Além disso, é possível detectar outros elementos importantes para a conservação de uma espécie. A diminuição na taxa de interações dentro de uma comunidade pode reduzir a sobrevivência de indivíduos (Giannini et al., 2012). Em plantas, o serviço de polinização pode ser considerado um elemento chave para a persistência de espécies, pois a sua ausência pode reduzir a taxa reprodutiva de indivíduos (Zhang et al., 2007). Em espécies de interesse econômico, a associação entre mudanças climáticas e a redução na taxa de polinização pode reduzir o tamanho e a qualidade dos frutos, o que potencialmente pode ocasionar um impacto econômico negativo para a sociedade.

A polinização é um processo ecossistêmico em que indivíduos de outras espécies realizam a troca de gametas entre plantas (Costanza et al., 1997). Em alguns casos, torna-se um serviço ecossistêmico ocasionado pelo benefício desse processo para os seres humanos (Costanza et al., 1997). Para alguns cultivos esse serviço é fundamental para manter a produtividade ao longo do tempo. Apesar disso, essa interação biológica pode ser vulnerável a diversos impactos, o que conseqüentemente pode afetar a produtividade em plantas (Stout, 2014). As mudanças climáticas, por exemplo, tem sido um dos fenômenos mais discutidos nos últimos anos, e o seu impacto na agricultura tem recebido bastante destaque (Nabout et al., 2011; Settele et al., 2016). A velocidade em que o clima está mudando pode dificultar com que as espécies consigam acompanhar áreas adequadas para sua sobrevivência (Jackson and Overpeck, 2000). Conseqüentemente, se espécies de plantas não conseguem acompanhar as mudanças climáticas, é bastante provável que as áreas adequadas diminuam e possivelmente reduz suas respectivas taxas reprodutivas. Além disso, a taxa reprodutiva de uma planta também pode ser afetada pelos polinizadores. Por esse motivo, as mudanças climáticas



também podem afetar a produtividade de plantas por meio de uma perda da sobreposição das áreas adequadas entre plantas e polinizadores (Schweiger et al., 2008).

Em plantas economicamente importantes, a redução da taxa reprodutiva também pode representar uma perda significativa para municípios que utilizam o extrativismo vegetal como fonte de renda (Gribel and Hay, 1993; Nabout et al., 2011). Entender os fatores que afetam a produtividade dessas plantas pode direcionar esforços para esse tipo de produção. Alguns estudos têm investigado como que as condições ambientais podem afetar a produtividade (Nabout et al., 2011). Em alguns casos utilizam-se características biológicas para entender quais fatores influenciam mais na produção de comida. Conhecer a produtividade potencial de alguns cultivos pode dar suporte na tomada de decisões sobre quais espécies cultivar (Zardo and Henriques, 2011). Isso pode gerar um benefício direto para produtores rurais, pois se a escolha for feita por meio de critérios bem estabelecidos é bastante provável que as consequências econômicas da produção agrícola não afetem municípios que necessitam desse tipo de produção para manter sua renda mensal. Além disso, entender como a produtividade potencial está distribuída no espaço pode dar suporte sobre onde cultivar. No entanto, para mapear essas áreas é necessário o uso de algumas ferramentas que auxiliem na delimitação dessas áreas. Um exemplo frequentemente utilizado na literatura é a Modelagem de Nicho Ecológico (em inglês, Ecological Niche Modeling - ENM).

Modelagem de Nicho Ecológico é um procedimento estatístico que associa registros de ocorrência a suas respectivas condições ambientais observadas (Peterson, 2006). A partir disso, estima áreas adequadas por meio da similaridade ambiental entre locais, comparando as condições ambientais de cada unidade de área à faixa de tolerância ambiental conhecida para a espécie alvo (Soberón and Nakamura, 2009). Esses modelos têm sido utilizados em diferentes abordagens de estudos ecológicos (Papeş and Gaubert, 2007; Velazco et al., 2017). Na literatura é possível listar algumas áreas de conhecimento importantes que utilizam essa ferramenta como suporte para tomadas de decisão (Marini et al., 2010). Como um exemplo, o estabelecimento de áreas prioritárias tem utilizado essa abordagem para selecionar áreas que potencialmente apresentarão condições ambientais favoráveis para a persistência de espécies (Pierce et al., 2005). Além disso, também é possível utilizar mapas de adequabilidade ambiental resultantes dos modelos de nicho para propor outras formas de pesquisa, como as áreas de alta produtividade agrícola e silvicultura.

Ao modelar plantas, a maioria dos estudos utilizam somente dados climáticos para prever a adequabilidade ambiental (Diniz-Filho et al., 2009; Marco-Júnior and Siqueira,

2009). No entanto, alguns estudos têm incorporados dados de solo na distribuição de plantas (Velazco et al., 2017). Então, assumindo que dados de solo não muda rapidamente com o tempo, é possível que ao integrar esses dados em modelos de distribuição, potencialmente algumas espécies não mudariam muito a suas distribuições, ficando restringidas no espaço geográfico. Então, ao considerar que o serviço de polinização também seja um fator importante na produtividade de plantas, é possível que a adição de variáveis de solo restrinja a distribuição da planta, o que poderia evidenciar uma incompatibilidade espacial entre as plantas e polinizadores (Morton and Rafferty, 2017; Polce et al., 2014; Schweiger et al., 2008). Além disso, pode-se esperar que o uso de dados de solo pudesse complementar o nicho conhecido da espécie a fim de estimar melhores predições para plantas.

A árvore Pequi (*Caryocar brasiliense* Camb. Caryocaraceae) é uma espécie endêmica e amplamente distribuída no bioma Cerrado. Essa é uma das espécies em que o extrativismo vegetal é bastante forte, principalmente por municípios que extraem o fruto dessa espécie para comercializar em mercados locais (Guedes et al., 2017; Nabout et al., 2011). Além disso, é uma espécie comum na culinária da região central do Brasil. O seu fruto é bastante comercializado devido aos seus diversos usos, desde extração de óleos e amêndoas até a alimentação da polpa do fruto (Araujo, 1995). Apesar da diversidade de possibilidades de uso desse fruto, para manter uma alta produtividade é necessário que a espécie seja polinizada, pois a taxa de autopolinização é baixa, necessitando de outras espécies para manter a produção de frutos. Seus polinizadores são duas espécies de morcegos nectarívoros (*Glossophaga soricina* e *Anoura geoffroyi*) (Gribel and Hay, 1993). Além disso, sua importância econômica atraiu o interesse de diversos pesquisadores na área de ecologia. Por esse motivo, existem vários estudos estimando parâmetros populacionais e distribuição potencial. Em particular, estudos investigando a distribuição potencial dessa espécie utilizam somente variáveis climáticas como preditores ambientais.

Nosso objetivo está dividido em dois capítulos em formato de artigos científicos. No primeiro capítulo estamos interessados em estudar o efeito das mudanças climáticas na sobreposição de áreas adequadas entre planta e polinizadores. Para isso, testaremos a hipótese de que em cenários de mudanças climáticas haverá uma perda de sobreposição entre áreas altamente adequadas. Além disso, testamos o pressuposto de que dados de solo podem restringir a distribuição da planta, fazendo com que a sua distribuição no espaço varie menos. No segundo capítulo estamos interessados em investigar a capacidade dos modelos de nicho ecológico em prever áreas com alta produtividade para o Pequi. Em ambos os capítulos

conduzimos os procedimentos de modelagem utilizando dois conjuntos de dados ambientais. Um deles é somente climático e no outro incorporamos os dados de solo.

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1 **The climate change impact on the mismatch between plant and pollinators distributions**  
2 **in the Brazilian Cerrado**

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13 **1. Introduction**

14           Pollination is an interspecific interaction with strong ecosystem effects on the stability  
15 of natural and human-dominated systems (Hadley and Betts, 2012). The lack of pollinators  
16 can constrain the development of key plant populations in the ecosystem (Potts et al., 2010)  
17 affecting, for example, the succession process (Devoto et al., 2012; Forup et al., 2008). On the  
18 other hand, the pollination service affects the productivity of agroecosystems (Kennedy et al.,  
19 2013; Marco and Coelho, 2004; Potts et al., 2007), with well-documented large economic  
20 effects (Gallai et al., 2009). Therefore, it has the potential to encourage the public interest in  
21 biodiversity conservation considering the economic advantages for society (Morandin and  
22 Winston, 2006), helping to build public policies for conservation (Chan et al., 2006). This  
23 approach was developed mainly in the discussions involving economically important crops  
24 (Cristina Giannini et al., 2013; Marco and Coelho, 2004). However, pollination is a  
25 vulnerable interaction, since the biological elements involved are sensitive to different  
26 impacts (Montero-Castaño and Vilà, 2012; Settele et al., 2016; Stout, 2014). Climate change,  
27 for example, can alter the distribution of climatically suitable areas for the survival of both the  
28 plant species and its pollinators (Giannini et al., 2012; Pyke et al., 2016), and for the  
29 occurrence of pollination, which in turn can have consequences in fruit productivity (Elias et  
30 al., 2017; Polce et al., 2014).

