

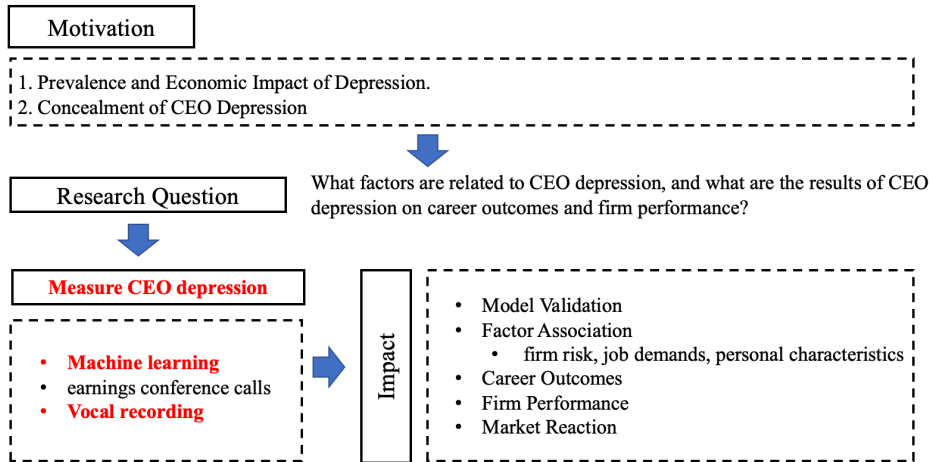
# Silent Suffering: Using Machine Learning to Measure CEO Depression

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# Design



# Comparison

## Commonalities:

- **Non-traditional Data Sources:**
  - Analyzing and predicting economic behavior or decision-making
  - (1)Digital footprints, (2)facial expressions, (3)vocal acoustic features
- **Paper2, 3**
  - Emotion analysis
- **Paper1, 2**
  - Predicting consumer default

## Differences:

- **Research Objects**
  - Digital Footprints: historical data,
  - Real-time information: facial expressions
  - CEO Depression: Earnings calls
- **Indicator Construction Methods:**
  - Paper 2,3: Machine learning
  - Paper 1: simple, easily accessible variables

## Motivation: CEO depression is high prevalence

- Depression is a pervasive condition
  - Depression and anxiety result in an estimated *12 billion lost workdays annually*, costing the global economy around *1 trillion* per year(WHO2023)
  - How to detect depression: PHQ, QIDS-SR...
- CEO Depression is high prevalence
  - *Pivotal role as decision-makers*, CEOs' emotional states significantly impact their careers, firms, and the broader economy.
  - The uncertainty and ambiguity inherent in CEOs' roles can exacerbate anxiety and depression
  - So, ***What factors are related to CEO depression, and what are the results of CEO depression on career outcomes and firm performance?***

## Motivation: CEO depression remains hidden

- CEO depression remains hidden
  - Due to stigma and concerns about professional reputation, Depression often underrecognized and undertreated.
  - Unlike more visible traits such as overconfidence, narcissism and extraversion
- Introduce a measure that captures CEO depression using CEOs' vocal acoustic features.
  - Understand the prevalence and severity of depression among CEOs.
  - Examine how CEO depression relates to career outcomes
  - Building the CEO depression measure and analyses

## Contribution: CEO depression effect

- Literature on how CEO characteristics explain variation in CEO turnover and firm investment...
  - Prior: static traits(military background, MBA degree, overconfidence, narcissism...)
  - Extend: dynamic psychological state: depression
- Understanding of managerial labor markets
  - Prior: the nonlinear effect of CEO optimism on turnover
  - Extend: quantify the association between CEO depression and career outcomes
- Examining the effect of general mood changes on economic agents' activity and decisions
  - Prior: use weather or hospital admissions as proxies for low mood
  - Extend: directly measure CEO depression from vocal acoustic features

## Contribution: introduce a measure captures CEO depression

- The impact of vocal cues in financial markets
  - Prior: how voices are perceived by investors, analysts, and auditors
  - Extend: analyze vocal acoustic features imperceptible to the human ear
- Literature on machine learning techniques in accounting and finance
  - Prior: predicting lending decisions, fundamental analysis
  - Extend: the first to use CEOs' vocal acoustic features to predict CEO depression

# Hypotheses

- Risk factors(CEO performance, firm risk, CEO linguistic patterns during conference calls, job demands, and personal characteristics and seasonal effects) is associated with CEO depression
- CEO depression is associated with career outcomes, including turnover, career prospects, TPS, compensation, and PPS
- CEO depression explains firm performance in the short or long term
- CEO depression explains abnormal returns around conference calls



# Training the depression detection model

## 1. DAIC data

- 1. American English proficiency
- 2. Gender balance
- 3. Neutral and depression-specific questions

## 2. Three subsamples

- 1. Training (107 observations, 57%)
- 2. Development (35 observations, 18%)
- 3. Testing (47 observations, 25%)

## 3. Feature extraction

- 1. Acoustic features: Extracted 6,373 ComParE and 89 GeMAPS features
- 2. YAMNet Model
- 3. PCA: Reduced feature dimensions.