31           The speed at which climate change occurs reduces the chance of species migrating  
32 under suitable conditions for their survival (Jackson and Overpeck, 2000). This can reduce the  
33 species distributional size, which may cause temporal and spatial mismatch between plants  
34 and pollinators, whereas, as a consequence, affects the fitness and population dynamics in  
35 plants (Hegland et al., 2009; Schweiger et al., 2008). Furthermore, the soil, as a static  
36 component, restricts plants' dispersion (Schweiger et al., 2008; Velazco et al., 2017), but do  
37 not affect directly the distribution of pollinators. A consequence of this would be the loss of  
38 plant and pollinators overlapping distribution under effect of climate change, decreasing the  
39 probability of pollination (Silva et al., 2014). The study of how climate change affects  
40 interactions, however, involves several methodological issues that have only recently begun to  
41 be circumvented (Morton and Rafferty, 2017).

42           The most common technique for the study the geographical distribution of species is  
43 the Ecological Niche Modeling (ENM). These models are simplifications of the patterns of  
44 species' distribution in the nature based on their environmental requirements (Peterson and  
45 Soberón, 2012). Their algorithms use observed environmental data at the occurrence records



46 to predict suitable areas by means of environmental similarity between sites, considering the  
47 assumption that the species' niche does not change. Usually, climate has been used as  
48 macroscale environmental data to these analyses (Ackerly et al., 2010; Pearson and Dawson,  
49 2003). This allows the study of species distribution in different scenarios, both in space and  
50 time. The ENMs are statistical procedures based mainly in theory of ecological niche, in  
51 which individuals of the same species are associated with a set of environmental conditions  
52 and resources that define and limit their distribution (Soberón, 2007).

53 The "Pequi" tree (*Caryocar brasiliense* Camb, Caryocaraceae) is an endemic species  
54 broadly distributed in the Cerrado ecoregion. The fruit of this species is used for direct  
55 consumption and for the extraction of other materials (Araujo, 1995), which makes it very  
56 important for regional economy. The reproductive system in this plant consists of  
57 hermaphrodite and auto compatible flowers, but the fitness in *C. brasiliense* depends in large  
58 part on the cross-pollination (Gribel and Hay, 1993). This helps to reduce the number of  
59 aborted seeds, which is strongly associated with high rates of self-pollination (Collevatti et al.,  
60 2009). The pollination service in this species is provided by two species of nectarivorous bats:  
61 *Glossophaga soricina* and *Anoura geoffroyi* (Gribel and Hay, 1993). These pollinators present  
62 broad geographical distribution and are considered not threatened (LC - low concerns)  
63 according to the IUCN criteria (IUCN, 2018). Nevertheless, some studies already predicted a  
64 change in the geographical distribution of *C. brasiliense* in climate change scenarios  
65 (Collevatti et al., 2011a; Nabout et al., 2011).

66 Our goal is to evaluate the climate change consequences on geographic range of plant-  
67 pollinator interactions. Here, we assume that soil data constraint plants distribution in climate  
68 change scenarios. We hypothesize that there will be a loss of overlap between present and  
69 future spatial distribution of plants and its pollinators in climate change scenarios. Otherwise,  
70 this analysis may provide a more realistic evaluation of climate change effects on a food  
71 resource ("Pequi") with high importance in a portion of Brazilian national market.

## 72 **2. Methods**

### 73 *2.1. Species Distribution Database and Data Treatment*

74 To describe completely the environmental niche of these species (Raes, 2012), we  
75 used all the occurrence records from a systematic review of the literature available on Web of  
76 Science and Scopus, with the search codes: ("*Caryocar brasiliense*" OR "*C. brasiliense*" OR  
77 "*Caryocar brasiliense* Cambess" OR "pequi" OR "piqui" OR "pequizeiro" OR "piquizeiro");

78 ("Glossophaga soricina" OR "G. soricina" OR "bat list"); ("Anoura geoffroyi" OR "A.  
79 geoffroyi" OR "bat list"). Moreover, we completed the occurrence records using two online  
80 databases: (1) GBIF (<https://www.gbif.org/>); and (2) SpeciesLink (<http://splink.cria.org.br/>).  
81 All data collected, especially those from online database, were evaluated considering various  
82 quality criteria (De Giovanni et al., 2012). We considered possible georeferencing errors  
83 (duplicated occurrence records, latitude and longitude exchange, and occurrence records  
84 outside the Neotropical region), and taxonomic identification accuracy (mainly for species of  
85 the "Pequi" tree). We removed all occurrences considered to be doubtful in relation to these  
86 criteria, which improved the quality and reliability of the data to be used in the models.

## 87 2.2. Environmental variables

88 We separated the environmental data in two sets to perform the ENMs: (1) climate-  
89 only; (2) climate and edaphic variables together. Climate data were obtained from WorldClim  
90 database (<http://www.worldclim.org/>), and the current climate scenario was based on  
91 interpolations of meteorological data collected in the period from 1960 to 1990. For the future  
92 scenario, we used 17 climate models for the year of 2070 from the Fifth Assessment Report  
93 (AR5) of the Intergovernmental Panel on Climate Change (IPCC). These data are climate  
94 projections calibrated using as baseline current climate, and we used only one (Representative  
95 Concentration Pathways 8.5) for the following climate models: ACCESS1-0, BCC-CSM1-1,  
96 CCSM4, CESM1-CAM5-1-FV2, CNRM-CM5, GFDL-CM3, GFDL-ESM2G, GISS-E2-R,  
97 HadGEM2-AO, HadGEM2-CC, HadGEM2-ES, INMCM4, IPSL-CM5A-LR, MIROC-ESM-  
98 CHEM, MIROC-ESM, MIROC5, MPI-ESM-LR, MRI-CGCM3, NorESM1-M. Edaphic data  
99 were obtained from the SoilGrids database available from International Soil Reference and  
100 Information Centre (ISRIC, <https://www.isric.online/>), which describe physical and chemical  
101 properties of the soil (Hengl et al., 2014). From this dataset, we selected eight variables that  
102 we considered important for Cerrado plants: (1) Depth to bedrock (R horizon) up to 200 cm;  
103 (2) Probability of occurrence of R horizon; (3) Cation exchange capacity of soil in cm olc/kg;  
104 (4) Clay content (0-2 micrometer) mass fraction in %; (5) Soil pH x 10 in H<sub>2</sub>O; (6) Soil pH x  
105 10 in KCl; (7) Silt content (2-50 micro meter) mass fraction in %; and (8) Sand content (50-  
106 2000 micro meter) mass fraction in %.

107 Both for the choice of climate and edaphic variables, and the general procedures of  
108 modeling we followed Velazco et al. (2017), who made a very complete account of the use of  
109 those variables for model plant distributions in South America. Original climate and edaphic

110 data has a spatial resolution of 30 arc-seconds ( $\approx 1 \text{ km}^2$  cell size). We upscaled them to 5 arc-  
111 minutes ( $\approx 10 \text{ km}$  cell size) for all analysis using the average value of the higher resolution  
112 cells into lower resolution cells. In addition, in order to reduce multicollinearity among the  
113 environmental variables, we used a Principal Component Analysis (PCA) in the two datasets  
114 (climate-only, climate and edaphic variables together) (De Marco and Nóbrega, 2018). PCA is  
115 a multivariate technique that reduces the dimensionality of a large dataset, initially correlated.  
116 A recent work has evaluated the use of PCA transformed variables into ENM and found an  
117 increase of accuracy for different algorithms. This also reduces the complexity of the  
118 algorithms of ENMs, which avoids the production of unrealistic predictions. We selected the  
119 axes that represent 95% or more of the total variation of the original environmental matrix.

### 120 *2.3. Modeling procedures and evaluation of models*

121 The ENMs are statistical procedures that relate the environmental variables with the  
122 occurrence points to predict suitable areas for the species. Here, we used five algorithms of  
123 ENMs with the purpose of measuring the uncertainty associated with these procedures,  
124 producing better consensus maps among them: (1) Maximum Entropy (MXS) reduce  
125 complexity and predict better models in specific situations (Phillips et al., 2017, 2004); (2)  
126 Random Forest (RDF) adjust the distribution models based on decision trees (Prasad et al.,  
127 2006); (3) Support Vector Machine (SVM) creates a separation line between the occurrence  
128 records and a set of absences (Guo et al., 2005); (4) Maximum Likelihood (MLK) estimates a  
129 probability distribution of occurrence based on the environmental conditions of the presence  
130 records (Royle et al., 2012); e (5) Gaussian (GAU) predicts the probability of occurrence  
131 based on adjustments made by Bayesian inference (Golding and Purse, 2016). Typically,  
132 these ENMs need real absences to be adjusted, with exception of the MXS, which adjust the  
133 models by the difference between the occurrence records and a sample of all background.  
134 Nevertheless, real absences are rare; as an alternative we used pseudo-absence to adjust these  
135 algorithms. For this purpose, we created a bioclimatic envelope as a pseudo-absence selection  
136 method and only unsuitable areas for the species are considered. This method is advised  
137 because it reduces the possibility of a real presence not recorded in our database to be  
138 included as a pseudo-absence in the analysis (Muscarella et al., 2014).

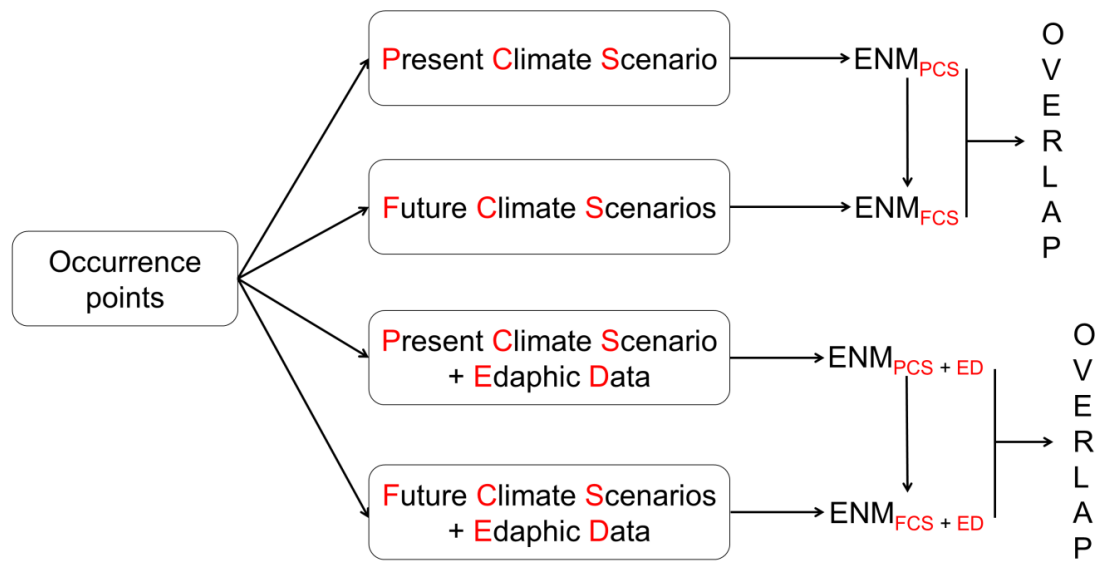
139 The ENMs evaluations were performed using geographic partition in a checkerboard  
140 format. This partition method subdivides the study extent equally and selects, alternately,  
141 training (to perform the model) and testing (to evaluate the models) occurrences (Muscarella

142 et al., 2014). The algorithms performances were estimated by the True Skill Statistics (TSS),  
143 using True Positive Rate (TPR) and True Negative Rate (TNR) metrics. This procedure has  
144 already been compared with other metrics and; therefore, considered as a simple and intuitive  
145 measure to evaluate the performance of the ENMs (Allouche et al., 2006).

146 In order to test the general hypothesis of this work, we produced consensus maps with  
147 only the grid cells that the models agreed to. However, there are many different techniques to  
148 produce consensus maps (Marmion et al., 2009). For the purpose of this study, we used the  
149 sum of models that present TSS values above average. In addition, we made presence and  
150 absence maps using a threshold at which the sum of the sensitivity and specificity is highest  
151 (Liu et al., 2011). We applied this consensus procedure initially to the models from different  
152 algorithms at the present environment. In the projection of the future climates the problem is  
153 more complex once we have the variation among the algorithms and among different climate  
154 scenarios. To allow an evaluation of the uncertainty derived from the climate scenarios, we  
155 performed the above method for all algorithms within each climate scenario separately. Thus  
156 we have 19 different final future distributions for each species and could evaluate our  
157 hypothesis for each of these results. All modeling procedures was done using  
158 ENM\_TheMetaLand package in software R  
159 ([https://github.com/andrefaa/ENM\\_TheMetaLand](https://github.com/andrefaa/ENM_TheMetaLand)).

#### 160 *2.4. Experimental Design and Analytical Procedures*

161 The experimental design of this study consists on the comparison between the  
162 potential pollination service estimated by the ENMs, considering the current and future  
163 climate scenarios. We assume that highly suitable areas for the bat species are a surrogate of  
164 its presence (and possibly abundance) (VanDerWal et al., 2009) and, thus, may be converted  
165 into a quantitative estimate of the availability of the pollination service. Our basic assumption  
166 is that the plant species will be constrained by both climate and edaphic variables, while the  
167 bats are only affected by climate. Thus, we expect that the plant will experience less  
168 distributional change than its pollinators, favoring a disconnection of its distributions. To  
169 evaluate these possibilities and the possible effect of edaphic variables as a constraint to plant  
170 distribution in future climates, we used both (1) climate and (2) climate + edaphic variables in  
171 plant models (Fig. 1). Otherwise, we modeled bat pollinator distributions based only on  
172 climate variables.

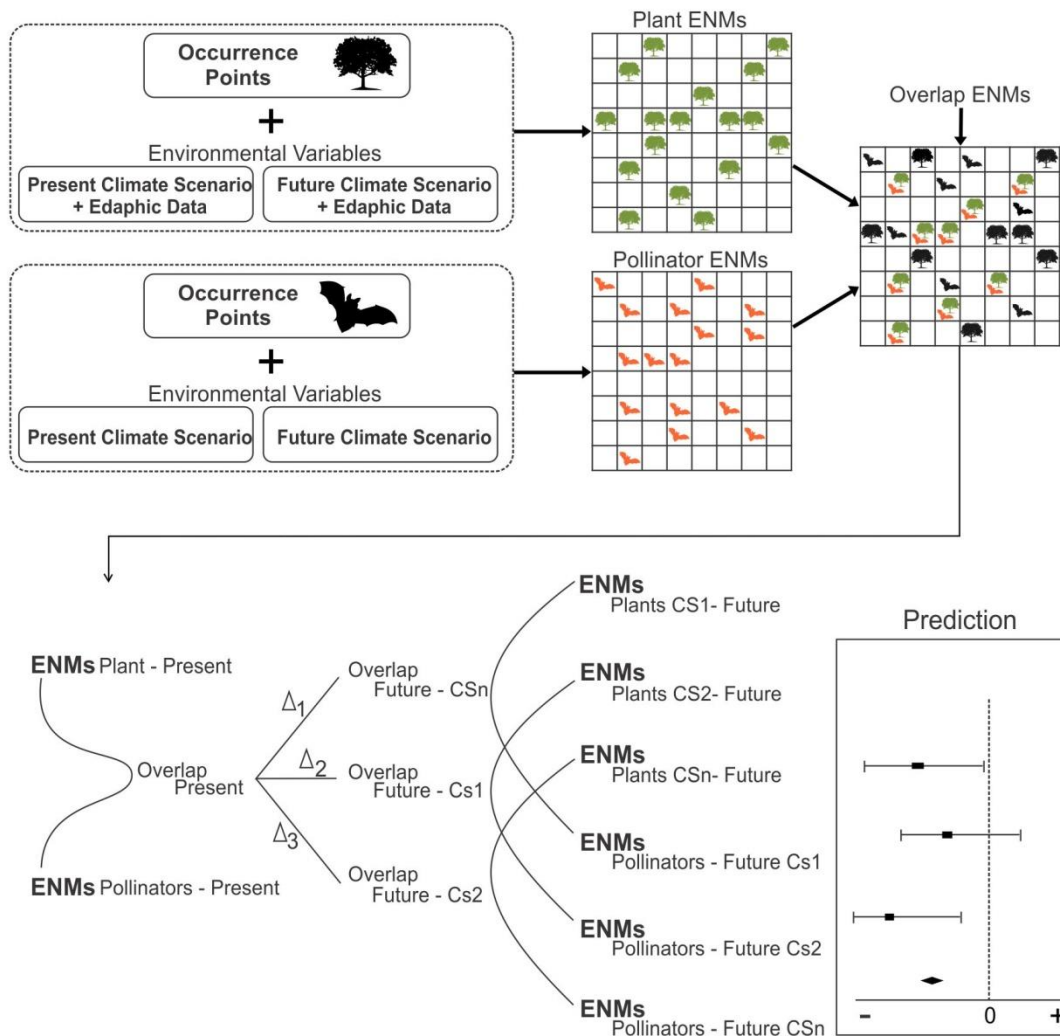


173

174 Fig. 1- Experimental design to test the assumption of edaphic effects on plant distribution  
 175 under different climate change scenarios. All models were done using occurrence points and  
 176 fitted in present climate scenario (PCS) and projected to future climate scenarios (FCS). The  
 177 inclusion of edaphic data (ED) in models is expected to produce smaller distributions under  
 178 future climate. The prediction is that the overlap between future and present distribution will  
 179 be smaller for models that include ED.

180 Our response variable was the difference in the proportion of overlapping area of each  
 181 bat pollinator and the plant distribution with edaphic data for the present and each future  
 182 climate scenario. This was done for each possible combination of interaction (plant and  
 183 pollinator 1; plant and pollinator 2; and plant with both pollinators). Present model comprise  
 184 only one value of overlap area, while future models include several values derived from  
 185 different climate scenarios. Thereby, we used an inference based on average and confidence  
 186 interval estimates to evaluate the hypothesis that the overlap among the distributions decrease  
 187 for future scenarios. Here, we use the confidence interval estimates to evaluate the effect size  
 188 that was represented by the forest plot analysis (Michael Borenstein, Larry V. Hedges, Julian  
 189 P. T. Higgins, 2009). We expected that the difference between the present and future to be less  
 190 than zero.