No	Machine Learning Model	Input	Output	RMSE	MAE	RMSE Baseline	MAE Baseline
1	Linear regression	ComParE and GeMAPS Features	Depression Score based on the PHQ-8 Scale	6.69	5.42	7.78	5.72
2	Linear regression	YAMNet Embeddings	Depression Score based on the PHQ-8 Scale	5.83	4.87	7.78	5.72
3	Support vector regression (SVR)	ComParE and GeMAPS Features	Depression Score based on the PHQ-8 Scale	6.16	4.95	7.78	5.72
4	Support vector regression (SVR)	YAMNet Embeddings	Depression Score based on the PHQ-8 Scale	5.71	4.67	7.78	5.72
5	Convolutional neural network (CNN)	Spectrogram images of audio signals	Depression Score based on the PHQ-8 Scale	6.51	5.64	7.78	5.72

# Method for measuring CEO depression

## 1. Data Collection & Screening

- Data Source: earnings conference calls (2010-2021), focusing on the CEO presentation at the beginning for longer speech duration.
- Final Sample: 14,608 firm-quarter observations from 421 firms and 826 CEOs

## 2. Model Application

- Predict Depression Score: Input features into trained depression detection model (YAMNet-based SVR)
- Indicator : Score  $> 10$  indicates depression (1), otherwise not depressed (0)

## 3. Validation

- Event Validation: Compare depression scores from 500 calls with other events within 14 days, average difference 4.66 (close to MAE 4.67)
- Factor Analysis: Test associations with CEO depression factors
- Known Cases: Verify model predictions for CEOs with known depression

## Results 1: factors associated with CEO depression

- Female and older CEOs have a lower probability of depression
- Factors related to firm risk are positively associated with CEO depression
- Factors related to CEO job demands are negatively associated with CEO depression.

Panel G: All variables using principal components analysis

Dependent Variable =	CEO Depression Indicator (1)	CEO Depression Score (2)
CEO Performance Factor	-0.0008 (-0.1539)	0.0447 (0.7249)
Firm Risk	0.0083 (0.9116)	0.2228** (2.3782)
CEO Speech Properties Factor	0.0029 (0.4017)	0.0219 (0.3917)
CEO Job Demands Factor	-0.0049 (-0.5475)	-0.1763*** (-2.7155)
Female	-0.1553*** (-4.8398)	-1.8800*** (-7.9623)
CEO Age	-0.0032*** (-2.6725)	-0.0293*** (-2.8286)
Season	-0.0041 (-0.3234)	-0.2269 (-1.4282)

## Results 2: CEO depression and CEO career outcomes

- No evidence that CEO depression is associated with CEO turnover
- Limited evidence that CEO depression is associated with higher TPS, higher compensation, and higher pay-performance sensitivity

Dependent Variable =	<i>CEO Turnover</i>		<i>CEO Departure to Larger Firm</i>		<i>CEO Departure to Smaller Firm</i>	
	(1)	(2)	(3)	(4)	(5)	(6)
<i>CEO Depression Indicator</i>	0.0051		-0.0003		-0.0006	
<i>Mode</i>	(0.4449)		(-0.1344)		(-0.2849)	
<i>Mean CEO Depression</i>		0.0002		-0.0001		-0.0004
<i>Score</i>		(0.1752)		(-0.2837)		(-1.2131)
<i>Sales Growth</i>	0.0069	0.0073	-0.0014	-0.0014	-0.0102	-0.0102
	(0.1788)	(0.1877)	(-0.4367)	(-0.4332)	(-0.7914)	(-0.7949)

## Results 3: CEO depression and firm performance

- No evidence that CEO depression is associated with firm performance in the short term or long term

Dependent Variable =	$Return_q$		$Return_{q+1, q+4}$	
	(1)	(2)	(3)	(4)
<i>CEO Depression Indicator</i>	-0.0008 (-0.1447)		-0.0040 (-0.7426)	
<i>CEO Depression Score</i>		-0.0007 (-1.5421)		-0.0004 (-1.0368)
<i>MB</i>	0.0051*** (5.2640)	0.0051*** (5.2586)	0.0013* (1.8532)	0.0013* (1.8462)
<i>Return Volatility</i>	0.2214 (1.1687)	0.2258 (1.1903)	1.1941*** (6.9158)	1.1948*** (6.9152)

## Results 4: CEO depression and stock market reactions

- No association between CEO depression and abnormal returns around conference calls
- Investors do not detect CEO depression during these calls

Dependent Variable =	CAR[0,+1]	
	(1)	(2)
<i>CEO Depression Indicator</i>	0.0004 (0.4358)	
<i>CEO Depression Score</i>		-0.0000 (-0.5672)
<i>CEO Tone</i>	0.0420*** (9.0547)	0.0420*** (9.0467)
<i>MB</i>	0.0000 (0.0256)	0.0000 (0.0070)

## Ideas

- 相对于消极情绪，抑郁特点在于其是长期的心理疾病，因此可跟踪同一个 CEO 更加长期的发言状态数据去衡量抑郁状态
- 加入控制变量：CEO 积极、消极情绪
- 综合语调、文本、表情、动作衡量 CEO 抑郁状态，考察 CEO 抑郁状态对决策和公司未来表现影响
- 抑郁症导致 CEO 对负面结果敏感性增加，风险承担能力降低，因此可以研究 CEO 抑郁对企业创新产出的影响，极端事件发生时对企业的决策影响