## Modeling Pollination Service by ENMs



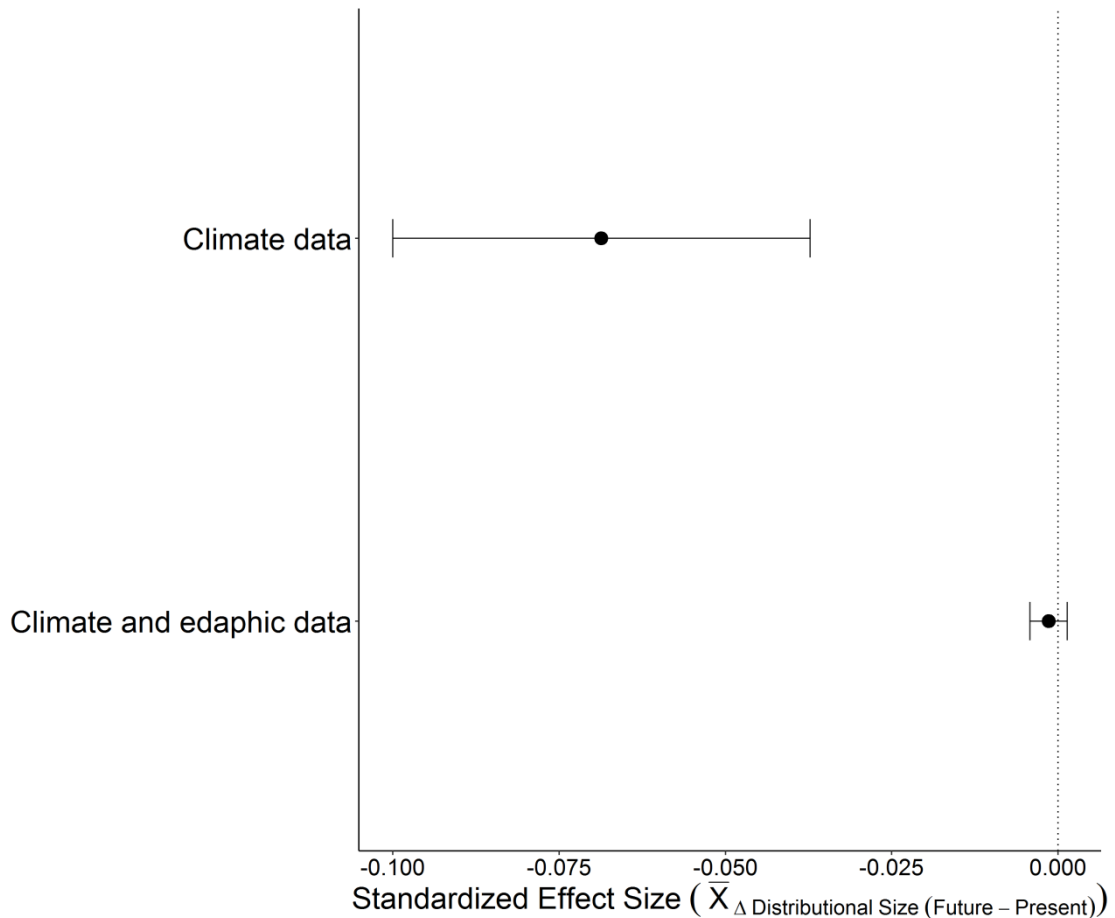
191

192 Fig. 2- Experimental design to test the hypothesis that the overlap among the distributions  
 193 decrease for future scenarios. The ENM procedures for plant use occurrence points and the  
 194 environmental variables (climate and climate-edaphic data), while for bat pollinators use only  
 195 climate data. After that, the binary map is generated for all species present in the interaction.  
 196 Using these maps, we measured the overlap between the plant and bat pollinators for all  
 197 combinations of interaction (for present and future climate scenarios). Thereby, we obtained  
 198 the difference value between present and future climate scenarios ( $\text{Overlap}_{\text{Future}} -$   
 199  $\text{Overlap}_{\text{Present}}$ ), in which the prediction result is that these difference are smaller than zero.

### 200 3. Results

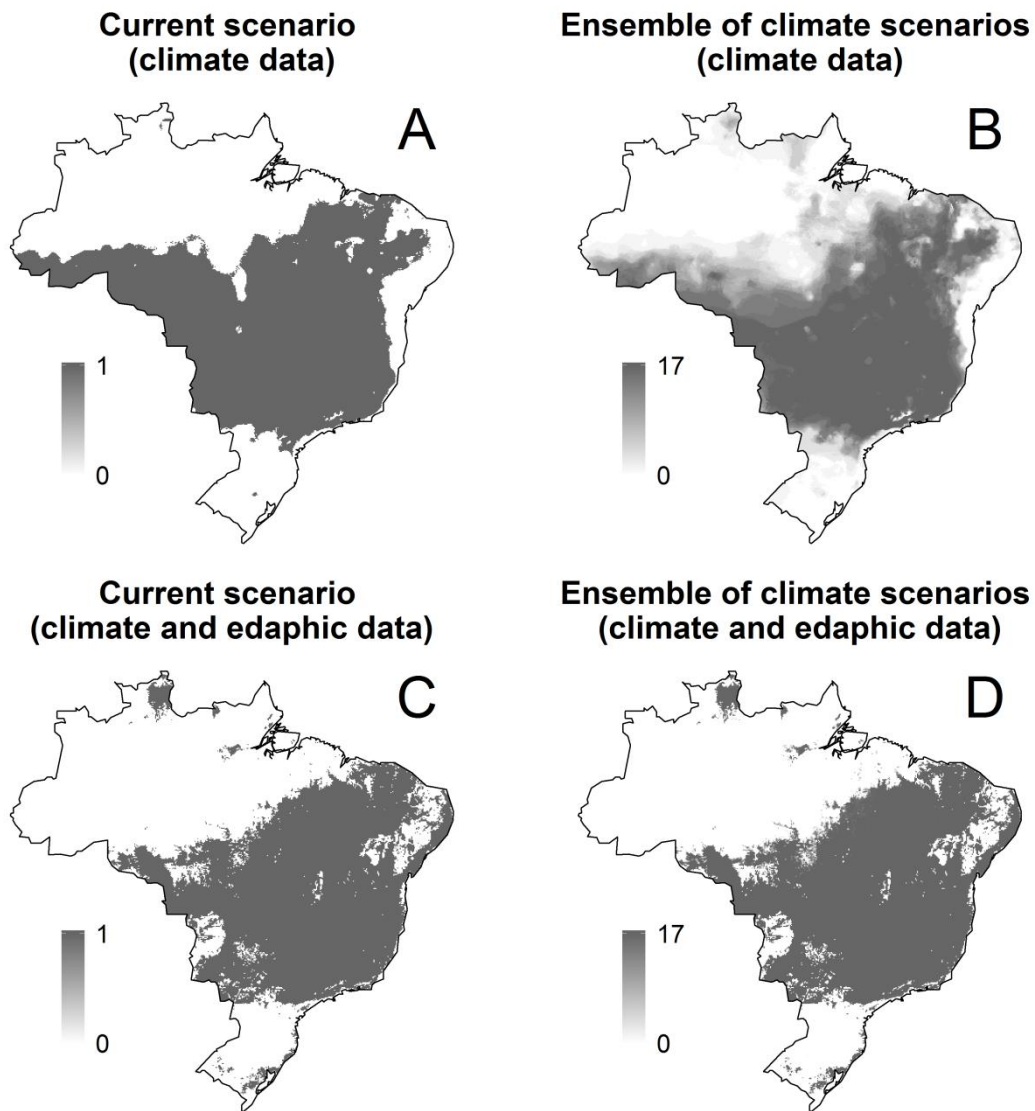
201 We found a small difference in the plants' distributional size when using edaphic  
 202 data in the ENM. This result supports our assumption that edaphic data constraint plants'

203 distribution. Furthermore, in future scenarios, the spatial distribution of “Pequi” remains  
204 significantly the same. On the other hand, the models using climate-only data evidenced more  
205 uncertainty on geographic distributional size (Fig. 3). This is because there are less accurate  
206 cell predictions in the North of the maps. Nevertheless, the geographic distributional sizes do  
207 not change with the combination of edaphic variables and climate data (Fig. 4), but is reduced  
208 in future if only climate data is used.



209

210 Fig. 3- Forest plot testing the effect of the addition of edaphic data in the plant’s modeling  
211 procedures. This analysis was performed using climate-only and with edaphic data added to  
212 climate database. Each model was projected for 17 future models. From this, we counted the  
213 number of cells for the present and each future model respecting Brazilian borders and  
214 calculated the difference between each future model and the present. This result was  
215 standardized using the current model of plant distribution.

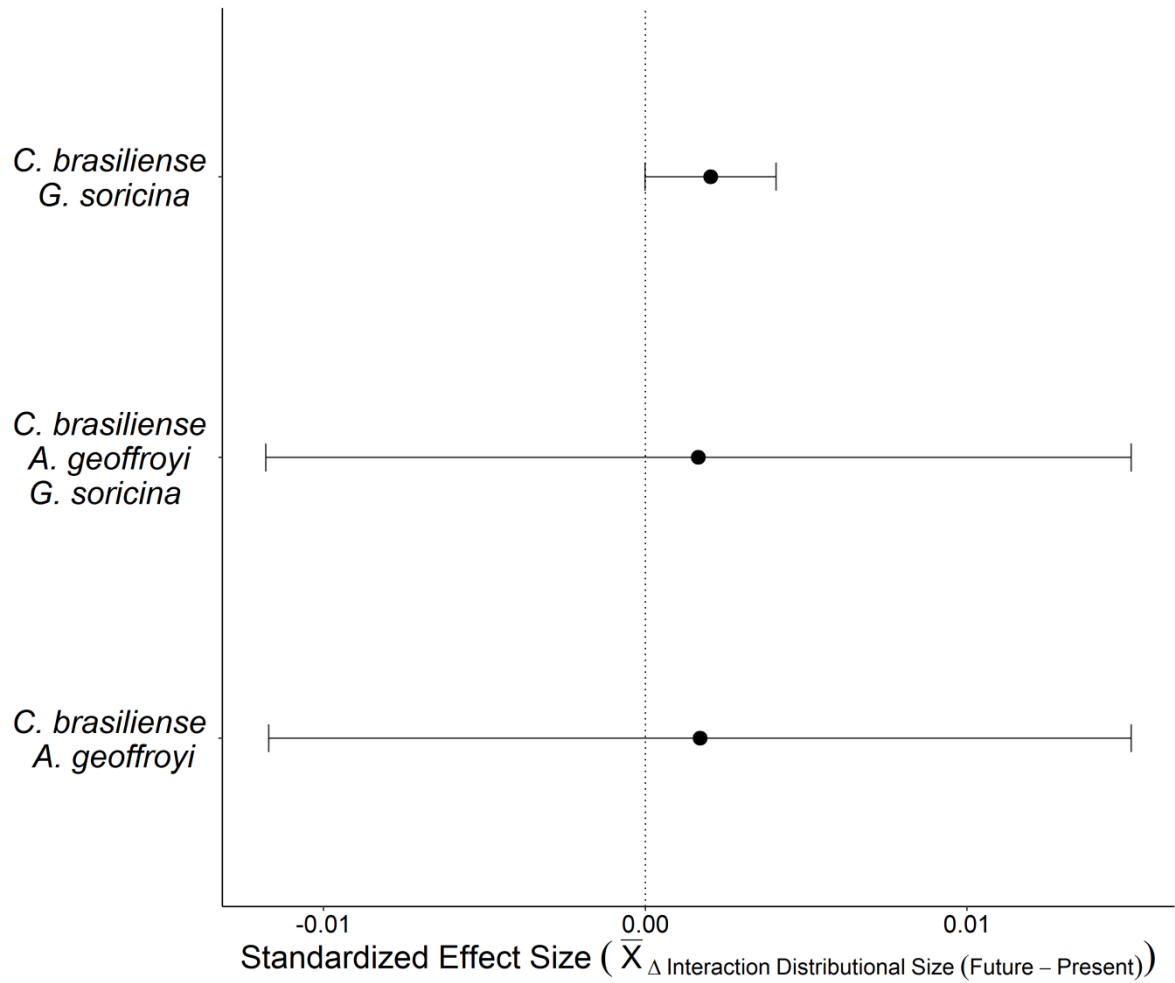


216

217 Fig. 4- Spatial distribution using different datasets for the current and future models in the  
 218 plant modeling.

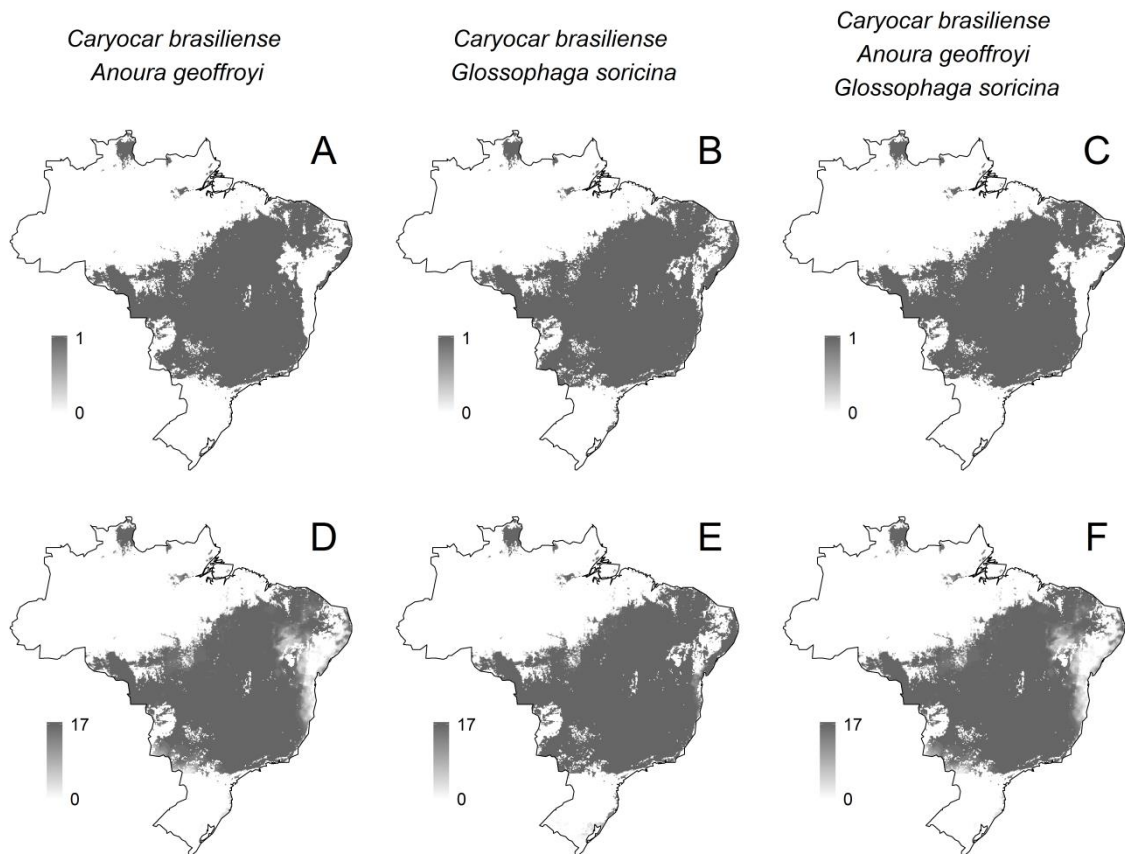
219 We also found that the overlapped distribution of bats and “Pequi”, as well as their  
 220 distributional range, do not change in future scenarios, indicating that the pollination service  
 221 for “Pequi” trees will be maintained (Fig. 5). We observed that *A. geoffroyi* restricts the  
 222 overlap with “Pequi” distribution. However, the interaction has lower spatial uncertain is *C.*  
 223 *brasiliense* and *G. soricina*, maintaining the overall level of overlap with “Pequi” distribution.





224

225 Fig. 5- Forest plot testing if interaction distributional sizes reduce in future scenarios. This  
 226 analysis was performed using climate and edaphic data for plant modeling and climate-only  
 227 for pollinators. In this analysis, the geographical distributions of the species were overlapped  
 228 for present and each future model. From this, the number of cells was counted for the present  
 229 and each future model respecting the Brazilian borders, and the difference between each  
 230 future model and the present. This result was standardized using the current model of plant  
 231 distribution.



232

233 Fig. 6- Overlap between the distributional sizes of all species in three combinations: *C.*  
 234 *brasiliense* and *A. geoffroyi*, present (A) and future (D); *C. brasiliense* and *G. soricina*,  
 235 present (B) and future (E); and *C. brasiliense*, *A. geoffroyi* and *G. soricina*, present (C) and  
 236 future (F). For each interaction there are 17 future combination models. All procedures were  
 237 done using edaphic data in the plant modeling.

## 238 DISCUSSION

239 Our results suggest that the use of the edaphic properties may restrict the distribution  
 240 of plants, indicating that our assumption is true for *C. brasiliense*. Specifically, we observed a  
 241 small difference between the current and future scenarios when using the association of  
 242 climatic and edaphic data as predictor variables. This may indicate that the use of edaphic  
 243 data produces better predictions about plant distributions. In the scientific literature most  
 244 studies involving ENMs for plants (including *C. brasiliense*) used only climate data as  
 245 environmental predictors (Gaikwad et al., 2011; Trethowan et al., 2011) (Collevatti et al.,  
 246 2011b, 2011c; Diniz-Filho et al., 2009; Nabout et al., 2011). In these studies, it is possible to  
 247 observe that the climatic conditions strongly alter the distribution of “Pequi”, which shifts

248 from the central to the southeast region of Brazil. Nevertheless, some studies have  
249 incorporated the edaphic data for modeling plants (Velazco et al., 2017), but did not compare  
250 the effect of different environmental predictors in their distributional size. Altogether, our  
251 models contrast the results found in the literature.

252 For *C. brasiliense*, it is possible that its large known distributional size is associated  
253 with large climate/edaphic tolerances, which explains both the large predicted current and  
254 future distributions. However, we cannot discard the hypothesis that not only environmental  
255 conditions but intrinsic demographic characteristics and external landscape structures affect  
256 its success in nature. For instance, interspecies and intraspecies competition, and the lack of  
257 pollinators may reduce its development. Otherwise, habitat loss may affect all those  
258 interactions besides a direct effect on its persistence. Despite this, our models indicate that the  
259 addition of edaphic variables for modeling plants with larger distributions, such as *C.*  
260 *brasiliense*, will possibly produce similar distribution under present and future climates  
261 scenarios. On the other hand, for species with restricted distribution, additional constrain due  
262 to soil will produce even smaller distributions under future climate scenarios. Thereby, our  
263 results comparing the different environmental predictors in ENM suggest that the use of  
264 edaphic data may produce more accurate predictions for the plants, including better future  
265 projections (Velazco et al., 2017). Consequently, these variables may contribute with the  
266 establishment of priority areas for conservation due better descriptions of species' niche  
267 (Velazco et al., 2019).

268 The analyses of the distributional size of the interaction between plant and  
269 pollinators under climate change scenarios suggest that it will not be affected in the future. As  
270 discussed earlier, one possible reason for this is the “Pequi” and its pollinators' distributional  
271 sizes, both with wide distribution. This condition allows us to say that, regardless of the areas  
272 where our ENMs indicate as suitable for this plant, its pollinators will potentially share the  
273 same area, because all species are broadly distributed. As a consequence, if the pollination  
274 service is the main condition for a high fitness, our results support the idea that *C. brasiliense*  
275 maintain its reproductive rates in future scenarios. This result is on the opposite direction of  
276 many ecological interactions models for plants and their pollinators under climate change  
277 (Jackson and Overpeck, 2000). We strongly suggest that two important points need to be  
278 observed in those studies and many need to be reviewed to a better evaluation of the problem.  
279 First, is necessary to make a more careful prediction about the importance of future climate  
280 change to disconnect the distributions of plants and its pollinators decrease with the increase

281 of the range-sizes of the plant and its pollinators. Secondly, that plant models lacking edaphic  
282 information may provide an incomplete understanding of plant distributional patterns and,  
283 according to our results, may predict higher differences than expected with better or more  
284 complete ENM models.

285

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471

1 **Using Ecological Niche Models to predict the potential production on “Pequi” fruits in**  
2 **Brazil**

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13

## 14    **1. Introduction**

15           Worldwide population growth put agriculture productivity as a major human concern  
16 (Carvalho, 2006; Ruttan, 2015). Both present-day restrictions and ongoing climate change  
17 jeopardize food production because it can alter the spatial distribution of fruitful species and  
18 their pollinators, for example (Olesen and Bindi, 2002). In plants, the environmental  
19 characteristics such as climate conditions, nutrients availability, pH and other soil properties  
20 are the major determinant of their productivity (Poorter and Nagel, 2000). Furthermore, these  
21 information are often used in agrometeorological models to identify suitable areas for  
22 agriculture (Hoogenboon, 2000), which inform farmers to better chose their investments. On  
23 the other hand, spatial models that indicate optimal conditions for crop production may help  
24 government agencies to optimize resource investment on agriculture (Matthews et al., 2008).  
25 Similarly, knowing the future distribution of suitable areas for crop production may subsidize  
26 the creation of strategic plans for the food production (Matthews et al., 2013).

27           Despite a lot of interest are currently on the most valuable agriculture commodities,  
28 such as sugar cane, soybean and corn (Martinelli et al., 2010), many less economically  
29 relevant tropical plants may have an important role in future economy. Pequi, tropical palms,  
30 Cupuaçu and Açaí, are important fruits of the Netropical realm which are majorly consumed  
31 in local markets but with potential to acquire worldwide relevance. Unfortunately, the  
32 knowledge on their response to climate and soil is still insipient. To solve the lack of  
33 information about the environmental conditions necessary for high plant productivity, some  
34 studies have been using the ecological niche theory to map optimum areas for food production  
35 (D'Alpoim Guedes and Butler, 2014). According to this theory, ecological niche is a set of  
36 environmental conditions necessary for a species to survive in a given location. Niche theory  
37 assumes that the environmental conditions required by species do not change through time  
38 (Wiens and Graham, 2005). This allows the projection of the current environmental condition  
39 to future scenarios, and the identification of possible suitable areas (Blanchard et al., 2015).  
40 There are other applications of this theory, such as to determine suitable areas for species and  
41 investigate their niche width (Morueta-Holme et al., 2010; Mykrä and Heino, 2017).  
42 Particularly, some studies are using suitability maps to estimate other environmental  
43 characteristics, as well as abundance of species and its probability of detection in nature  
44 (Mendes and De Marco, 2018).

45           The ENM is a statistic advanced tool used in ecological studies that can predict  
46 suitable areas based on a range of environmental tolerance of focal organisms (Guisan and  
47 Thuiller, 2005). This method associate occurrence records on geographic space to  
48 environmental conditions in which the species live. This association is a proxy for the  
49 species' fundamental niche, and is used by the model to estimate the suitability of new areas  
50 of occurrence (Kearney and Porter, 2009). Most ENMs are performed using only climate data,  
51 but some studies have investigated plant distribution adding edaphic data to the environmental  
52 dataset (Velazco et al., 2017; Walthert and Meier, 2017). (Especially) For plants, soil data  
53 may improve the fit of the models due to their biological requirements. Therefore, the  
54 performance of these models is dependent of the combination of biologically important  
55 variables. Because ENMs indicate best suitable areas for a certain species, and we can  
56 securely assume that this is reflected in their reproductive success (talvez citação aqui), they  
57 can be very useful to predict the productivity of economically important crops (sei lá, frutas,  
58 plantas, etc).

59           The “Pequi” tree (*Caryocar brasiliense* Camb, Caryocaraceae) has a strong economic  
60 impact in many municipalities due to its several uses by society (Guedes et al., 2017). For  
61 instance, the “Pequi” fruit may be used directly in the culinary and in the extraction of  
62 secondary products, such as oils. It is a broadly distributed species in Brazilian territory, and  
63 is pollinated by two nectarivorous bats (*Glossophaga soricina* and *Anoura geoffroyi*) (Gribel  
64 and Hay, 1993), also of wide distribution. Based on this, our aim is to test the prediction  
65 power of climate and edaphic variables on current fruit productivity of “Pequi” in the  
66 Brazilian Cerrado. Specifically, we hypothesize that the addition of edaphic variables to a  
67 climate dataset will improve model predictions. We performed Ecological Niche Models  
68 using climate with and without edaphic variables to explain the spatial variation in “Pequi”  
69 fruit production. Considering the economic importance of this plant, we also evaluated how  
70 climate change could affect both the mean productivity and its spatial distribution in Cerrado.

71

## 72 **2. Material and methods**

### 73 *2.1. “Pequi” distributional database*

74           We developed our database through a systematic review seeking occurrence records  
75 of “Pequi” on the Web of Science and Scopus platforms using the search code (“*Caryocar*  
76 *brasiliense*” OR “*C. brasiliense*” OR “pequi” OR “piquei” OR “pequizeiro” OR “piqueizeiro”).

77 We complemented the data using the GBIF (<https://www.gbif.org/>) and SpeciesLink  
78 (<http://splink.cria.org.br/>) distribution databases. To reduce data uncertainties, we applied four  
79 quality criteria: latitude and longitude exchange, duplicated records, records outside the  
80 Neotropical realm, and occurrence records with possible problems of taxonomic  
81 identification. Occurrences were corrected when possible and eliminated otherwise.

## 82 *2.2. Productivity data*

83 We obtained the “Pequi” fruit production data from the Brazilian Institute of  
84 Geography and Statistics (IBGE; [www.ibge.gov.br](http://www.ibge.gov.br)) using the Automatic System Recovery  
85 (SIDRA) in the category “Production of Vegetable Extraction and Forestry”. Altogether, we  
86 collected the amount of fruits produced in 242 Brazilian Cerrado’s municipalities in 2017.  
87 These data contained the amount of fruit produced (in ton) for each municipality, and we  
88 converted it to kilograms. In addition, we calculated the area for all municipalities producing  
89 “Pequi” and correlated the fruit production and area.

## 90 *2.3. Environmental variables*

91 Our environmental database contains two variable types: climate and edaphic. We  
92 obtained 19 bioclimatic variables from WorldClim database (<http://www.worldclim.org/>) for  
93 present and future scenarios in a Representative Concentration Pathways 8.5 projected for  
94 2070. We also used 17 future climate models from the Fifth Assessment Report (AR5) of the  
95 Intergovernmental Panel on Climate Change (IPCC). We obtained the edaphic data from  
96 International Soil Reference and Information Centre (ISRIC, <https://www.isric.online/>)  
97 (Hengl et al., 2014)(Hengl et al., 2014) and selected eight variables that we considered  
98 important for the Cerrado region: (1) Depth to bedrock (R horizon) up to 200 cm; (2)  
99 Probability of occurrence of R horizon; (3) Cation exchange capacity of soil in cmolc/kg; (4)  
100 Clay content (0-2 micrometer) mass fraction in %; (5) Soil pH x 10 in H<sub>2</sub>O; (6) Soil pH x 10  
101 in KCl; (7) Silt content (2-50 micro meter) mass fraction in %; and (8) Sand content (50-2000  
102 micro meter) mass fraction in %. We obtained the edaphic data in a spatial resolution of 250  
103 meters and upscaled them to 5 arc-minutes ( $\approx 10$  km cell size), in a similar resolution to the  
104 climate data. For this, we used the average value of the cells in higher resolution. In general,  
105 we separated these variables in two datasets: (1) climate-only (hereafter climate data); (2)  
106 climate and edaphic together (hereafter environmental data). For each dataset, we performed a  
107 Principal Component Analysis (PCA) to reduce the multicollinearity (De Marco and Nóbrega,

108 2018), decreasing the complexity of the ENM algorithms. Then, we select the axes in which  
109 the sum represents at least 95% of the total variation of the data.

#### 110 2.4. Ecological Niche Modeling (ENMs)

111 The ENM is a statistical procedure used in several types of studies, as well as  
112 potential distributions (Gallien et al., 2010), and how species respond to climate change  
113 (Walther et al., 2002). Here, we proposed the use of this method to estimate the potential fruit  
114 production of “Pequi” through of the estimated model between the productivity and suitability  
115 for each municipalities of the Brazil. As most of current and future distribution of this species  
116 are in the Brazilian Cerrado (see Chapter 1), this specially represent this biome that is  
117 considered the origin of the distribution of this species. The ENM procedures relate the  
118 species’ occurrence records to environmental variables, in order to estimate the environmental  
119 suitability. We chose five modeling algorithms to perform this analysis: (1) Maximum  
120 Entropy (MXS) (Phillips et al., 2017, 2004); (2) Random Forest (RDF) (Prasad et al., 2006);  
121 (3) Support Vector Machine (SVM) (Guo et al., 2005); (4) Maximum Likelihood (MLK)  
122 (Royle et al., 2012); e (5) Gaussian (GAU) (Golding and Purse, 2016). Some of these  
123 algorithms require absence data to fit the models. In those cases, we generated pseudo-  
124 absences in areas environmentally dissimilar to known occurrence points to avoid misplace of  
125 absences in areas with high probability of species occurrence. To accomplish that, we  
126 produced a simple initial distribution model using the BIOCLIM (Busby, 1991) and only  
127 chose pseudo-absences outside those predicted areas around the real points.

128 Two important steps are required to check how well the algorithms were successful.  
129 First, we separate the dataset into two equal-size training and testing data applying a spatial  
130 partition that subdivides the geographic extent into a checkerboard pattern, which controls the  
131 effects of spatial autocorrelation on the independency of the training and test dataset  
132 (Muscarella et al., 2014). The size of the checkerboard cells are chosen as to minimize spatial  
133 autocorrelation, by the I-Moran, and maximize model transferability, measured by the  
134 environmental similarity between the datasets using the multivariate environmental similarity  
135 surface (MESS). Second, we estimate the algorithms performance through the True Skill  
136 Statistics (TSS,  $TSS = sensitivity + specificity - 1$ ). This procedure is considered simple and  
137 intuitive in the evaluation of ENMs’ performance (Allouche et al., 2006). . Otherwise

138 To reduce the errors associated with each algorithm, we produced consensus maps to  
139 summarize the geographic information in a single layer. Despite of several ensemble methods

140 listed in the literature (Marmion et al., 2009), we opted to use only the mean of a subset of  
141 algorithms with high adjust(ed?) value ( $>$  overall mean TSS) , for both present and future  
142 scenarios. All procedures were performed using ENM\_TheMetaLand package in software R  
143 ([https://github.com/andrefaa/ENM\\_TheMetaLand](https://github.com/andrefaa/ENM_TheMetaLand)).

#### 144 2.5. Analytical procedures

145 There are some studies discussing the relation of the predictions of ENM and  
146 population parameters, such as abundance (Tôrres et al., 2012; VanDerWal et al., 2009).  
147 Those studies emphasize the theoretical expectation that the ENM may represent constrains  
148 on the maximum values of the parameter, not its mean values. In fact, environmental  
149 suitability may control for the maximum expected yield for the plant under the best  
150 conditions, but many other factor affect the observed values such as competition, pests, etc.  
151 Thus, we applied a quantile regression to estimate the relation of different quartiles in  
152 interpret this theoretical possibility. In order to test the relationship between the plant's  
153 productivity and suitability, we performed quantile regressions for each environmental  
154 dataset. This regression is often used by ecologists due the possibility of finding better  
155 relations between variables (Cade and Noon, 2003). Furthermore, the quantile regression  
156 presents better fit for outliers. Additionally, we performed ordinary least square regressions to  
157 describe a general model for these relationships.

158 In order to describe the geographic pattern of “Pequi” productivity, we used  
159 municipality as our sampling unit. Thus, we used the average suitability derived from both  
160 climates only and climate-soil ensembles of each municipality as our predictor in both  
161 quantile and the common Ordinary Least Square (OLS) procedures. We applied quantile  
162 regressions at 5%, 50%, 90% and 95% quantiles. To linearize the relation we applied a logit  
163 transformation ( $\ln\left(\frac{x}{1-x}\right)$ ) on the suitability values and productivity data.

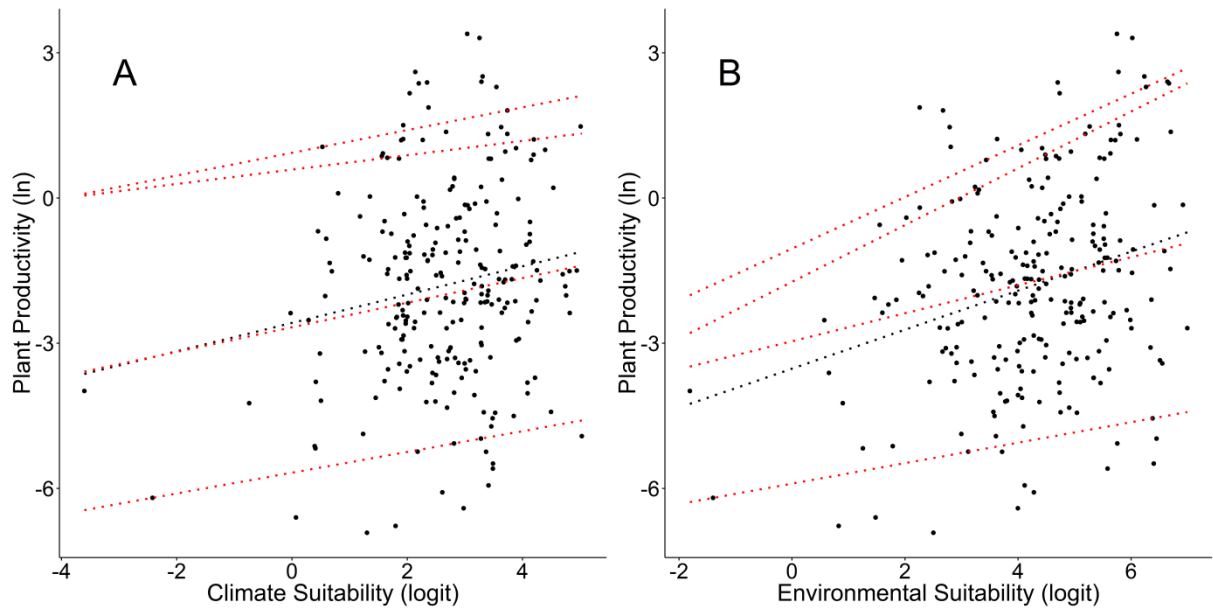
164

### 165 3. Results

166 We found that the edaphic data better explained the “Pequi” productivity (Fig. 1a,  
167 1b). Edaphic data improved the model fit compared to those with climate-only data (Table 1).  
168 The overall triangular distribution of data supports the use of the quantile regression.  
169 Specifically, there was statistically significant relation for the 90% quantile, suggesting that



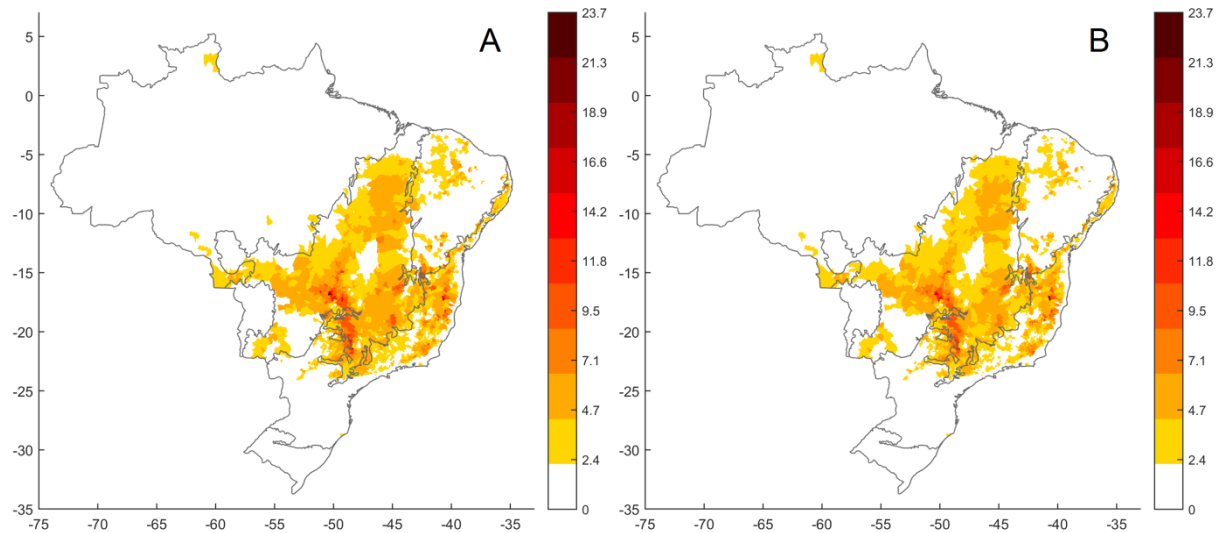
170 maximum productivity of “Pequi” fruits are related to suitability estimate using climate and  
171 edaphic data (Pseudo  $R^2 = 0.084$ ;  $p < 0.001$ ) (Table 1). However, no regressions had a strong  
172 explanation between data. Furthermore, no regressions using climate-only data for the  
173 suitability were significant (Table 1).



174

175 Fig. 1- Quantile (red lines) and ordinary least square regression (black lines) between “Pequi”  
176 fruit production and suitability predicted by climate-only (A) and climate+edaphic data (B).  
177 Both regressions were made with quantiles 0.05, 0.5, 0.9 and 0.95.

178 Despite the low fit of regressions, we used the quantile regression of 0.9 to estimate  
179 the productivity due its significant result and the greatest fit between data. We observed a  
180 small difference between the current and future models: there is a retraction of the high  
181 productivity areas, in which some municipalities will lose favorable environmental conditions  
182 for good fruit production in the future (Fig. 2). However, the geographic pattern of the  
183 “Pequi’s” productivity doesn’t change, being that the most central municipalities will remain  
184 with high values for fruit production. In addition, the distribution of productivity seems to be  
185 concentrated within the limits of the Cerrado biome. A special highlight is the existence of a  
186 prediction for “Pequi” trees in areas of Roraima (Northern region of Brazil), where it is  
187 known to occur vegetation similar to Cerrado.



188

189 Fig. 2- Productivity maps of “Pequi” fruits for current (A) and future scenario (B). These  
 190 maps were constructed using the Brazilian municipalities as sample unit. For each  
 191 municipality the average suitability was calculated using the environmental dataset with  
 192 addition of edaphic data. Then, environmental suitability was used to estimate the maximum  
 193 yield fruit production. This was done using the parameters of the quantile regression of 90%.

194 **4. Discussion**

195 Our results suggest that the models created by the association between climate and  
 196 edaphic data better explains “Pequi” productivity. In addition, this environmental dataset  
 197 improves the fit of the models mainly for the high fruit production. This result may mean that  
 198 the addition of edaphic data also improves the known tolerance range, approximating of the  
 199 species’ realized niche. The realized niche may be considered as accessible geographic areas  
 200 where the environmental conditions are favorable for the persistence of species (Vetaas,  
 201 2002). In this case, if edaphic data better describes the realized niche, is possible that, for plant  
 202 species, this type of environmental dataset may indicate more accurate suitability areas. In  
 203 addition, we found that the spatial distribution of productivity suffers a small contraction for  
 204 core areas of the Cerrado biome, indicating a small reduce of fruit production in future  
 205 scenarios. From this result, we may say that the “Pequi” tree will adapt to climate changes,  
 206 presenting a potential use by the Brazilian economy.

207 Theoretically, niche responses may determine the maximum fitness of individuals  
 208 rather than its observed mean values (VanDerWal et al., 2009). We could consider that  
 209 ecophysiological constraint on plants may determine its maximum yield under determined

210 edaphic and climatic conditions. However, no single plant must remain under these edge  
211 conditions. In fact, real plants lose part of its fitness due to a large range of possible  
212 demographic phenomena. For instance, intra-specific and inter-specific competitions are  
213 among the most suggested causes for this kind of loss. To exemplify this, distance between  
214 plants in forestry plots explain a large amount of productivity variance in a demographic  
215 phenomenon called self-thinning (Weller, 1987). Otherwise, other biological characteristics  
216 also may affect the fitness in plants. For instance, the availability of pollinators may reduce  
217 the fitness mainly in species which the self-pollination rate is low. For instance, the “Pequi”  
218 tree presents low self-pollination, depending strongly of pollinator for its reproductive rate  
219 (Gribel and Hay, 1993). Also, the presence of agricultural pests may be one of responsible  
220 factors for the low productivity in several crops. All those aspects considered small predictive  
221 power for an OLS model of mean productivity is highly expected. Otherwise, control of  
222 productivity under strict management practices (distance between plants, pest control and soil  
223 corrections) may provide productivity estimates closer to the maximum yield expected under  
224 an ecophysiological model. Unfortunately, this information is not currently available for the  
225 majority of crops, including the Pequi. Considering that the available information (used in this  
226 study) was mean productivity at the municipality level, our predictive power and observed  
227 relations should be considered as good.

228 “Pequi” productivity distribution had a minor general change from present to future  
229 climate scenarios. Although this result points to a different direction than the majority of  
230 studies in this field (Collevatti et al., 2012, 2011; Nabout et al., 2011), it may have a strong  
231 theoretical support. For example, it is expected that species with large current geographic  
232 distributions, such as Pequi, have developed ecophysiological responses for different  
233 climates. Thus, they are naturally adapted to climate changes, at least to a certain degree.  
234 Otherwise, our results show that the expectation of large changes in some studies may result  
235 from the incomplete access to plant ecological requirements by only including climate  
236 variables into the analysis. Our models including edaphic properties, which display only  
237 minor changes under future climate scenarios, represent a better description of “Pequi”  
238 production for the current scenario. The inclusion of soil variables in ENM studies has long  
239 been suggested in plant studies but only recently the existence of worldwide soil grids and a  
240 detailed evaluation of its use (Velazco et al., 2017) made this more effective. These are the  
241 main reasons to explain the differences between our results and the similar study of Nabout et  
242 al. (2011), which only used climate data to model “Pequi” distribution and found a large

243 change in future climates, especially in the southeastern area. Our study does not support the  
244 existence of those large effects, mainly due to the stabilizing effect of the soil constraining  
245 plant distribution (see chapter 1 results). Those results also suggest the need to revise other  
246 similar studies to include soil variables to better understand the real climate change effects on  
247 plant species distribution.

248 Our results about the spatial predictions indicate that main municipalities for  
249 maximum “Pequi” production are distributed within the limits of the Cerrado biome. In  
250 addition, in future scenarios it is expected only a retraction on distribution of municipalities  
251 with high potential of fruit production, indicating that “Pequi” tree presents no evident threats  
252 under conditions of future climate changes. This may imply that the “Pequi” fruit production  
253 has a potentially profitable market in Brazilian forestry.

254 The most important strategy to deal with ongoing climate change is to adapt (Adger  
255 et al., 2005). Most of international programs try to find ways to adapt our current economy  
256 and well-being to inexorable climate change consequences. The existences of some crops that  
257 are expected to not change under climate change scenarios are one of those strategies. Here,  
258 we show that “Pequi” productivity satisfy this criteria. Currently, the consumption of “Pequi”  
259 is restricted to parts of the central region of Brazil and some parts of the northeastern areas.  
260 Otherwise, popularization of its use could be easily developed in a convenient time-frame.  
261 This exercise highlights the importance to seek for other possible native plants with similar  
262 responses as potential crops to invest in Brazilian climate adaptation programs.

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## **CONSIDERAÇÕES FINAIS**

Os resultados derivados da Modelagem de Nicho Ecológico indicam que o serviço de polinização em Pequi não será fortemente afetado, pois as espécies envolvidas nessa interação possuem amplas distribuições geográficas e dificilmente haverá uma incompatibilidade espacial em cenários de mudanças climáticas. Da mesma forma, poucos municípios produtores de Pequi perderão a sua produtividade como consequência das mudanças climáticas. Isso pode indicar que essa espécie de planta é uma ótima alternativa para manter a economia de alguns municípios do Cerrado brasileiro, bem como uma fonte de recurso alimentar para a população local. Assim, esse tipo de cultivo pode ser considerado uma forte alternativa de segurança alimentar ao longo do tempo pela possibilidade de se manter em crises climáticas.